1 Long-Term Atmospheric Emissions for the Coal Oil Point

2 Natural Marine Hydrocarbon Seep Field, Offshore California

3 Ira Leifer¹, Christopher Melton¹, Donald R. Blake²

4 ¹Bubbleology Research International, Solvang, CA 93463, United States

⁵ ²University of California, Irvine, Department of Chemistry, Irvine, CA 92697, United States

6 Correspondence to: Ira Leifer (Ira.Leifer@bubbleology.com)

7 Abstract. In this study, we present a novel approach for assessing nearshore seepage atmospheric emissions through 8 modeling of air quality station data, specifically, a Gaussian plume inversion model. Three decades of air quality 9 station meteorology and total hydrocarbon concentration, THC, data were analysed to study emissions from the Coal 10 Oil Point marine seep field offshore California. THC in the seep field directions was significantly elevated and Gaussian with respect to wind direction, θ . An inversion model of the seep field anomaly, THC'(θ), derived 11 atmospheric emissions. The model inversion is for the far field, which was satisfied by gridding the sonar seepage and 12 13 treating each grid cell as a separate Gaussian plume. This assumption was validated by offshore in situ offshore data 14 that showed major seep area plumes were Gaussian. Plume air sample THC was 85% methane, CH4, and 20% carbon 15 dioxide, CO₂, similar to seabed composition, demonstrating efficient vertical plume transport of dissolved seep gases. 16 Air samples also measured atmospheric alkane plume composition. The inversion model used observed winds and 17 derived the three-decade-average (1990-2021) field-wide atmospheric emissions of 83,500±12,000 m³ THC day⁻¹. 18 Based on a 50:50 air to seawater partitioning, this implies seabed emissions of 167,000 m³ THC dy⁻¹. Based on 19 atmospheric plume composition, C_1 - C_6 alkane emissions were 19, 1.3, 2.5, 2.2, 1.1, and 0.15 Gg yr⁻¹, respectively. If 20 CH₄ emissions were dispersed over the \sim 6.3 km² of 25x25 m² bins with sonar values above noise, we find 5.7 μ M m⁻ 21 ² s⁻¹. The approach can be extended to derive emissions from other dispersed sources such as landfills, industrial sites, 22 or terrestrial seepage if source locations are constrained spatially.

23

24 1 Introduction

25 1.1 Seepage and methane

26 On decadal timescales, the important greenhouse gas, methane, CH₄, affects atmospheric radiative balance far more

27 strongly than carbon dioxide, CO₂ (IPCC, 2007, Fig. 2.21), yet CH₄ has large uncertainties for many sources (IPCC,

28 2013). Since pre-industrial times, CH₄ emissions have risen by a factor of ~2.5, and after stabilizing in the 1990s and

- 29 early 2000s, resumed rapid growth since 2007 (Nisbet et al., 2019). The significantly shorter lifetime of CH₄ than CO₂
- 30 argues for CH₄ regulatory priority as emission reductions (and changes to the radiative balance) manifest more quickly
- 31 as atmospheric concentrations decrease (Shindell, Faluvegi, Bell, & Schmidt, 2005). Further impetus for a CH4 focus
- 32 is a recent estimate that 40% CH₄ emissions reductions are feasible at no net cost for the oil and gas, O&G, industry

33 (IEA, 2020), a major anthropogenic CH₄ source (IPCC, 2014). This is particularly salient given a recent estimate that

35

For 2008-2017, global CH₄ top-down emissions estimates are 576 Tg yr⁻¹; 1 Tg=10¹² g, (550-594 Tg yr⁻¹) whereas 36 bottom-up approaches find 737 Tg yr⁻¹ (594-881 Tg yr⁻¹). Anthropogenic sources for 2008-2017 were estimated at 37 38 336-376 Tg CH₄ yr⁻¹ based on bottom-up estimates. Natural sources include wildfires, wetlands, hydrates, and 39 geological seepage among others. Bottom-up estimates for natural sources are higher than top-down estimates 40 including for geological sources (Saunois et al., 2020). Geological sources (including seepage) are estimated at 63-80 Tg CH₄ yr⁻¹ with marine seepage estimated to contribute 20-30 Tg CH₄ yr⁻¹ (Etiope, Ciotoli, Schwietzke, & Schoell, 41 42 2019) or 5-10 Tg CH₄ yr⁻¹ (Saunois et al., 2020). For comparison, marine non-geological CH₄ emissions are estimated 43 at 4-10 Tg yr⁻¹. The broad range of this emissions estimate is based in part on the uncertainty in the fraction of seabed 44 emissions that reaches the atmosphere and the uncertainty in overall seabed emissions. Further complexity in assessing 45 geological seepage CH4 emissions arise because both seepage and O&G emissions source from the same geological

46 reservoirs (Leifer, 2019) and thus are isotopically similar (Schwietzke et al., 2016).

47

Seepage is where the migration of petroleum hydrocarbon gases and fluids in the lithosphere escape to the hydrosphere and/or atmosphere from the reservoir formation which underlies a capping layer that seals the formation, allowing hydrocarbon accumulation. Thus, seepage requires a migration pathway through the capping layer (Abrams, 2005), or a capping layer that eroded away leaving an outcropping of the reservoir formation.

52

53 Marine seepage is widespread in every sea and ocean (Judd & Hovland, 2007). Quantitative seepage estimates (for 54 global budgets) are limited (though growing); see Leifer (2019) review and below for more recent. Fluxes for 55 individual marine seep vents and seep areas have been reported for the Gulf of Mexico (Johansen et al., 2020; Leifer & MacDonald, 2003; Römer et al., 2019; T. C. Weber et al., 2014), the Black Sea (Greinert, McGinnis, Naudts, Linke, 56 & De Batist, 2010), the southern Baltic Sea (Heyer & Berger, 2000), various sectors of the North Sea (Borges, 57 58 Champenois, Gypens, Delille, & Harlay, 2016; Leifer, 2015; Römer et al., 2017), offshore Norway (Muyakshin & 59 Sauter, 2010; Sauter et al., 2006) offshore Svalbard in the Norwegian Arctic (Veloso-Alarcón et al., 2019), offshore 60 Pakistan (Römer, Sahling, Pape, Bohrmann, & Spieß, 2012), the arctic Laptev Sea (Leifer, Chernykh, Shakhova, & 61 Semiletov, 2017), the East Siberian Arctic Sea (Shakhova et al., 2013), the South China Sea (Di, Feng, Tao, & Chen, 62 2020), New Zealand's Hikurangi Margin (Higgs et al., 2019), the Cascadia Margin (Riedel et al., 2018), and the Coal Oil Point (COP) marine hydrocarbon seep field, hereafter COP seep field, in the northern Santa Barbara Channel, 63 64 offshore Southern California (Hornafius, Quigley, & Luyendyk, 1999), and for numerous individual vents in the field 65 (Leifer, 2010).

66

67 Most seep emission estimates are snapshot from short-term field campaigns. Seep emissions vary on timescales from

tidal (Leifer & Boles, 2005; Römer, Riedel, Scherwath, Heesemann, & Spence, 2016) to seasonal (Bradley, Leifer, &

69 Roberts, 2010) to decadal (Fischer, 1978; Leifer, 2019). Additional temporal variability arises from transient emissions

half of recent CH₄ increases are from the O&G industry (Jackson et al., 2020).

- pulses lasting seconds to minutes (Greinert, 2008; Schmale et al., 2015) to decades (Leifer, 2019). This shortcoming
is being addressed by benthic (seabed) observatories and cabled observatories, e.g., Wiggins, Leifer, Linke, and
Hildebrand (2015); Greinert (2008), Kasaya et al. (2009); Römer et al. (2016); Scherwath et al. (2019). Still, benthic
observatories are costly and thus uncommon.

74

75 Seepage contributes to oceanographic budgets and to a lesser extent, atmospheric budgets due to water column losses 76 with significant uncertainty in the partitioning. As a result, uncertainty in the atmospheric contribution is much larger 77 than the (significant) uncertainty in seabed emissions. Seepage partitioning between the atmosphere and ocean - where 78 microbial degradation occurs on timescales inversely related to concentration (Reeburgh et al., 1991), which depends 79 primarily on depth (Leifer & Patro, 2002) with little to none of deepsea seabed emissions reaching the atmosphere, 80 e.g., Römer et al. (2019). In contrast, very shallow seepage (meter scale) largely entirely reaches the atmosphere both 81 by direct bubble-meditated transfer and diffusive transport. For intermediate depths, the ocean/atmospheric 82 partitioning is complex and depends on depth, bubble flux, bubble size distribution, bubble interfacial conditions, and 83 other characteristics (Leifer & Patro, 2002). Whereas the indirect diffusive flux (proximate and distal) depends on 84 bubble dissolution depth (Leifer & Patro, 2002), turbulence vertical transport in the winter wave-mixed layer (Rehder,

85 Keir, Suess, & Rhein, 1999), microbial oxidation losses, and exchange through the sea-air interface.

86

A range of approaches have been used to estimate the sea-air flux. The most common is by measuring the atmospheric
and water concentrations and applying air-sea gas exchange theory for the measured wind speeds, e.g., Schmale,
Greinert, and Rehder (2005) for Black Sea seepage under weak wind speeds.

90

Sea-air exchange is a diffusive turbulence transfer process that depends on the air-sea concentration difference and the piston velocity, k_T , which depends on gas physical properties, wind speed, *u* (Liss & Duce, 2005), wave development (Zhao, Toba, Suzuki, & Komori, 2003), wave breaking (Liss & Merlivat, 1986), and surfactant layers at low wind speeds that suppress gas exchange (Frew et al., 2004). k_T increases rapidly and non-linearly with *u* and has been parameterized as piecewise linear functions (Wanninkhof, Asher, Ho, Sweeney, & McGillis, 2009) or as a cubic function (Nightingale et al., 2000). Air-sea gas exchange theory is for (relatively) homogeneous atmospheric and oceanographic fields (concentrations, winds, and wave development), and thus is inappropriate for point-source

- 98 (bubble-plume) emissions and for the near-field downcurrent plume, which tend to be heterogeneous.
- 99

Another approach uses seabed bubble size measurements or an assumed bubble size distribution to initialize a numerical bubble propagation model to predict direct bubble-mediated atmospheric fluxes (Leifer et al., 2017; Römer et al., 2017; Schneider von Deimling et al., 2011). The dissolved portion that evades to the atmosphere could be addressed by a dispersive model coupled to an air-sea gas exchange model, though studies have not yet addressed this component.

106 An alternate approach is to derive atmospheric emissions by plume inversion. Leifer, Luyendyk, Boles, and Clark

107 (2006) derived emissions for a blowout from Shane Seep in the COP seep field by a plume inversion. This neglected

- 108 the portion that dissolves during bubble rise and drifts downcurrent, out of the bubble plume's vicinity before sea-air
- 109 gas transfer into the atmosphere. Note dissolved gas evasion in the plume vicinity contributes to the inversion
- 110 emissions estimate.
- 111

112 **1.2 Study motivation**

113 In this study, we present a novel approach for assessing nearshore seepage atmospheric emissions – air quality station 114 data modeling, specifically using a Gaussian plume inversion model. This model requires that source locations are 115 mapped, spatially stable, and lie within a fairly constrained distance range band. These conditions are met for the COP seep field, which is near the West Campus air quality Station (WCS). COP seep field lies in shallow coastal waters of 116 117 northern Santa Barbara Channel, CA. Spatial constraint is provided by geological structures, such as faults, that 118 constrain emission locations. The Gaussian plume model assumes the source is in the far field, whereas WCS is in the 119 nearfield for the extensive COP seep field. To satisfy the far field criterion, the source was gridded and each grid cell's 120 emissions treated as a distinct (distant) Gaussian plume. This characterization was validated in an offshore survey of 121 several focused COP seep field seepage areas, which were well-modeled as Gaussian plumes.

122

Thus, this study demonstrates an approach to deriving emissions from air quality station data for an area source such as natural marine seepage. This approach could be used to derive emissions from other dispersed sources such as

- 125 landfills, industrial sites, or natural terrestrial seepage where the source locations can be constrained spatially.
- 126

127 **1.3 Water column marine seabed seepage fate**

Seep seabed CH₄ partitioning between the atmosphere and water column depends on seabed depth and emission character – as bubbles, bubble plumes (Leifer & Patro, 2002), or dissolved CH₄. Dissolved CH₄ migration through the sediment is oxidized largely by near seabed microbes (Reeburgh, 2007), termed the microbial filter, negating its contribution, leaving only bubble-mediated flow.

132

As seep bubbles rise, they dissolve, losing gas to the surrounding water at a rate that decreases with time; smaller and more soluble gases dissolve faster than larger and less soluble gases, i.e., fractionation (Leifer & Patro, 2002). Additionally, larger bubbles transport their contents upwards more efficiently than smaller bubbles (Leifer et al., 2006). Sufficiently large bubbles reach the sea surface with a significant fraction of their seabed CH₄ from depths of even hundreds of meters (Solomon, Kastner, MacDonald, & Leifer, 2009). There are synergies, too with higher plume fluxes driving a stronger upwelling flow that transports plume fluids with dissolved gases upwards towards the surface

139 where air-sea gas exchange drives evasion (Leifer, Jeuthe, Gjøsund, & Johansen, 2009). Another synergy arises from

elevated dissolved plume CH₄ concentration (Leifer, 2010; Leifer et al., 2006), which slows dissolution. Also, bubbles
are oil-coated, which slows dissolution.

142

Moreover, gases in bubbles that dissolve in the wave-mixed layer (or reach it by the upwelling flow) then diffuse to the air-sea interface due to wave and wind turbulence. Note, microbial degradation removes a portion of the dissolved CH₄, which therefore never reaches the air-sea interface. Thus, there are two timescales that govern the fraction that evades – the microbial degradation timescale, which decreases as concentrations increase, and the diffusion timescale, which decreases with increasing wind speed. As a result, there is a dissolved plume that drifts downcurrent, from

- 148 which evasion creates a linear-source atmospheric plume, with dissolved plume concentrations slowly decreasing with 149 time (downcurrent distance) from sea-air gas exchange losses, microbial oxidation, and dispersion.
- 150

151 **1.4 Atmospheric Gaussian plumes**

Strong focused atmospheric plumes are created from the seep plume bubble bursting at the sea surface and from dissolved gas evasion within the bubble surfacing footprint. This evasion is enhanced by water-side turbulence from rising and bursting bubbles (Leifer et al., 2015). Atmospheric plume evolution is described by the Gaussian plume model (Hanna, Briggs, & Hosker Jr., 1982), which relates downwind concentrations to wind transport and turbulence dispersion and is the basis of the inversion calculation (see Supp. Sec. S1 for details).

157

158 **1.5 Setting**

159 **1.5.1 The Coal Oil Point seep field**

The COP seep field (**Fig. 1**) is one of the largest seep fields in the world, with estimated 1995-1996 seabed emissions, E_B , of $1.5 \times 10^5 \pm 2 \times 10^4$ m³ THC dy⁻¹ (Hornafius et al., 1999). Hereafter emissions and concentrations are for total hydrocarbon, THC, unless noted. Clark, Washburn, Hornafius, and Luyendyk (2000) estimated that half the COP seep field E_B reach the atmosphere in the near field. This is due to shallowness, bubble oiliness, high plume bubble densities, and turbulence mixing within the wave mixed layer.

165

Geological structures play a critical role in the spatial distribution of seepage (Leifer, Kamerling, Luyendyk, & Wilson, 2010), which lies along several trends in waters from a few meters to ~85 m deep. These trends follow geologic structures including anticlines, synclines, and faults in the reservoir formation, the Monterey Formation. Faults provide migration pathways with seepage scattered non-uniformly along the trends, including focused seep areas that are highly active, localized, and often are associated with crossing faults and fractures (Leifer et al., 2010). Seepage in these areas typically surrounds a focus and decreases with distance, primarily along linear trends (Leifer, Boles, Luyendyk, & Clark, 2004). See **Supp. Table S3** for informal names and locations of selected focused seep areas.

174 1.5.2 Coal Oil Point seep field emissions and composition

175 COP seep field sources from the South Ellwood oil field whose primary source rock is Monterey Formation, which is

176 immature to marginally mature. Petroleum gases from marine organic materials have relatively higher proportion of

ethane, propane, butane, etc., relative to methane as compared to petroleum gases from terrestrial organic materials. 177

178 The wet gas fraction (C_2-C_5/C_1-C_5) indicates a thermogenic origin of greater than 0.05 (Abrams, 2017). Of the 179 saturated alkanes, the alkenes (olefins) are of biological origin. Additionally, the ethane/ethene ratio and

180

propane/propene ratios can be indicators of seep gas biogenic modification with values above 1000 indicating purely

- 181 thermogenic origin (Abrams, 2017; Bernard, Brooks, & Zumberge, 2001).
- 182

183 In this study, we analyse WCS (located at 34° 24.897'N, 119° 52.770'W) atmospheric THC. Clark, Washburn, and

184 Schwager (2010) report average seep field seabed CH₄, CO₂, and non-methane hydrocarbons (NMHC), of 76.7, 15.3,

185 and 7.7%, respectively, with Trilogy Seep seabed compositions of 67, 21, and 7.8%, respectively. With respect to

186 alkanes, seabed bubbles are 90.4% CH₄ and 8.6% NMHC. CO₂ rapidly escapes the bubbles and is negligible (<1%)

187 at the sea surface. At the sea surface, CH_4 in bubbles is ~90% with NMHC making up the remaining 10%, neglecting

air gases (Clark et al., 2010). Note, whereas seep THC is predominantly CH4, THC from terrestrial directions arises 188

189 from NMHC from traffic and other anthropogenic sources as well as CH₄ from pipeline leaks, terrestrial seeps, etc.

190

191 1.5.3 Northern Santa Barbara Channel climate

192 Diurnal and seasonal wind cycles are important to the atmospheric transport of COP seep field emissions. The Santa 193 Barbara climate is Mediterranean with a dry season and a wet seasons when storms occur infrequently (Dorman & 194 Winant, 2000). The semi-permanent eastern Pacific high-pressure system plays a dominant controlling role in weather 195 in the Santa Barbara coastal plain. This high-pressure system drives light winds and strong temperature inversions that 196 act as a lid that restricts convective mixing to lower altitudes. The coastal California boundary layer is shallow, 0 to 800 m (Edinger, 1959), generally 240-300 m around Santa Barbara (Dorman & Winant, 2000). Additionally, coastal 197 198 mountains provide physical barriers to transport (Lu, Turco, & Jacobson, 1997).

199

200 As a coastal environment, the land/sea breeze is important to overall wind-flow patterns with weak offshore night 201 winds and stronger onshore afternoon winds (Dorman & Winant, 2000). In coastal Santa Barbara, warming on 202 mountaintops and more interior arid lands relative to cooler marine temperatures drives the sea breeze. Downslope 203 nocturnal flows warm nocturnal surface temperatures, moderating the coastal diurnal temperature cycle (Hughes, Hall, 204 & Fovell, 2007).

205

206 Typical morning winds are calm and offshore and often accompanied by a cloud-filled marine boundary layer, 50-207 150 m thick (Lu et al., 1997). The marine layer usually (but not always) "burns off" mid-morning after which temperatures rise, the boundary layer thickens and winds shift clockwise from offshore to eventually prevailing 208

- 209 westerlies aligned with the coastal mountains. Midday through late afternoon and even evening, winds strengthen,
- 210 often leading to whitecapping before the boundary layer collapses and winds resume the nocturnal pattern.
- 211

212 2 Methods

213 2.1 West Campus Station data

WCS data includes wind speed, u, and direction, θ , by a vane anemometer (010C,020C, Met One, Grants Pass, OR) and THC concentration, C, by a Flame Ionization detector (51i-LT, Thermo Scientific, MA). WCS is maintained by the Santa Barbara County Air Pollution Control District. Daily instrument calibration occurs after midnight, rendering C unavailable 00:50 to 02:09 local time, LT. WCS was improved significantly in 2008 from 1-hour to 1-minute time resolution, which allowed far higher values of C and u due to the shorter averaging times. Data analysis uses custom routines as well as standard routines and functions in MATLAB (MathWorks, MA).

220

First, WCS data were quality controlled to remove all values of *C* during the daily calibration, as well as to interpolate neighboring values that were unrealistically low, i.e., *C* less than 1.6 ppm in the 1990s and 1.85 ppm in the 2000s. Data since 2008 were smoothed by nearest-neighbor averaging, yielding 3-minute time resolution. Data prior to 2008 were hourly and were not smoothed. Wind data were nearest-neighbor averaged after decomposing into north and east components, followed by recalculation of *u* and θ .

226

227 2.2 In situ marine surveys

Offshore *in situ* survey data were collected by the *F/V Double Bogey*, a 12-m, 9-ton, fishing vessel with a near waterline deck (~0.2 m) and low overall profile (cabin at ~2.2 m). A sonic anemometer (VMT700, Vaisala) was mounted on a 6.5-m tall, 5-cm (2") diameter aluminum mast and measured 3D winds. Continuous, CH₄ and CO₂ data were collected 5 Hz by a Cavity Enhanced Absorption Spectroscopy (CEAS) analyzer (FGGA, LGR Inc., San Jose, CA). Vessel location and time were from a Global Positioning System (GPS) at 1 Hz (19VX HVS, Garmin, KS). CH₄ and CO₂ calibration with a greenhouse gas air calibration standard (CH₄: 1.981 ppmv; CO₂: 404 ppmv, Scott Marin,

- 234 CA, purchased 2015, Sigma Aldritch, St Louis, MO).
- 235

Data are real time integrated and visualized in Google Earth on a portable computer (Spectre360, HP) using custom software, written in MATLAB (MathWorks, MA) for AutoMObile trace Gas (AMOG) Surveyor, described elsewhere (Leifer, Melton, Fischer, et al., 2018; Leifer, Melton, Manish, & Leen, 2014; Leifer, Melton, Tratt, et al., 2018; Leifer et al., 2016). Real-time visualization facilitates adaptive surveys, wherein the survey route is modified based on realtime data to improve outcomes (Thompson et al., 2015) - in this case to facilitate plume tracking and to ensure transects

- 241 were near orthogonal to the wind.
- 242

- 243 Accurate, absolute winds are calculated from relative winds after accounting for vessel motion and filtering for non-
- 244 physical velocity changes due to GPS uncertainty (Leifer, Melton, Fischer, et al., 2018). Filtering removes transient
- winds that are not relevant to plume transport. The filter interpolates GPS positions flagged as unrealistic.
- 246
- 247 Whole air samples were collected in evacuated 2-liter stainless steel canisters, which were filled gently over ~1 minute
- from ~1 m above the sea surface. The filled canisters were analyzed in the Rowland/Blake laboratory at the University
- of California, Irvine for carbon monoxide, CO, CH₄, and C₂-C₇ organic compounds. Samples were analysed by a gas
- 250 chromatography multi-column/detector analytical system utilizing flame ionization detection.
- 251

252 **2.3 Seep plume emissions model**

253 The plume inversion model is a three-step process (Leifer, Melton, Fischer, et al., 2018; Leifer, Melton, Tratt, et al., 2018; Leifer et al., 2016). Emissions from focused seep areas were derived from offshore data by first fitting Gaussian 254 function(s) to orthogonal transect C' data, termed the data model. C' is relative to C outside the plume, derived by 255 256 linear interpolation across the plume transect. The data model is derived by error minimization using a least-squares 257 linear-regression analysis (Curve fitting toolbox, MathWorks, MA). Next, the Gaussian plume model (Eqn. S1; Supp. 258 Figs. S1; S2) is fit to the data model. Transect data are collected close to orthogonal to the wind direction and are projected in the wind direction onto an orthogonal plane. See Leifer, Melton, Tratt, et al. (2018) for a validation study 259 260 of the plume inversion model by comparison with remote sensing-derived emissions (which are largely insensitive to 261 transport). The study found in situ and remote-sensing derived emissions agreed within 11%.

262

263 **2.4 Seep field emissions model**

The inversion model is based on gridding the seep field into numerous small additive Gaussian plumes that represent the area emissions and was written in MATLAB (MathWorks, MA). This assumes that each sea-surface grid cell contributes a Gaussian plume, an assumption that was tested with offshore survey data downwind of several active seep areas.

268

The definition of area versus point source depends on the relevant length scales – an area source is well approximated as a point-source plume if sufficiently downwind (far field), where the distance for "sufficiently downwind" depends on the area source dimensions and meteorological conditions. Whereas WCS is near field for the entire seep field plume, it is far field for the small plumes from each grid cell.

273

The area source was based on a Sept. 2005 sonar survey sonar return, ω , map (Fig. 1), see Leifer et al. (2010) for sonar survey details. Simulations used sonar data gridded at a hybrid 22/56-m in a UTM coordinate system, with origin at WCS. Specifically, gaps in the 22-m map were filled from the 56-m map (**Supp. Fig. S3**). The probability distribution of ω was used to identify the noise level (**Supp. Fig. S4**) as in Leifer et al. (2010).

279 The model calculates a Gaussian plume for E_{ij} for grid cell *i* and *j*, for each grid cell with ω above noise. $C'_{Sim}(x, y)_{ij}$ are calculated for the observed $u(\theta)$ in wind direction θ and a typical Santa Barbara channel boundary layer, BL=250 280 m. The initial $E_{i,i}$ is by scaling such that the integrated sonar return ($\int \omega(\mathbf{x},\mathbf{y})$) scales to $E_A=1.5 \times 10^5$ m³ dy⁻¹, i.e., E_B from 281 282 Hornafius et al. (1999). The Gaussian plume is calculated in a Cartesian coordinate system (Supp. Fig. S5A), rotated 283 to θ , and the interpolated linearly to double the spatial resolution. Then, the rotated plume is regridded to UTM 284 coordinates using the ffgrid.m function (Supp. Fig. S5B). Interpolation removes gaps in the regridded plume map. 285 Then, the regridded plume is renormalized to ensure total mass is conserved before and after these operations. Rotated regridded plumes are translated to the seep field grid and added, yielding $C'_{sim}(x, y)$, the simulated seep field plume 286 287 anomaly (Supp. Fig. S5C).

288

289 The model scans θ for the seep directions (110°< θ <330°) and calculates the simulated plume anomaly, $C'_{sim}(\theta)$ at

WCS, which is compared with the observed $C'_{Obs}(\theta)$ at WCS. Hereafter, C_{Obs} and C_{Sim} and their anomalies refer to values at WCS. $C'_{Obs}(\theta)$ is defined:

292
$$C'_{Obs}(\theta) = C_{Obs}(\theta) - \min(C_{Obs}(\theta))$$
(1)

with the minimum typically from the west in a direction with no known seepage. Specifically, $C'_{obs}(\theta)$ was calculated by subtracting the minimum in the annualized observed $C'_{obs}(t,\theta)$ each year, *t*, after applying a 7-year running average.

296

Emissions from suburban communities, light industry, and commercial centers enhance $C'_{Obs}(\theta)$ for the north to east (~350-70°) sectors. Removal of these terrestrial emissions was by fitting a Gaussian function to $C'_{Obs}(\theta)$ for 330°< θ <30° with the residual yielding $C'_{Obs}(\theta)$. This only affected $C'_{Obs}(\theta)$ for overlapping directions corresponding to the fields' eastern edge.

301

Simulations were run at angular resolutions of 2°. Higher angular resolution produced small artifacts for the 22/56-m
 sonar grid while the 11-m sonar grid was overly sparse due to the distance between sonar tracks (Supp. Fig. S3).
 304

The source map is ω in units of decibels, whereas emissions are in units of moles m⁻² s⁻². Given that the relationship between ω and bubble density (emissions) is complex and non-linear (Leifer et al., 2017), there is poor agreement

between $C'_{sim}(\theta)$ and $C'_{obs}(\theta)$. Thus, a correction function, $K(\theta)$, is applied to emissions for each grid, E(i,j), along each θ and the model rerun. $K(\theta)$ is defined,

309
$$K(\theta) = \frac{C'_{Obs}(\theta)}{C'_{Sim}(\theta)}$$
(2)

Initially, K=1, but in subsequent iterations, $K(\theta)$ is scaled as in **Eqn. 2** to adjust E_A in cells along θ . Because $K(\theta)$ weights closer seeps more than more distant seeps, a distance-varying correction function, $K(r,\theta)$, was calculated such that,

313
$$\int_{r=0}^{r=\infty} E_A(r,\theta) = \int_{r=0}^{r=\infty} K(r,\theta) E_A(r,\theta) dr$$
(3)

where *r* is distance from WCS. Simulations that shifted WCS northwards showed E_A varied nearly linearly with distance. Accounting for off-axis plume contributions requires several iterations to achieve *Convergence*, which was defined,

317 Convergence =
$$\frac{\sum C'_{sim}(\theta) \sum C'_{obs}(\theta)}{\sum C'_{obs}(\theta)}$$
 (4)

318 Iterations continued to *Convergence* of 1% or better – typically 4 to 5 iterations. Simulations suggest wind veering, 319 ψ , was important, which was implemented by calculating $C'(\theta)$ and assigning it to $C'(\theta + \psi)$.

320

321 3 Results

322 **3.1 Offshore** *in situ* surveys

An offshore COP seep field survey measured *in situ* C_{CH4} and *u* on 28 May 2016. Data were collected from the Santa Barbara harbor (~7.5 km east of the seep field, **Fig. 2A; Supp. Fig. S6**) to offshore Naples, several kilometers west of the seep field. Overall winds were easterly with an onshore component near Campus Point and a broad (6-km wide)

326 offshore flow west of COP that shifts to along coast near Naples (Fig. 2A, white arrows). Observed winds veered

- $\sim 10^{\circ}$ from east to the west sides of the seep field, roughly comparable to the shift in coastline orientation.
- 328

Plumes are apparent downwind of major seeps, with the largest plume associated with the Trilogy Seep (**Fig 2B**). Strong plumes also are evident downwind of the La Goleta Seep and Patch Seep. Notably, the Seep Tent Seep plume was very weak. The Seep Tent Seep was the dominant seep area in the COP seep field from its appearance in June

332 1973 (Boles, Clark, Leifer, & Washburn, 2001) until recent years.

333

Additionally, the offshore survey identified focused plumes from beyond the extent of the seep field's 2005-sonar map. Specifically in the Goleta Bay, which has been noted (Jordan et al., 2020), and offshore Haskell and Sands

- Beaches, areas of abandoned oil wells, and off Naples Point (**Fig. 2A**, red arrow).
- 337

338 Plume alkane C' were determined by the difference between two "background" air samples collected immediately

- outside the plume and three Trilogy Seep plume air samples. CH4 was 88.5% of THC, with ethane, propane, and
- butane at 3.1%, 4.2%, and 2.76%, respectively, with pentane, hexane, and heptane at 1.11, 0.13, and 0.04%,
- 341 respectively (**Table 1**). THC molecular weight is 19.6 g mole⁻¹ based on a composition weighting. Branched alkanes

- were detected, with 2-methylpentane and 3-methylpentane comprising 0.21%, each, as well as simple aromatics, e.g.,
- benzene and toluene, with concentrations of 0.044 and 0.100 ppm, respectively.
- 344

The observed wet gas fraction, $\sum_{n=2}^{5} C_n / \sum_{n=1}^{5} C_n$ was 0.11 indicating a thermogenic origin - greater than 0.05

- (Abrams, 2017) and thus derived from marine organic materials. Although the olefins ethene and ethyne were
 detectable at 0.02% and 0.004%, respectively, butene was not detected. These olefins primarily derive from microbial
- 348 processes (Abrams, 2017), thus, the ethane/ethyne ratio of 6200 also strongly indicates a thermogenic source (Bernard
- 349 et al., 2001). Atmospheric CO₂ was elevated by 12 ppm. Given that CO₂ completely dissolves from bubbles well
- before reaching the sea surface (Clark et al., 2010), this demonstrates efficient vertical transport of dissolved seep
- 351 gases to the sea surface.
- 352
- 353 Plumes for the Trilogy Seeps, La Goleta Seep, and Seep Tent Seep were inverse modeled to derive emissions for each

plume. For the Trilogy Seeps, the average u across the plume was 5.9 m s⁻¹, insolation was full sun, and the source

- height was set at 25 m based on Trilogy's atmospheric plume being buoyant. Model surface concentrations for Trilogy
- B plume are shown in **Fig 2A**. The other two seeps are far less intense and used a 1-m source height.
- 357

E for Trilogy A was 1.28 Gg CH₄ yr⁻¹ (5600 m³ CH₄ dy⁻¹), whereas Trilogy B and C contributed 0.06 and 0.07 Gg 358 $CH_4 \text{ yr}^{-1}$, respectively, for a total of 6200 CH₄ m³ dy⁻¹. Note, plume origins and the sonar seep bubble plume locations 359 360 do not precisely match because the sonar map is for near the seabed, and currents deflect the bubble surfacing location, 361 up to ~40 m. La Goleta Seep released 4000 m³ CH₄ dy⁻¹ and the Seep Tent Seep released 310 m³ CH₄ day⁻¹ with 362 almost no surface bubble expression. For comparison, Clark et al. (2010) used a flux buoy, which measures near surface bubble fluxes, and found Trilogy Seep emissions of 5500 and 4200 m³ THC dy⁻¹ and 930 m³ THC day⁻¹ for 363 La Goleta Seep in 2005 and 5700 m³ THC dy⁻¹ for the Seep Tent Seep in 2002. During the cruise, surface bubble 364 plumes were not observed for the Seep Tent Seep, although its bubble plume had been a perennial and dominant 365 366 feature since its appearance. Note, Clark et al. (2010) reported THC in near sea surface bubbles was 91% CH4.

367

368 3.2 West Campus Station

369 3.2.1 Temporal trends

WCS is 500 m from the coast (to the southwest) at 11-m altitude and 850 m almost due south to the 11-m altitude bluffs of Coal Oil Point (**Fig. 1**). Terrain slopes gently towards the coast to the southwest and towards a lagoon to the south-southeast, rising again to the southeast to the COP bluffs. This flat relief likely has small to negligible effect on

- 373 wind speed and direction, although differential land-ocean heating could influence winds. Wind veering for the coast
- to the east of COP is likely due to the orientation of the coastline and bluffs.
- 375

The WCS improvements in 2008 (Fig. 3-dashed line) allowed far higher values of C and u (Supp. Fig. S7A,7B).

377 Comparison of the probability distributions of u and C, $\phi(u)$ and $\phi(C)$, respectively, before and after the upgrade did

- not suggest biases were introduced (Supp. Fig. S7C,7D). Specifically, changes in the average and median values and
 in the baseline after 2008 were from better measurement of higher value events (gusts and short positive *C* anomalies).
- 380

- 381 Significant daily, seasonal, and interannual variations are apparent in the day-averaged *u* and *C* (Fig. 3). The calmest
- 383 Winds have strengthened since a minimum in 1995-1996, moreso for the seep directions with stronger winds becoming

season is late summer to fall, whereas spring is the windiest and most variable due to synoptic systems (Fig. 3A).

- more frequent and moreso for summer than winter (Supp. Figs. S8, S9).
- 385

386 Trends in C reflect trends in both seep field emissions and ambient C. C is higher in fall and spring (Fig. 3B). Given

that stronger winds decrease C through dilution, this suggests the seasonal variation in C underestimates the seasonal

variation in emissions. Several studies have shown increased emissions under higher wave regimes (storminess),

389 reviewed in Leifer (2019) and proposed from wave pumping. Storms increase evasion from higher wave turbulence

- 390 and breaking-wave bubbles, which sparge dissolved CH₄ and other trace gases down to the seabed in shallow (<100
- m) waters (Shakhova, Semiletov, Salyuk, et al., 2010). Note, u, θ , and C' correlate with time of day. For example,
- 392 north generally reflects weak, offshore nocturnal winds with no seep contribution.
- 393

394 3.2.2 Spatial heterogeneity

Calculating the angular-resolved average C, $C_{ave}(\theta)$, for the complete dataset with respect to θ shows the highest Cfrom the main seep field direction (155-250°, **Fig. 4**). For the seep directions, $C_{ave}(\theta)$ was poorly fit by a single Gaussian function but well fit (R^2 =0.997) by two Gaussian functions with peaks at 178° and 198° corresponding to the Seep Tent and Trilogy Seeps' directions, respectively (**Fig. 4A, 4B**). Notably, the fit residual showed a linear increasing trend, $dC_{ave}(\theta)/d\theta$, of 0.17 ppb degree⁻¹ from 180 to 210° (**Supp. Fig. S9B**) consistent with evasion from a dissolved downcurrent plume that drifts west-northwest along the coast (Leifer, 2019).

401

The average *C* anomaly, $C'_{ave}(\theta)$, was calculated from the average of $C_{Obs}(\theta)$ after **Eqn. 1** with terrestrial anthropogenic sources to from the north to northeast removed. The minimum in $C_{Obs}(\theta)$ was at 270°, a direction with no mapped seepage that also is at the dissolved plume's approximate shoreward edge. **Fig. 4A** shows $C_{Obs}(\theta)$ before removal of terrestrial emissions, which do not overlap in any significant manner with seep field emissions.

406

There is a strong, focused peak in $C_{max}(\theta)$ at $\theta \sim 190^\circ$, close to the Seep Tent Seep direction (**Fig. 4E, 4F**), which is fairly isolated on the Ellwood Trend (**Fig. 1**). This peak also is close to the direction of Tonya Seep on the inshore seep trend and to the small, unnamed area of seepage to the west of Trilogy Seep along the Red Mountain Fault trend. The θ -resolved maximum $C(\theta)$, $C_{max}(\theta)$, remains elevated through $\sim 270^\circ$, far west of the $C_{ave}(\theta)$ peak at $\sim 200^\circ$. This strongly suggests that the seep field extends further to the west-northwest than current maps. These data cannot be explained by dissolved plume outgassing, which would affect $C_{ave}(\theta)$ but not $C_{max}(\theta)$.

- 414 $C(\theta)$ enhancements for non-seep directions (**Fig. 4A,4B**) show a peak at ~35°, corresponding to the direction of a 415 commercial center amid suburban development. This could result from terrestrial seepage and natural gas pipeline 416 leakage and/or THC emissions from communities and traffic.
- 417
- 418 Neglecting the synoptic system, topographic forcing from the east-west Santa Ynez range means that prevailing winds
- 419 are westerlies and are the strongest (Fig. 4C, 4D). North winds (320-15°) largely are weak as are winds from due
- south; however, the sea breeze strengthens winds rapidly away from due south. θ peaks in the maximum winds (1-
- 421 minute sustained), $u_{max}(\theta)$, correspond to the west and east peaks in $u_{ave}(\theta)$ with strengths to 16 m s⁻¹. Interestingly,
- 422 there also are strong north (0-30°) winds or downslope flow, termed sundowner winds, a highly localized and
- 423 infrequent phenomenon. The overlap of $u_{med}(\theta)$ and $u_{ave}(\theta)$ shows winds largely are normally distributed.
- 424

425 The median C, $C_{med}(\theta)$, and average C, $C_{ave}(\theta)$, have similar shapes, albeit with lower values at all θ (Fig. 4A),

426 indicating *C* is not normally distributed. This is shown in the wind direction-resolved wind speed probability 427 distribution, $\phi(\theta, u)$ (Fig. 5A), defined such that

428
$$\int \phi(\theta, u) du = 1, \quad \int \phi(\theta, C) dC = 1$$
 (5)

429 $\phi(\theta, u)$ is very narrow (y-axis) for the northeast (~45°) where winds are largely weak and broad for the east-southeast 430 (70-135°) and the prevailing westerlies (250-280°). The east-southeast distribution skews to the south (stronger winds 431 extend further from the south - offshore), whereas the prevailing westerly wind distribution skews to the northeast (as 432 does the coastline).

433

In the seep direction, $\phi(C,\theta)$ extends to much higher values than from non-seep directions (Fig. 5B). $\phi(C,\theta)$ is asymmetric with θ extending further to the west than the seep field extent (240°) and then decreasing more abruptly than the decrease to the east. This asymmetry is expected given the seep field's asymmetric orientation relative to WCS (eastern seepage is more distant). Emissions beyond the field's mapped western edge arise from downcurrent plume outgassing and potentially contributions from unmapped seeps.

439

440 **3.2.3 Seep field diurnal emissions cycle**

441 C and u for the seep field direction, u_{seep} , and C_{seep} , respectively, follow diurnal patterns that are not the same as the 442 overall diurnal pattern due to the wind direction constraint and because C_{seep} depends on u_{seep} . The dependency arises 443 because higher u dilutes emissions, decreasing C, but higher u also increases dissolved plume evasion and bubble-444 mediated emissions from higher swell (after a delay for wave buildup). Diurnal winds in coastal regions feature a shift 445 between weak nocturnal offshore winds that veer to onshore winds in the morning - the sea breeze circulation. This 446 was explored in time and direction segregated u and C and seep direction averaged u_{seep} , and C_{seep} for 90-270° (Fig. 6). Data were segregated by θ for pre- and post-2008 (when station improvements facilitated better wind 447 448 characterization, particularly for night winds, which are seldom from the seep field direction, see Supp. Fig. S10 for

- 449 1991-2007). $u(\theta,t)$ and $C(\theta,t)$ were 2D Gaussian kernel smoothed with a 1-bin standard deviation (contours based on 450 a 3 bin standard deviation) by the imgaussfilt.m algorithm (MATLAB, MathWorks, MA) after interpolating the 451 calibration data gap 24:00-01:00.
- 452

Early morning (01:00–03:00) u_{seep} are stronger because typical nocturnal winds are northerlies (land breeze), coming from the south largely during storms. These are accompanied by elevated C_{seep} implying greater emissions despite enhanced dilution from stronger winds. The minimum in both u_{seep} and C_{seep} occur in the early morning (04:00-08:00), with both increasing slightly through midday (~12:00). C_{seep} follows an afternoon trend of an overall decrease to a minimum at ~20:00 before increasing into the late evening.

458

Underlying these trends are complex temporal spatial patterns. u for the north to northeast reaches a maximum around noon and peak around 16:00; while C for northeast to east is low in the morning reaching a peak to the east in the afternoon and likely reflects terrestrial sources. This pattern in $C(t,\theta)$ extends to nearly 130°. Beyond the seep field's western edge, u is elevated from the prevailing direction (270°), with C elevated throughout the morning. There also is a short-lived peak in u around noon at ~300°, which corresponds to a short-lived depressed C. These could be consistent with wave development time, transport time, and sparging of the downcurrent plume; however, interpretation based on these patterns largely is speculative.

466

467 **3.3 Overall seep field emissions**

468 **3.3.1 Overall emissions**

Average atmospheric emissions, E_A , for 1990-2020 were derived by an iterative Gaussian plume model, initialized with the 2005 sonar map (**Fig. 7A**). An emissions sensitivity study on the effect of grid resolution was conducted for resolutions from 11 to 225 m and a 22/56-m hybrid grid (**Fig. S3**). Simulations used moderate insolation to derive the turbulence parameters and stability class, a 250-m *BL*, a typical Santa Barbara Channel marine values (Edinger, 1959; Rahn, Parish, & Leon, 2017), and 2° angular resolution (Hanna et al., 1982). Simulations were run iteratively until convergence, typically within 5 iterations (**Supp. Fig. S11**). Sensitivity studies found the distance weighting function,

475 $K(r,\theta)$, was linear (Supp. Fig. S12).

476

477 Simulations could not reproduce observations in the Platform Holly direction (θ =238°). Thus, a source was added for

the platform area, which improved simulation-observational agreement in this wind direction. Since significant seep

bubbles plumes generally are not observed in the platform's vicinity, these emissions could arise from incomplete

- 480 combustion from flaring.
- 481

The model-derived, E_A , for 1990-2020 was 83,500 m³ dy⁻¹ (Fig. 7). Using a composition-weighted molecular mass of

483 19.6 g mole⁻¹ implies 27 Gg THC yr⁻¹. Atmospheric seep gas is 88.5% CH₄, implying 19 Gg CH₄ yr⁻¹ seep emissions

484 (Table 1). Given that CH₄ is 73% of THC, non-methane hydrocarbon (NMHC: C₂-C₇) emissions are 9,500 m³ dy⁻¹

and Reactive Organic Carbon (ROC) gases emissions of 6.0 Gg yr⁻¹ - ROC are organic species excluding CH4
including alkanes and aromatic compounds. For reference, Santa Barbara County 2018 ROC emissions were ~27 tons
dy⁻¹ (9.9 Gg yr⁻¹) (ourair.org/emissions-inventory, SBAPCD). The largest NMHC was propane with emissions of 3510
m³ dy⁻¹, followed by ethane at 2590 m³ dy⁻¹. The NMHC components of THC are conservative (do not react

- significantly) on the typical the transport time from the seep field to WCS (20-30 minutes).
- 490

Seabed emissions, E_B , are necessarily significantly greater than E_A as E_A misses the fraction of emissions that remain in the water column, E_W , at least in the field's near downcurrent. There are two notes, the model E_A includes evasion from the dissolved plume in the area covered by the seep field sonar map. Secondly, the model does not include E_A from the dissolved fraction that evades beyond the seep field extent. For the seep field area and near downcurrent area, Clark et al. (2000) estimated a 50:50 air/water partitioning, implying seabed emissions, E_B , 1990-2020 of 167,000 m³ dy⁻¹ or 54 Gg yr⁻¹. A comparison of E_A versus ω showed a very steep increase with ω for $E_A = 1-10$ g s⁻¹ m⁻² with

- 497 rollover at *ω*~0.022 (**Supp. Fig. S13**).
- 498

499 Insights were provided by how the model partitioned emissions between different seep areas. Particularly notable is 500 the model's treatment of the Trilogy Seep area - the second strongest seep area after the Seep Tent Seep over 1990-501 2020. The model re-assigned Trilogy Seep emissions to seepage to the west, representing Trilogy Seep emissions as 502 unrealistically weaker than other, smaller seeps, such as IV Super Seep. One likely contributor to this re-assignment 503 is wind veering. Also suggesting wind veering is the model's assignment of strong emissions to the field's eastern and 504 western edges despite weak sonar returns. In a comparison of the Seep Tent Seep and La Goleta Seep areas, the model 505 emphasized the Seep Tent Seep whereas La Goleta Seep emissions were shifted to inshore seepage. This re-506 partitioning was greatly reduced for a $\pm 10^{\circ}$ wind veer, which also lessened the strengthening of emissions from the 507 field's western edge relative to sonar (Supp. Fig. S14). Given the lack of field data between the seep field and WCS 508 on wind veering, further wind veering analysis was not conducted.

509

510 **3.3.2 Seep field sector emissions**

511 To investigate sub-field scale emissions, the seep field was segregated into three sectors: inshore, offshore east, and 512 offshore west (Fig. 1). Based on integrating sonar return, ω , the inshore seepage contributes 40% of the field's ω with 513 the offshore seep trend split between 9% for the west and 51% for the east. Supporting this comparison is the similarity 514 in the normalized sonar return probability distribution, $\phi_n(\omega)$, for the inshore seeps and offshore east seeps (Fig. 8). In 515 contrast, $\phi_n(\omega)$ for the offshore west seepage differed dramatically despite the similarity in geology along the anticline 516 underlying the offshore seep trend (Leifer et al., 2010). This likely results in part from the interaction between 517 migration and production from Platform Holly. Although the normalized atmospheric emissions probability 518 distribution, $\phi_n(E_A)$, for the inshore and offshore seeps are similar over most of the range (except the weakest, $E_A < 0.02$ 519 $g s^{-1}$), significant differences are evident between offshore east and west seepage. Offshore east seepage is more 520 dispersed and favors weaker seepage compared to offshore west seepage and compared to $\phi_n(\omega)$.

The weakest seepage ($\omega < 0.02$) contributes negligibly to overall sonar return and had no notable inshore-offshore $\phi_n(\omega)$ difference. The largest difference is between the strongest seepage ($\omega > 0.5$) for the inshore and offshore seeps. Specifically, there is a strong peak at $\omega \sim 0.45$ and nothing stronger for the inshore seeps, whereas offshore $\phi_n(\omega)$ continued to $\omega \sim 0.7$. The E_A probability distribution, $\phi_n(E_A)$, for the strongest inshore seepage was similar to $\phi_n(E_A)$ for strong offshore seepage. However, this masked a significant east-west offshore seepage difference. Specifically, $\phi_n(E_A)$ for strong seepage was reduced far more for offshore east seepage than offshore west seepage, and the reverse for weak seepage.

529

These distributions suggest that controlling geological structures (fractures, fault damage zones, and chimneys in the capping Sisquoc Formation) are the same for inshore seepage and offshore east seepage, with the primary difference for the strongest seepage in these two sectors which are of similar strength – the inshore Trilogy Seeps provide focused emissions, whereas the offshore east La Goleta Seeps are comparatively dispersed and far oilier.

534

535 Although, ω is not E_4 , E_4 followed the 40:60 partition in ω between inshore and offshore seepage. Interestingly, the 536 E_A partitioning between the offshore east and offshore west differed significantly from sonar partitioning with 21% of 537 E_A from offshore west and 38% from offshore east. This greatly accentuated the E_A Seep Tent Seep area. In part, this 538 arises from a diurnal cycle bias - WCS observes the offshore west seeps for afternoon/evening westerly winds, which 539 are stronger, whereas WCS observes the offshore east seeps when winds are weaker, earlier in the day (Fig. 6B). 540 Winds increase bubble emissions from wave hydrostatic pumping and dissolved gas evasion. Also potentially 541 contributing is saturation of ω at very high bubble-density bubble plumes, such as the Seep Tent Seep and Trilogy 542 Seep (Leifer et al., 2017). Saturation would imply an under-estimate of ω for the strongest seep area emissions which 543 are for the west offshore seepage, altering the west:east ω ratio (9%:51%).

544

545 **3.3.3 Uncertainty and emissions sensitivity**

Given the number of sources with poorly characterized variability, uncertainty is best assessed by Monte Carlo simulations; however, this was unfeasible due to the simulations' computational demands. Thus, emissions uncertainty was investigated by sensitivity studies (**Fig. 9**). Where data were available, uncertainty due to a specific parameter was estimated from the data. Specific parameters studied included sonar resolution, angular resolution, $\delta\theta$, wind speed, *u*, concentration anomaly, *C*', boundary layer height, *BL*, wind veering, ψ , spatial northing offset, *Y*, and the inshore and offshore seepage partitioning, ζ . Sensitivity study details are presented in **Supp. Sec. S7.4**.

552

553 The contribution to uncertainty from $\delta\theta$, C', ψ , and spatial offsets within the seep trends were minimal – just a few

percent or less. Moderate uncertainty was identified for BL and ζ . For example, for BL ranging from 150 to 350 m,

555 mean E_A uncertainty was 6%. Although u has strong sensitivity, combined with BL it does not as u counters BL –

- lower u corresponds to higher BL. There still is uncertainty, though in the value of BL, which is not measured.
- 557 Assessing uncertainty in ζ was more challenging as there is no verification data on variability in the E_A partitioning
- between the inshore and offshore seep trends. The mean E_A uncertainty for -50% < ζ < 50% is 11.5% from a polynomial

559 fit. Still, the consistency in seepage location between sonar surveys spanning decades (Leifer, 2019) suggests only

- 560 modest changes in ζ over the multi-decade time period of model averaging. Total uncertainty was taken as 15% based
- 561 on the sum of uncertainty in *BL* and ζ , each averaged to the nearest 5%.
- 562

563 3.4 Ellwood Field emissions

564 $C(\theta)$ increases to the northeast with a peak at 290-320° corresponding to the direction towards abandoned wells off 565 Haskell Beach (**Fig. 10**). Emissions from this area – either from natural seepage or leaking wells – were noted in the

offshore survey data near Haskell Beach (Fig. 2A). Additionally, $C_{max}(\theta)$ shows a 22-ppm peak in in this direction

- well above $C_{ave}(\theta)$ (Fig. 4F), consistent with transient releases from natural seep and/or abandoned well emissions.
- Ellwood field production continued through the 1970s with wells drilled into the geological structures that allowed oil accumulation (Olson, 1983); including faults that provide migration pathways (Leifer et al., 2010). There are many abandoned wells from these oil fields and from others fields on the Goleta Plains, beaches, and shallow near-coastal waters to the west-northwest of WCS (offshore Haskell Beach and onshore around Naples Point). Currently, active wells only are found at the La Goleta Gas field (a natural gas storage field), east of WCS.
- 574

Faults associated with these anticlines provide migration pathways and are approximately aligned with the coast in a series of roughly parallel faults extending onshore (Minor et al., 2009). The onshore/coastal Ellwood field (northwest of the South Ellwood field) sources from the primarily sandstone Vaqueros Formation (Olson, 1983), whose main trap is an anticline at the western edge of the North Branch Western More Ranch Fault (NBWMRF). Offshore seepage tracks some of these faults, e.g., the Isla Vista Fault trend corresponds to an offshore seep trend in Goleta Bay that includes the Goleta Pier Seep, whereas wells follow the NBWMRF trend offshore of Haskell Beach.

581

582 4 Discussion

583 4.1 Atmospheric seep field observations

584 4.1.1. Air quality station

585 A range of approaches are available to evaluate marine seepage CH4 emissions: in situ approaches including direct

586 capture (Leifer, 2015; Washburn, Johnson, Gotschalk, & Egland, 2001), fluid flow measurements (Leifer & Boles,

587 2005), video (Leifer, 2015), and remote sensing approaches that include active acoustics, i.e., sonar (Hornafius et al.,

1999), dissolved in situ (Marinaro et al., 2006), and passive acoustics (Wiggins et al., 2015). Remote sensing is the

589 best approach for long-term monitoring to capture shifts in emissions between vents. To date, only sonar remote 590 sensing has provided quantitative seep plume (seabed) emissions. Notably, sonar ranges are up to a few hundred 591 meters, far less than the size scales of many seep fields, while high power-demands typically require a cabled 592 observatory for long-term observations.

593

This study demonstrated that air quality station data can provide the long-term continuous data needed to capture seasonal variations including emissions during storms and transient events, which field campaigns likely miss. For example, sonar surveys tend to occur during summer when seas are calmer and more predictable and when seepage is weakest (**Fig. 3**); however, not during storms when emissions likely are enhanced.

598

599 The approach derived atmospheric trace gas emissions for a dispersed area source constrained by sonar seepage maps 600 from long-term air quality and meteorology data. This approach can be extended to terrestrial seepage if the source 601 can be constrained spatially (due to geology); although nearby anthropogenic sources may complicate emissions 602 assessments. Other terrestrial sources such as landfills, O&G production fields, or industrial sites - if spatially 603 constrained – could be addressed by this approach, particularly if isolated from other confounding sources. The use 604 of cavity enhanced absorption spectrometers that can speciate gases like CH₄ and C₂H₆ could enable discrimination 605 some confounding sources as well as better characterization of emissions. Although an onshore station can address 606 nearshore seepage, further offshore seepage could be addressed by a moored station. Moored stations also could 607 include in situ aqueous chemical sensors, current measurements.

608

609 4.1.2 *In situ* atmospheric surface surveys

Atmospheric emissions were assessed for three seep areas - zone of focused seepage - by an atmospheric *in situ* survey approach wherein downwind data are collected orthogonal to the wind direction in a transect that spans the plume (background to background on the plume's edges). This approach was developed for terrestrial sources (Leifer, Melton, Tratt, et al., 2018) yet remains unused for offshore marine seepage, which often are area sources. In this study, this was addressed by gridding the area source and treating each grid as a far-field point source. Gaussian plume inversion requires distant source(s), i.e., far field. Surveys of three strong seep areas all were well characterized by the Gaussian plume model.

617

One advantage of atmospheric surveys is rapidity - a single transect of a few minutes is sufficient to derive emissions for a seep area. In comparison, a flux buoy survey can require many hours to a day (Clark et al., 2010), during which forcing factors (waves, tides, etc.) change significantly. Also rapid are seep area sonar surveys (Wilson, Leifer, & Maillard, 2015) allowing a combined sonar and atmospheric survey to repeat characterize emissions and sea-air partitioning within a few hours. With respect to the entire COP seep field, whereas a sonar survey requires two to three days (Leifer et al., 2010), a downwind atmospheric survey is far more rapid, requiring perhaps an hour. This allows repeat field emissions measurements over a tidal cycle.

626 4.2 Seep field emissions

627 4.2.1 Total emissions

628 To date, only two estimates of COP seep field seabed emissions, E_{B} , have been published. Hornafius et al. (1999) estimated $E_B=1.5 \times 10^5$ m³ dy⁻¹ (64 Gg yr⁻¹) based on sonar surveys covering 18 km² from Nov. 1994 – Sep. 1996, 629 collected during the summer to late fall seasons. This value excluded Seep Tent collection. A 4.1 km² sonar survey in 630 631 Aug.-Sep. 2016 estimated $E_B=24,000 \text{ m}^3 \text{ dy}^{-1}$ (Padilla, Loranger, Kinnaman, Valentine, & Weber, 2019), significantly 632 lower, which in part arises from field subsampling, but also could arise from long-term changes; however, neither 633 study addressed temporal variability. The sonar surveys occurred in summer and fall when seepage activity is at a 634 minimum, whereas winter and early spring feature much higher activity associated with large transient events and 635 storms (Bradley et al., 2010).

636

Hornafius et al. (1999) used an engineered bubble plume to calibrate emissions, an approach also used in Leifer et al. 637 (2017). Due to technology limitations at the time, the strongest seepage was clipped or saturated, i.e., underestimated, 638 and the survey did not cover shallow seepage. Thus, the Hornafius et al. (1999) emissions estimate is a lower limit for 639 640 summer/fall emissions. The Padilla et al. (2019) survey was calibrated by an inverted seep flux buoy suspended at 23 641 m. This differs significantly from seep flux buoy measurements (Washburn et al., 2001), which are collected in surface 642 drift mode. Surface drift mode ensures a horizontal orientation for the buoy and an absence of lateral velocity 643 difference between the capture device and currents – either of which decreases capture efficiency from 100%, biasing 644 derived emissions low. Further, the Padilla et al. (2019) survey was calibrated 1 month after the sonar surveys, whereas 645 the 1995 engineered plume calibration by Hornafius et al. (1999) was contemporaneous. The Hornafius et al. (1999) 646 approach accounts (partially) for dissolution between the seabed and survey depth window, albeit air dissolves slower 647 than methane. Dissolution losses between the seabed and the depth window can be addressed by a numerical bubble 648 model (Leifer et al., 2017).

649

The Gaussian plume model-derived E_A was 8.4×10^4 m³ dy⁻¹. Based on the Clark et al. (2000) assessment that half the seabed seepage reaches the atmosphere, $E_B=1.7 \times 10^5$ m³ dy⁻¹; very similar to $E_B=1.5 \times 10^5$ m³ dy⁻¹ from Hornafius et al. (1999). This agreement is coincidental as it neglects seasonal and interannual trends. For example, Bradley et al. (2010) found 1994-1996 emissions were well below the average for 1990-2008, increasing significantly after 2008.

654

655 4.2.2 Methane and non-methane hydrocarbon emissions

Analysis of atmospheric samples provided a picture of the complexity of atmospheric emissions that arises from the

multiple pathways underlying atmospheric emissions. Specifically, as bubbles rise, they lose lighter and more soluble

- gases faster (deeper in the water column), leading to differences between evasion from dissolved gases and direct
- bubble transport (Leifer & Clark, 2002). Thus, bubble-mediated transport enhances larger alkanes relative to smaller

alkanes leaving more of the smaller alkanes in the water column. For strong seeps, bubble plumes are associated with

- strong upwelling flows (Leifer et al., 2009), which transport dissolved gases to the sea surface where they outgas.
- 662 Additionally, oil (as droplets and bubble coatings) enhances alkane transport due to slower dissolution and diffusion
- 663 of larger alkanes through oil.
- 664

Atmospheric plume concentrations were 11.5% NMHC and 88.5% CH₄, very similar to Hornafius et al. (1999) who referenced the Seep Tent composition (88% CH₄, 10% NMHC, and 2% nitrogen) as very similar to the reservoir composition. Note, Clark et al. (2010) observed Trilogy near sea surface bubbles with 5.7% to 7.9% NMHC and 52.4 to 79.7% CH₄, demonstrating significant partitioning. The similarity between the atmospheric and seabed composition despite the difference in the bubble composition demonstrates efficient dissolved gas transfer to the sea surface.

670

671 COP seep field seabed emissions are orders of magnitude greater than typically reported for other seep areas, e.g., 672 summary Römer et al. (2017) where emissions for 12 different seep areas including sites in the North Sea, Pacific 673 northwest, Gulf of Mexico, etc., were 2-480 tons yr⁻¹, multiple orders of magnitude less than COP seep field seabed 674 emissions. Römer et al. (2017) used a bubble model for Dogger Bank seepage in the North Sea to estimate emissions 675 for observed atmospheric CH₄ plumes. The model estimated direct atmospheric bubble-mediated emissions of 21.7 ton yr⁻¹, 20% of seabed emissions. For the Tommelieten Seeps (in 70-m water) Schneider von Deimling et al. (2011) 676 677 estimated 4% of the 0.024 Gg CH₄ yr⁻¹ seabed emissions, i.e., ~1 Mg CH₄ yr⁻¹ reached the atmosphere by bubble-678 mediated transfer. Schneider von Deimling et al. (2011) used a bubble model based on an assumed bubble size and 679 neglected diffusive flux. These diffusive fluxes include bubble dissolution into the wave mixed layer in the local area. 680 A few studies have directly measured atmospheric fluxes by an air-sea gas transfer model. For example, Schmale, 681 Beaubien, Rehder, Greinert, and Lonmbardi (2010) found seep air fluxes of 0.96-2.32 nmol m⁻² s⁻¹, much higher than 682 the ambient Black Sea flux of 0.32-0.77 nmol m⁻² s⁻¹. In the Black Sea, ambient emissions arise from microbially produced CH₄ in shelf and slope sediments (Reeburgh et al., 1991). Di, Feng, and Chen (2019) estimated 7.7 nmol m⁻ 683 ² s⁻¹ for the shallow South China Sea based on an air-sea gas transfer model. If we disperse COP seep field atmospheric 684 emissions of 1.15×10^9 M yr⁻¹ over the ~6.3 km² of 25×25 m² bins with emissions, we find 5.7 μ M m⁻² s⁻¹, three orders 685 686 of magnitude greater.

687

Recent estimates of total global geo-CH₄ sources from a bottom-up approach are 45 Tg yr⁻¹ with submarine seepage 688 contributing 7 Tg yr⁻¹ (Etiope & Schwietzke, 2019), implying COP seep field contributes 0.25% of the bottom-up 689 690 submarine emissions. However, an estimate of pre-industrial CH4 emissions (not confounded with fossil fuel production emissions) based on ice core ¹⁴CH₄ suggested 1.6 Tg geo-CH₄ yr⁻¹ emissions (Hmiel et al., 2020). This 691 692 estimate, if accurate, would imply the COP seep field contributes an astounding 1% of global seep emissions 693 (submarine and aerial) and is difficult to reconcile with the COP seep field and other top seepage estimates. For 694 example, CH₄ atmospheric emissions for the Lusi hydrothermal system of 0.1 Tg yr⁻¹ (Mazzini et al., 2021), a hotspot 695 in the Laptev Sea of 0.9 Tg yr⁻¹ into shallow seas (Shakhova, Semiletov, Leifer, et al., 2010), and for the East Siberian

- Arctic Sea using eddy covariance of 3.0 Tg yr^{-1} (Thornton et al., 2020). Thus, COP seep field emissions either play a significant role in global seep emissions or indicate that geo-gas emissions are less tightly constrained.
- 698

699 COP seep field C₂H₆ emissions were 1.27 Gg C₂H₆ yr⁻¹. For reference, this is 11% of the 11.4 Gg C₂H₆ yr⁻¹ in 2010

- for the South Coast Air Basin (SCAB), which includes Los Angeles (Peischl et al., 2013). Globally, Simpson et al.
- (2012) and Höglund-Isaksson (2017) found 11.3 and 9.7 Tg C_2H_6 yr⁻¹ in 2010, respectively. C_2H_6 has been increasing since 2010 due to increased O&G production emissions (Helmig et al., 2016). Globally, seeps are estimated to
- rontribute 2-4 Tg C₂H₆ yr⁻¹ (Etiope & Ciccioli, 2009), and from ice cores, 2.2-3.5 Tg yr⁻¹(Nicewonger, Verhulst,
- Avdin, & Saltzman, 2016). This suggests the seep field contributes 0.03-0.06% of global seep emissions.
- 705
- Seep THC was 4.2% propane, implying emissions of 2.5 Gg C₃H₈ yr⁻¹. Global propane emissions are 10.5 Tg yr⁻¹
- 707 (Pozzer et al., 2010), with 1-2 Tg yr⁻¹ estimated for seeps (Etiope & Ciccioli, 2009). This suggests the COP seep field
- contributes 0.05-0.1% of the global seep budget. Oceans are estimated to contribute 0.35 Tg C₃H₈ yr⁻¹ (Pozzer et al.,
- 709 2010), less than geological seepage contribution.
- 710

711 Based on an evaluation of the COP seep field emissions with respect to global seep ethane and propane emissions,

- 712 COP seep field contribution to global geo-CH₄ emissions are consistent with recent global geo-gas CH₄ emissions
- estimates of 45 Tg yr⁻¹ (0.04%) (Etiope et al., 2019), not the significantly lower pre-industrial estimates of global geo-
- 714 CH₄ emissions, e.g., 1.6 Tg yr^{-1} (1.15%) (Hmiel et al., 2020).
- 715
- Global butane emissions are 14 Tg C₄H₁₀ yr⁻¹ (Pozzer et al., 2010), higher than ethane and propane. COP seep field butane (C₄) and pentane (C₅) emissions were 2.2 Gg C₄H₁₀ yr⁻¹ and 1.1 Gg C₅H₁₂ yr⁻¹, respectively, thus combined C₂-C₅ emissions are 7.1 Gg yr⁻¹, compared to 65 Gg yr⁻¹ from the entire SCAB, i.e., COP seep field contributes ~5% the SCAB. COP C₂-C₅ emissions are significantly above that of the La Brea area, estimated at 1.7 Gg yr⁻¹ (Weber et al., 2017). Note, COP seep field atmospheric C₂-C₅ emissions certainly are larger, potentially significantly, as larger alkanes also are emitted from oil slicks but were not considered for this study, and furthermore, the atmospheric plume from the slicks was not sampled for this study.
- 723
- Both benzene and toluene were detected with estimated emissions of 5000 and 1300 kg yr⁻¹, respectively. These emissions likely are underestimates, potentially significantly, due to neglecting the oil slick evaporation contribution.
- Both gases are of significant health concerns, as are alkanes like pentane and hexane.
- 727

728 4.3 Downcurrent emissions

The seep field concentration, $C'(\theta)$, anomaly was centered at θ -200° and well matched the location of the seep field,

- and moreover, was well described by a dual Gaussian function (Fig. 4B). This was surprising given that the seep field
- is asymmetric with respect to a 200° axial line from WCS to COP. Underlying this seeming discrepancy is that WCS

winds are weakest from due south and strongest from the west (prevailing) and also stronger to the east-southeast(Fig. 4C).

734

The residual of the Gaussian fit increased in the downcurrent direction (Supp. Fig. S9B), consistent with evasion from

- the downcurrent dissolved plume and seepage from this area. The dissolved plume roughly follows the coast, extending as far as $\sim 280^{\circ}$ from WCS due to the coastline shift from northwest to west around Haskell Beach (**Fig. 2**),
- $\sim 30^{\circ}$ beyond the seep field's sonar mapped western edge (**Fig. 1**). As prevailing winds are westerlies (paralleling the
- coastal mountains), downcurrent plume evasions decrease with distance due to dispersion and also as surface waters
- 740 become depleted by evasion. Evasion increases non-linearly with *u*, particularly for winds that include wave breaking
- 741 (Nightingale et al., 2000); however, higher winds also dilute emissions. Note, there are no mapped seeps in this area.
- 742

Dissolved plume emissions also likely occur from east of the field, leading the model to emphasize seepage at the field's eastern extent, too. Specifically, strong prevailing afternoon westerly surface winds drive a near-surface dissolved plume eastwards. When these westerly winds calm down late in the evening, easterly winds transport evasion from this east-displaced dissolved plume towards WCS. Additionally, it also is possible that the COP seep field extends further east than mapped in sonar surveys, at least during some seasons.

748

749 **4.4 Focused seep area emissions**

Trilogy Seep area emissions were estimated at 6,200 m³ CH₄ dy⁻¹ in May 2016. For comparison, Clark et al. (2010) 750 751 found 5500 and 4200 m³ THC dy⁻¹ (4,900 and 3,700 m³ CH₄ dy⁻¹) for Trilogy Seep as measured by flux buoy for near 752 surface bubble fluxes in Sept. 2005. Note, the plume inversion approach also includes outgassing of near surface 753 waters that have enhanced C_{CH4} from plume dissolution, which the flux buoy approach does not include. Although 754 Clark et al. (2010) found surface bubbles had undetectable CO₂, the atmospheric plume's CO₂ to CH₄ concentration 755 ratio was comparable to the seabed bubble concentration ratio. This demonstrates significant upwelling flow transport 756 of seabed water to the sea surface where dissolved gases evade near where the bubble plume surfaces. This near-plume 757 evasion contributes to the atmospheric plume. Note, these emissions neglect downcurrent emissions. A 50:50 758 atmosphere:ocean partitioning suggests 2016 Trilogy Seep emissions were ~40% lower than in 2005 - a difference 759 within the difference between the two 2005 Trilogy Seep measurements Clark et al. (2010).

760

In contrast, agreement was very poor for the Seep Tent Seep, for which Clark et al. (2010) mapped emissions of 5700 m³ day⁻¹ (5000 m³ CH₄ day⁻¹) in Nov. 2002 whereas this study found 310 m³ CH₄ dy⁻¹. This discrepancy was readily apparent with almost no visible surface bubble expression in May 2016, whereas the Seep Tent Seep has been a perennial feature since its appearance. The absence of more than a few scattered bubbles at the sea surface (the boil in 2000 was driven by a 1-2 m s⁻¹ upwelling - Leifer, Clark, and Chen (2000) - indicates that most emissions are from evasion. A buoyancy plume associated with the rising oil (thick oil slicks surface above the Seep Tents) as well as methane dissolved in the oil likely are transporting the observed, focused CH₄ emissions.

- 769 This is remarkable given that the seep field's geofluid migration "center" in recent decades has been the Seep Tent
- 770 Seep (Bradley et al., 2010), which was the largest seep in the field in 2010 (Clark et al., 2010). The Seep Tent Seep
- consists of emissions not captured by the Seep Tents two large (33-m square) steel capture tents on the seafloor. For
- reference, the Seep Tents captured $\sim 16,800 \text{ m}^3$ gas dy⁻¹ in the early 2000s (Boles et al., 2001). Bradley et al. (2010)
- found in WCS data that when overall seep field emissions decreased to a minimum in 1995, they were focused on the
- Seep Tent Seep direction. Note, the Seep Tent Seep was observed first in 1970 as a boil visible from 1.6-km distant.
- The seepage was tented in Sept. 1982 (Boles et al., 2001).
- 776

Underlying these observations are several factors. First, the Seep Tent Seep is modern – since 1978 – as it was not mapped in a 1953 seep survey (Leifer, 2019). At the time it was first reported as a sea boil visible over a kilometer distant (Boles et al., 2001). Since installation, overall Seep Tent production has diminished (Boles et al., 2001) by a factor of 3 from 1984 to 1995. Some fraction of this trend could have resulted from the expansion of active seepage beyond the seep tents. Perhaps more significantly, the Seep Tent Seep lies over one of the Platform Holly wells (Leifer et al., 2010; Fig. 3C), creating the potential of linkage between well production (including stimulation) and Seep Tent

- 783 production and thus Seep Tent seepage (the uncaptured portion).
- 784

785 **4.6 Diurnal trend and bias**

786 The diurnal wind patterns typical of the coastal Pacific marine environment are weak offshore (northerly) night winds 787 that shift to from the east in the morning and then swing to from the south. In afternoon they strengthen and shift to prevailing westerlies, continuing to late in the evening (Bradley et al., 2010). Note, WCS seep emissions require winds 788 789 to "probe or scan" across the seep field, and thus miss the strong afternoon prevailing winds when emissions are 790 expected to be higher. This is because higher wind speeds increase sea-air gas emissions of dissolved near-surface 791 gases (Nightingale et al., 2000) and increase emissions from higher hydrostatic pressure fluctuation driven by wave 792 height (Leifer & Boles, 2005). Given that prevailing winds are westerlies, higher afternoon emissions will generally 793 (but not always) drift eastwards, missing WCS.

794

The diurnal wind pattern from the seep field direction is different from the overall (direction-independent) diurnal pattern. Typical nocturnal winds are quite weak, $1.5-1.7 \text{ m s}^{-1}$ (**Fig. 6**). The strongest diurnal wind change was from late night to morning, a 20% decrease. Onshore winds (seep direction) in the middle of the night are from synoptic systems and were associated with the highest *C*'. Winds increase by a few percent to an early afternoon peak, decreasing through early evening before increasing again later in the night.

800

The diurnal trend for *C* from the seep direction followed the diurnal wind cycle, increasing by ~ 20 ppb and peaking ~ 2 hours later in the day than winds (15:00 versus 13:00 for *C* compared to *u*, respectively). This may reflect the lag

- in wave development with respect to wind strengthening and transport time. Based on sensitivity studies, the diurnal cycles in *u* and *C* correspond to variations of \sim 7% and \sim 9% in *E*_A.
- 805

Although efforts were made to characterize the diurnal cycle from WCS data, WCS data poorly sample the seep field for the higher wind speeds that occur in the afternoon which primarily are westerlies. Note, non-linearity in sea-air evasion with u means the model use of average u underestimates E_A . Thus, the contribution of the prevailing afternoon winds to diurnal emissions is significantly underestimated from WCS data. It is worth noting, though, that this factor only affects 25-33% of diurnal emissions. As the true diurnal cycle cannot be derived from WCS data, field data of repeat transects spanning the different phases of a diurnal cycle are needed.

812

813 4.7 Future needs and improvements

The sensitivity studies identified areas for improvement and data gaps. These are described in brief below and in more detail in **Supp. Sec. S8**. The largest uncertainty was with regards to partitioning between the inshore and offshore seep trends, which could be determined by a second air quality station, preferably including speciation such as by CEAS analyzers of CH₄ and C₂H₆. Further simulations could add grid cells for evasion corresponding to the downcurrent plumes to assess their contribution. The model was limited by available workstation power; however, additional computation power could open improvements such as simulating a range of wind speed based on the wind speed probability distribution with respect to wind direction, $\phi(u, \theta)$.

821

Additional field work and data also are needed. Another important sensitivity was to boundary layer height, *BL*, which varies diurnally and seasonally (Dorman & Winant, 2000) and could be derived from ceilometer data (Münkel, 2007). Another significant concern is afternoon seep field emissions that bypass WCS, which could be addressed by field work and a second air quality station at a different downwind direction from the seep field. Mapping offshore wind fields to characterize wind veering across the seep field is needed to allow simulations to provide insights at the seep area size-scale.

828

829 5 Conclusions

In this study, data from an onshore air quality station located downwind of a large marine seep field was analysed to derive the three-decade-averaged seep field emissions using an inversion model. The modeled emissions were similar to reported emissions; however, this was coincidental given that prior reported emissions were during a period of field quiescence. Highlighting the significance of the COP seep field, ethane and propane emissions suggest the COP seep field contributes 0.04% and 0.12% of the global seep budget, respectively. As a result, COP seep field emissions of 19 Tg CH₄ yr⁻¹ are consistent with global geo-gas budgets of 45 Tg yr⁻¹, but inconsistent with significantly lower

emissions estimated from ice core isotopic data. Additionally, the approach could be adapted to air quality station data

for other sources including terrestrial seeps, production fields, etc., if the sources are spatially constrained and isolated
 from confounding sources.

839

- **Data availability.** All data needed to evaluate the conclusions in the paper are present in the paper and/or the Supplementary Materials and/or were submitted to the Mendeley Data Repository, see Leifer, Ira (2020),
- 842 "Seep_Air_Data", Mendeley Data, V1, http://dx.doi.org/10.17632/znzhxkftm8.1
- 843
- 844 **Supplement.** The supplement contains additional supporting figures and details to complement the manuscript and
- an interactive map file as a Google Earth archive of the offshore survey data that are presented in Fig. 2.
- 846
- Author Contributions. IL Developed and conducted the study, analysed data, and wrote the manuscript. CM analysed
 data and edited the manuscript. DB analysed air sample data and edited the manuscript.
- 849
- 850 **Competing interests.** The authors declare that they have no conflict of interest.
- 851
- 852 Acknowledgements. We would like to gratefully acknowledge the SBCAPCD for providing data from their ongoing
- 853 monitoring program and the contribution of Marc Moritsch and Joel S. Cordes in particular for help with these data,
- and Doug Wilson for the processed sonar data. We recognize the skill and participation of vessel captains Jeff Wright
- and Tony Vultaggio and editorial review by Charlotte Marston, Bubbleology Research International.
- 856
- Financial Support. This work was supported by Plains All American Pipeline and the Bubbleology Research
- 858 International, Internal Research and Development (IRAD) fund.

860 **References**

- Abrams, M. A. (2005). Significance of hydrocarbon seepage relative to petroleum generation and entrapment. *Marine and Petroleum Geology*, 22(4), 457-477.
- 863 doi:10.1016/j.marpetgeo.2004.08.003
- Abrams, M. A. (2017). Evaluation of near-surface gases in marine sediments to assess
 subsurface petroleum gas generation and entrapment. *Geosciences*, 7(2), 35.
 doi:10.3390/geosciences7020035
- Bernard, B. B., Brooks, J. M., & Zumberge, J. (2001, 16-19 September 2001). *Determining the origin of gases in near-surface sediments*. Paper presented at the AAPG Hedberg
 Conference, Vancouver BC, Canada.
- Boles, J. R., Clark, J. F., Leifer, I., & Washburn, L. (2001). Temporal variation in natural
 methane seep rate due to tides, Coal Oil Point area, California. *Journal Geophysical Research Oceans, 106*(C11), 27,077-027,086. doi:10.1029/2000JC000774
- Borges, A. V., Champenois, W., Gypens, N., Delille, B., & Harlay, J. (2016). Massive marine
 methane emissions from near-shore shallow coastal areas. *Scientific Reports*, *6*, 27908.
 doi:10.1038/srep27908
- Bradley, E. S., Leifer, I., & Roberts, D. A. (2010). Long-term monitoring of a marine geologic
 hydrocarbon source by a coastal air pollution station in Southern California. *Atmospheric Environment, 44*(38), 4973-4981. doi:10.1016/j.atmosenv.2010.08.010
- CDOGGR. (2018). Well Finder. Retrieved from
 <u>https://www.conservation.ca.gov/dog/Pages/Wellfinder.aspx</u>. Retrieved 6 May 2019,
 from California Department of Conservation
 https://www.conservation.ca.gov/dog/Pages/Wellfinder.aspx
- 882 <u>https://www.conservation.ca.gov/dog/Pages/Wellfinder.aspx</u>
- Clark, J. F., Washburn, L., Hornafius, J. S., & Luyendyk, B. P. (2000). Natural marine
 hydrocarbon seep source of dissolved methane to California coastal waters. *Journal Geophysical Research Oceans, 105*, 11,509-511,522. doi:10.1029/2000JC000259
- Clark, J. F., Washburn, L., & Schwager, K. (2010). Variability of gas composition and flux
 intensity in natural marine hydrocarbon seeps. *Geo-Marine Letters*, *30*, 379-388.
 doi:10.1007/s00367-009-0167-1
- Di, P., Feng, D., & Chen, D. (2019). The distribution of dissolved methane and its air-sea flux in
 the plume of a seep field, Lingtou Promontory, South China Sea. *Geofluids*, 2019,
 3240697. doi:10.1155/2019/3240697
- Bi, P., Feng, D., Tao, J., & Chen, D. (2020). Using time-series videos to quantify methane
 bubbles flux from natural cold seeps in the South China Sea. *Minerals*, 10(3), 216.
 doi:10.3390/min10030216
- Borman, C. E., & Winant, C. D. (2000). The structure and variability of the marine atmosphere
 around the Santa Barbara Channel. *Monthly Weather Review*, *128*(2), 261-282.
 doi:10.1175/1520-0493(2000)128<0261
- Edinger, J. G. (1959). Changes in the depth of the marine layer over the Los Angeles Basin. *Journal of Meteorology*, 16(3), 219-226. doi:10.1175/15200469(1959)016<0219:citdot>2.0.co;2
- Etiope, G., & Ciccioli, P. (2009). Earth's degassing: A missing ethane and propane source.
 Science, 323(5913), 478-478. doi:10.1126/science.1165904
- Etiope, G., Ciotoli, G., Schwietzke, S., & Schoell, M. (2019). Gridded maps of geological
 methane emissions and their isotopic signature. *Earth System Science Data*, 11(1), 1-22.
 doi:10.5194/essd-11-1-2019

- Etiope, G., & Schwietzke, S. (2019). Global geological methane emissions: An update of topdown and bottom-up estimates. *Elementa: Science of the Anthropocene*, 7.
 doi:10.1525/elementa.383
- Fischer, P. J. (1978). Oil and Tar Seeps, Santa Barbara Basin, California. In D. J. Everitts, R. G.
 Paul, C. F. Eaton, & E. E. Welday (Eds.), *California Offshore Gas, Oil and Tar Seeps*(pp. 1-62). Sacramento, California: California State Lands Commission.
- Freeworldmaps (Cartographer). (2020). Physical Map of California. Retrieved from
 https://www.freeworldmaps.net/united-states/california/map.html
- Frew, N. M., Bock, E. J., Schimpf, U., Hara, T., Haußecker, H., Edson, J. B., ... Jähne, B.
 (2004). Air-sea gas transfer: Its dependence on wind stress, small-scale roughness, and
 surface films. *Journal of Geophysical Research: Oceans, 109*(C8), C08S17.
 doi:10.1029/2003JC002131
- Greinert, J. (2008). Monitoring temporal variability of bubble release at seeps: The
 hydroacoustic swath system GasQuant. *Journal of Geophysical Research*, *113*, C07048.
 doi:10.1029/2007JC004704
- Greinert, J., McGinnis, D. F., Naudts, L., Linke, P., & De Batist, M. (2010). Atmospheric
 methane flux from bubbling seeps: Spatially extrapolated quantification from a Black Sea
 shelf area. *Journal of Geophysical Research*, *115*. doi:10.1029/2009jc005381
- Hanna, S. R., Briggs, G. A., & Hosker Jr., R. P. (1982). *Handbook on Atmospheric Diffusion* (J.
 S. Smith Ed.): Technical Information Center, U.S. Department of Energy.
- Helmig, D., Rossabi, S., Hueber, J., Tans, P., Montzka, S. A., Masarie, K., ... Pozzer, A. (2016).
 Reversal of global atmospheric ethane and propane trends largely due to US oil and
 natural gas production. *Nature Geoscience*, 9(7), 490-495. doi:10.1038/ngeo2721
- Heyer, J., & Berger, U. (2000). Methane emission from the coastal area in the Southern Baltic
 Sea. *Estuarine, Coastal and Shelf Science, 51*(1), 13-30. doi:10.1006/ecss.2000.0616
- Higgs, B., Mountjoy, J. J., Crutchley, G. J., Townend, J., Ladroit, Y., Greinert, J., & McGovern,
 C. (2019). Seep-bubble characteristics and gas flow rates from a shallow-water, highdensity seep field on the shelf-to-slope transition of the Hikurangi subduction margin. *Marine Geology*, 417, 105985. doi:10.1016/j.margeo.2019.105985
- Hmiel, B., Petrenko, V. V., Dyonisius, M. N., Buizert, C., Smith, A. M., Place, P. F., ...
 Dlugokencky, E. (2020). Preindustrial 14CH4 indicates greater anthropogenic fossil CH4
 emissions. *Nature*, 578(7795), 409-412. doi:10.1038/s41586-020-1991-8
- Höglund-Isaksson, L. (2017). Bottom-up simulations of methane and ethane emissions from
 global oil and gas systems 1980 to 2012. *Environmental Research Letters*, *12*(2), 024007.
 doi:10.1088/1748-9326/aa583e
- Hornafius, S. J., Quigley, D. C., & Luyendyk, B. P. (1999). The world's most spectacular marine
 hydrocarbons seeps (Coal Oil Point, Santa Barbara Channel, California): Quantification
 of emissions. *Journal Geophysical Research Oceans, 104*(C9), 20,703-720,711.
 doi:10.1029/1999JC900148
- Hughes, M., Hall, A., & Fovell, R. G. (2007). Dynamical controls on the diurnal cycle of
 temperature in complex topography. *Climate Dynamics*, 29(2), 277-292.
 doi:10.1007/s00382-007-0239-8
- IEA. (2020). *Methane Tracker 2020*. Retrieved from Paris: <u>https://www.iea.org/reports/methane-</u>
 <u>tracker-2020</u>

- IPCC. (2013). Working Group 1 Contribution to the IPCC Fifth Assessment Report Climate
 Change 2013-The Physical Science Basis. Retrieved from IPCC Secretariat, Geneva,
 Switzerland:
- IPCC. (2014). Climate Change 2014: Synthesis Report. Contributions of Working Groups I, II
 and III to the Fifth Assessment Report of the Intergovernmental Panel on Climate
 Change. Retrieved from Geneva, Switzerland: <u>http://www.ipcc.ch/pdf/assessment-</u>
 report/ar5/syr/SYR AR5 FINAL full wcover.pdf
- Jackson, R. B., Saunois, M., Bousquet, P., Canadell, J. G., Poulter, B., Stavert, A. R., ...
 Tsuruta, A. (2020). Increasing anthropogenic methane emissions arise equally from
 agricultural and fossil fuel sources. *Environmental Research Letters*, 15(7), 071002.
 doi:10.1088/1748-9326/ab9ed2
- Johansen, C., Macelloni, L., Natter, M., Silva, M., Woosley, M., Woolsey, A., . . . MacDonald, I.
 R. (2020). Hydrocarbon migration pathway and methane budget for a Gulf of Mexico
 natural seep site: Green Canyon 600. *Earth and Planetary Science Letters*, 545, 116411.
 doi:10.1016/j.epsl.2020.116411
- Jordan, S. F. A., Treude, T., Leifer, I., Janßen, R., Werner, J., Schulz-Vogt, H., & Schmale, O.
 (2020). Bubble-mediated transport of benthic microorganisms into the water column:
 Identification of methanotrophs and implication of seepage intensity on transport *Scientific Reports*, 10(1), 4682. doi:10.1038/s41598-020-61446-9
- Judd, A., & Hovland, M. (2007). *Seabed fluid flow: The impact on geology, biology and the marine environment*. Cambridge, UK: Cambridge University Press.
- Kasaya, T., Mitsuzawa, K., Goto, T.-n., Iwase, R., Sayanagi, K., Araki, E., . . . Nagao, T. (2009).
 Trial of multidisciplinary observation at an expandable sub-marine cabled station "OffHatsushima Island Observatory" in Sagami Bay, Japan. *Sensors*, 9(11), 9241-9254.
 doi:10.3390/s91109241
- Leifer, I. (2010). Characteristics and scaling of bubble plumes from marine hydrocarbon seepage
 in the Coal Oil Point seep field. *Journal Geophysical Research*, *115*(C11), C11014.
 doi:10.1029/2009JC005844
- Leifer, I. (2015). Seabed bubble flux estimation by calibrated video survey for a large blowout
 seep in the North Sea. *Journal of Marine and Petroleum Geology*, 68B, 743-752.
 doi:10.1016/j.marpetgeo.2015.08.032
- Leifer, I. (2019). A synthesis review of emissions and fates for the Coal Oil Point marine
 hydrocarbon seep field and California marine seepage. *Geofluids*, 2019(4724587), 1-48.
 doi:10.1155/2019/4724587
- Leifer, I., & Boles, J. (2005). Turbine tent measurements of marine hydrocarbon seeps on
 subhourly timescales. *Journal of Geophysical Research-Oceans, 110*(C1), C01006.
 doi:10.1029/2003jc002207
- Leifer, I., Boles, J. R., Luyendyk, B. P., & Clark, J. F. (2004). Transient discharges from marine
 hydrocarbon seeps: Spatial and temporal variability. *Environmental Geology*, 46(8),
 1038-1052. doi:10.1007/s00254-004-1091-3
- Leifer, I., Chernykh, D., Shakhova, N., & Semiletov, I. (2017). Sonar gas flux estimation by
 bubble insonification: Application to methane bubble flux from seep areas in the outer
 Laptev Sea. *The Cryosphere*, 11(3), 1333-1350. doi:10.5194/tc-11-1333-2017
- Leifer, I., & Clark, J. F. (2002). Modeling trace gases in hydrocarbon seep bubbles: Application
 to marine hydrocarbon seeps in the Santa Barbara Channel. *Geologiya I Geofizika*, 47(7),
 572-579.

- Leifer, I., Clark, J. F., & Chen, R. F. (2000). Modifications of the local environment by natural
 marine hydrocarbon seeps. *Geophysical Research Letters*, 27(22), 3711-3714.
 doi:10.1029/2000GL011619
- Leifer, I., Jeuthe, H., Gjøsund, S. H., & Johansen, V. (2009). Engineered and natural marine
 seep, bubble-driven buoyancy flows. *Journal of Physical Oceanography*, *39*(12), 30713090. doi:10.1175/2009JPO4135.1
- Leifer, I., Kamerling, M., Luyendyk, B. P., & Wilson, D. (2010). Geologic control of natural
 marine hydrocarbon seep emissions, Coal Oil Point seep field, California. *Geo-Marine Letters, 30*(3-4), 331-338. doi:10.1007/s00367-010-0188-9
- Leifer, I., Luyendyk, B. P., Boles, J., & Clark, J. F. (2006). Natural marine seepage blowout:
 Contribution to atmospheric methane. *Global Biogeochemical Cycles*, 20(3), GB3008.
 doi:10.1029/2005GB002668
- Leifer, I., & MacDonald, I. (2003). Dynamics of the gas flux from shallow gas hydrate deposits:
 interaction between oily hydrate bubbles and the oceanic environment. *Earth and Planetary Science Letters*, 210(3-4), 411-424. doi:10.1016/S0012-821X(03)00173-0
- Leifer, I., Melton, C., Fischer, M. L., Fladeland, M., Frash, J., Gore, W., ... Yates, E. L. (2018).
 Atmospheric characterization through fused mobile airborne and surface in situ surveys:
 Methane emissions quantification from a producing oil field. *Atmospheric Measurement Techniques*, 11(3), 1689-1705. doi:10.5194/amt-11-1689-2018
- Leifer, I., Melton, C., Manish, G., & Leen, B. (2014). Mobile monitoring of methane leakage.
 Gases and Instrumentation, July/August 2014, 20-24.
- Leifer, I., Melton, C., Tratt, D. M., Buckland, K. N., Chang, C., Frash, J., . . . Yurganov, L.
 (2018). Validation of mobile in situ measurements of dairy husbandry emissions by
 fusion of airborne/surface remote sensing with seasonal context from the Chino Dairy
 Complex. *Environmental Pollution*, 242(Pt B), 2111-2134.
- 1021 doi:10.1016/j.envpol.2018.03.078
- Leifer, I., Melton, C., Tratt, D. M., Buckland, K. N., Clarisse, L., Coheur, P., . . . Yurganov, L.
 (2016). Remote sensing and in situ measurements of methane and ammonia emissions
 from a megacity dairy complex: Chino, CA. *Environmental Pollution, 221*, 37-51.
 doi:10.1016/j.envpol.2016.09.083
- Leifer, I., & Patro, R. (2002). The bubble mechanism for methane transport from the shallow
 seabed to the surface: A review and sensitivity study. *Continental Shelf Research*, 22(16),
 2409-2428. doi:10.1016/S0278-4343(02)00065-1
- Leifer, I., Solomon, E., Schneider v. Deimling, J., Coffin, R., Rehder, G., & Linke, P. (2015).
 The fate of bubbles in a large, intense bubble plume for stratified and unstratified water: Numerical simulations of 22/4b expedition field data. *Journal of Marine and Petroleum Geology*, 68B, 806-823. doi:10.1016/j.marpetgeo.2015.07.025
- Liss, P. S., & Duce, R. A. (2005). *The sea surface and global change*: Cambridge University
 Press.
- Liss, P. S., & Merlivat, L. (1986). Air-sea gas exchange rates: Introduction and synthesis. In P.
 Buat-Ménard (Ed.), *THe Role of Air-Sea Exchange in Geochemical Cycling* (Vol. 185).
 Dordrecht: Springer.
- Lu, R., Turco, R. P., & Jacobson, M. Z. (1997). An integrated air pollution modeling system for
 urban and regional scales: 1. Structure and performance. *Journal of Geophysical Research: Atmospheres, 102*(D5), 6063-6079. doi:10.1029/96jd03501

1041 Marinaro, G., Etiope, G., Bue, N. L., Favali, P., Papatheodorou, G., Christodoulou, D., ... Rolin, 1042 J.-F. (2006). Monitoring of a methane-seeping pockmark by cabled benthic observatory (Patras Gulf, Greece). Geo-Marine Letters, 26(5), 297-302. doi:10.1007/s00367-006-1043 1044 0040-4 Mazzini, A., Sciarra, A., Etiope, G., Sadavarte, P., Houweling, S., Pandey, S., & Husein, A. 1045 1046 (2021). Relevant methane emission to the atmosphere from a geological gas 1047 manifestation. Scientific Reports, 11(1), 4138. doi:10.1038/s41598-021-83369-9 1048 Minor, S. A., Kellogg, K. S., Stanley, R. G., Gurrola, L. D., Keller, E. A., & Brandt, T. R. (Cartographer). (2009). Geologic Map of the Santa Barbara Coastal Plain Area, Santa 1049 1050 Barbara County, California. Retrieved from https://pubs.usgs.gov/sim/3001/ Münkel, C. (2007). Mixing height determination with lidar ceilometers - Results from Helsinki 1051 Testbed. Meteorologische Zeitschrift, 16, 451-459. doi:10.1127/0941-2948/2007/0221 1052 Muvakshin, S. I., & Sauter, E. (2010). The hydroacoustic method for the quantification of the gas 1053 flux from a submersed bubble plume. Oceanology, 50(6), 995-1001. 1054 doi:10.1134/S0001437010060202 1055 1056 Nicewonger, M. R., Verhulst, K. R., Aydin, M., & Saltzman, E. S. (2016). Preindustrial atmospheric ethane levels inferred from polar ice cores: A constraint on the geologic 1057 sources of atmospheric ethane and methane. Geophysical Research Letters, 43(1), 214-1058 221. doi:https://doi.org/10.1002/2015GL066854 1059 Nightingale, P. D., Malin, G., Law, C. S., Watson, A. J., Liss, P. S., Liddicoat, M. I., ... Upstill-1060 Goddard, R. C. (2000). In situ evaluation of air-sea gas exchange parameterizations using 1061 novel conservative and volatile tracers. Global Biogeochemical Cycles, 14(1), 373-387. 1062 1063 doi:10.1029/1999GB900091 Nisbet, E. G., Manning, M. R., Dlugokencky, E. J., Fisher, R. E., Lowry, D., Michel, S. E., ... 1064 White, J. W. C. (2019). Very strong atmospheric methane growth in the 4 years 2014-1065 1066 2017: Implications for the Paris Agreement. Global Biogeochemical Cycles, 33(3), 318-342. doi:10.1029/2018GB006009 1067 Olson, D. J. (1983). Surface and subsurface geology of the Santa Barbara Goleta Metropolitan 1068 area, Santa Barbara County, California. (MS). Oregon State University, Retrieved from 1069 https://ir.library.oregonstate.edu/concern/graduate thesis or dissertations/v692tb957?loc 1070 ale=it 1071 1072 Padilla, A. M., Loranger, S., Kinnaman, F. S., Valentine, D. L., & Weber, T. C. (2019). Modern assessment of natural hydrocarbon gas flux at the Coal Oil Point seep field, Santa 1073 Barbara, California. Journal of Geophysical Research: Oceans, 124(4), 2472-2484. 1074 doi:10.1029/2018jc014573 1075 Peischl, J., Ryerson, T. B., Brioude, J., Aikin, K. C., Andrews, A. E., Atlas, E., . . . Parrish, D. D. 1076 (2013). Quantifying sources of methane using light alkanes in the Los Angeles basin, 1077 1078 California. Journal of Geophysical Research: Atmospheres, 118(10), 4974-4990. 1079 doi:10.1002/jgrd.50413 Pozzer, A., Pollmann, J., Taraborrelli, D., Jöckel, P., Helmig, D., Tans, P., ... Lelieveld, J. 1080 (2010). Observed and simulated global distribution and budget of atmospheric 1081 C₂-C₅ alkanes. Atmospheric Chemistray and Physics, 10(9), 1082 4403-4422. doi:10.5194/acp-10-4403-2010 1083 Rahn, D. A., Parish, T. R., & Leon, D. (2017). Synthesis of observations from the Precision 1084 1085 Atmospheric Marine Boundary Layer Experiment (PreAMBLE). Monthly Weather Review, 145(6), 2325-2342. doi:10.1175/mwr-d-16-0373.1 1086

- 1087 Reeburgh, W. S. (2007). Oceanic methane biogeochemistry. *Chemical Reviews*, 107(2), 486-513.
 1088 doi:10.1021/cr050362v
- Reeburgh, W. S., Ward, B. B., Whalen, S. C., Sandbeck, K. A., Kilpatrickt, K. A., & Kerkhof, L.
 J. (1991). Black Sea methane geochemistry. *Deep Sea Research Part A. Oceanographic Research Papers*, *38*, S1189-S1210. doi:<u>https://doi.org/10.1016/S0198-0149(10)80030-5</u>
- Rehder, G., Keir, R. S., Suess, E., & Rhein, M. (1999). Methane in the Northern Atlantic
 controlled by microbial oxidation and atmospheric history. *Geophysical Research Letters*, 26(5), 587-590. doi:10.1029/1999GL900049
- Riedel, M., Scherwath, M., Römer, M., Veloso, M., Heesemann, M., & Spence, G. D. (2018).
 Distributed natural gas venting offshore along the Cascadia margin. *Nature Communications*, 9(1), 3264. doi:10.1038/s41467-018-05736-x
- Römer, M., Hsu, C.-W., Loher, M., MacDonald, I. R., dos Santos Ferreira, C., Pape, T., ...
 Sahling, H. (2019). Amount and fate of gas and oil discharged at 3400 m water depth
 from a natural seep site in the Southern Gulf of Mexico. *Frontiers in Marine Science*,
 6(700). doi:10.3389/fmars.2019.00700
- Römer, M., Riedel, M., Scherwath, M., Heesemann, M., & Spence, G. D. (2016). Tidally
 controlled gas bubble emissions: A comprehensive study using long-term monitoring data
 from the NEPTUNE cabled observatory offshore Vancouver Island. *Geochemistry, Geophysics, Geosystems, 17*(9), 3797-3814. doi:10.1002/2016GC006528
- Römer, M., Sahling, H., Pape, T., Bohrmann, G., & Spieß, V. (2012). Quantification of gas
 bubble emissions from submarine hydrocarbon seeps at the Makran continental margin
 (offshore Pakistan). *Journal of Geophysical Research: Oceans, 117*(C10), C10015.
 doi:10.1029/2011jc007424
- Römer, M., Wenau, S., Mau, S., Veloso, M., Greinert, J., Schlüter, M., & Bohrmann, G. (2017).
 Assessing marine gas emission activity and contribution to the atmospheric methane
 inventory: A multidisciplinary approach from the Dutch Dogger Bank seep area (North
 Sea). *Geochemistry, Geophysics, Geosystems, 18*(7), 2617-2633.
- 1114 doi:10.1002/2017gc006995
- Saunois, M., Stavert, A. R., Poulter, B., Bousquet, P., Canadell, J. G., Jackson, R. B., . . .
 Zhuang, Q. (2020). The global methane budget 2000-2017. *Earth System Science Data*, *12*(3), 1561-1623. doi:10.5194/essd-2019-128
- Sauter, E. J., Muyakshin, S. I., Charlou, J.-L., Schlüter, M., Boetius, A., Jerosch, K., . . . Klages,
 M. (2006). Methane discharge from a deep-sea submarine mud volcano into the upper
 water column by gas hydrate-coated methane bubbles. *Earth and Planetary Science Letters*, 243(3-4), 354-365. doi:10.1016/j.epsl.2006.01.041
- Scherwath, M., Thomsen, L., Riedel, M., Römer, M., Chatzievangelou, D., Schwendner, J., . . .
 Heesemann, M. (2019). Ocean observatories as a tool to advance gas hydrate research. *Earth and Space Science, 6*(12), 2644-2652. doi:10.1029/2019ea000762
- Schmale, O., Beaubien, S. E., Rehder, G., Greinert, J., & Lonmbardi, S. (2010). Gas seepage in
 the Dnepr paleo-delta area (NW-Black Sea) and its regional impact on the water column
 methane cycle. *Journal of Marine Systems, 80*(1-2), 90-100.
 doi:10.1016/j.jmarsys.2009.10.003
- Schmale, O., Greinert, J., & Rehder, G. (2005). Methane emission from high-intensity marine
 gas seeps in the Black Sea into the atmosphere. *Geophysical Research Letters*, 32(7),
 L07609. doi:10.1029/2004gl021138

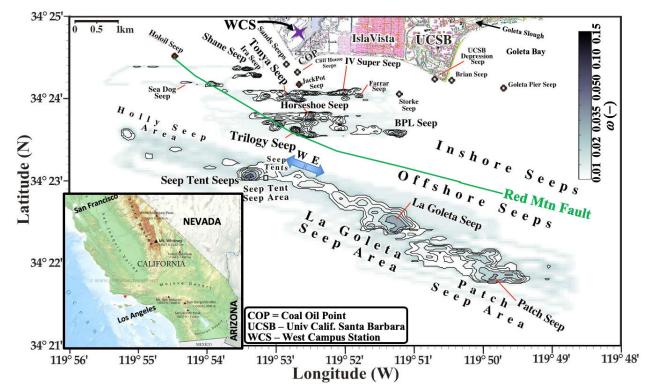
1132	Schmale, O., Leifer, I., Stolle, C., Schneider von Deimling, J., Krause, S., Kießlich, K.,
1133	Treude, T. (2015). Bubble transport mechanism: Indications for a gas bubble-mediated
1134	innoculation of benthic methanotrophs into the water column. Continental Shelf
1135	Research, 103, 70-78. doi:10.1016/j.csr.2015.04.022
1136	Schneider von Deimling, J., Rehder, G., Greinert, J., McGinnis, D. F., Boetius, A., & Linke, P.
1137	(2011). Quantification of seep-related methane gas emissions at Tommeliten, North Sea.
1138	Continental Shelf Research, 31, 867-878. doi:10.1016/j.csr.2011.02.012
1139	Schwietzke, S., Sherwood, O. A., Bruhwiler, L. M. P., Miller, J. B., Etiope, G., Dlugokencky, E.
1140	J., Tans, P. P. (2016). Upward revision of global fossil fuel methane emissions based
1141	on isotope database. Nature, 538(7623), 88-91. doi:10.1038/nature19797
1142	Shakhova, N., Semiletov, I., Leifer, I., Rekant, P., Salyuk, A., & Kosmach, D. (2010).
1143	Geochemical and geophysical evidence of methane release over the East Siberian Arctic
1144	Shelf. Journal of Geophysical Research, 115(C8), C08007. doi:10.1029/2009JC005602
1145	Shakhova, N., Semiletov, I., Salyuk, A., Iossoupov, V., Kosmach, D., & Gustafsson, O. (2010).
1146	Extensive methane venting to the atmosphere from sediments of the East Siberian Arctic
1147	Shelf. Science, 327, 1246-1249. doi:10.1126/science.1182221
1148	Shakhova, N., Semiletov Igor P., Leifer, I., Sergienko, V., Salyuk, A., Kosmach, D.,
1149	Gustafsson, O. (2013). Ebullition and storm-induced methane release from the East
1150	Siberian Arctic Shelf. Nature Geoscience, 7, 64-70. doi:10.1038/ngeo2007
1151	Shindell, D. T., Faluvegi, G., Bell, N., & Schmidt, G. A. (2005). An emissions-based view of
1152	climate forcing by methane and tropospheric ozone. Geophysical Research Letters, 32,
1153	L04803. doi:doi:10.1029/2004GL021900
1154	Simpson, I. J., Sulbaek Andersen, M. P., Meinardi, S., Bruhwiler, L., Blake, N. J., Helmig, D.,
1155	. Blake, D. R. (2012). Long-term decline of global atmospheric ethane concentrations and
1156	implications for methane. <i>Nature</i> , 488(7412), 490-494. doi:10.1038/nature11342
1157	Solomon, E., Kastner, M., MacDonald, I. R., & Leifer, I. (2009). Considerable methane fluxes to
1158	the atmosphere from hydrocarbon seeps in the Gulf of Mexico. <i>Nature Geoscience</i> , 2,
1159	561-565. doi:10.1038/NGEO574
1160	Thompson, D., Leifer, I., Bovensman, H., Eastwood, M., Fladeland, M., Frankenberg, C.,
1161	Thorpe, A. K. (2015). Real-time remote detection and measurement for airborne imaging
1162	spectroscopy: A case study with methane. Atmospheric Measurement Techniques, 8, 1-
1163	46. doi:10.5194/amtd-8-1-2015
1164	Thornton, B. F., Prytherch, J., Andersson, K., Brooks, I. M., Salisbury, D., Tjernström, M., &
1165	Crill, P. M. (2020). Shipborne eddy covariance observations of methane fluxes constrain
1166	Arctic sea emissions. Science Advances, 6(5), eaay7934. doi:10.1126/sciadv.aay7934
1167	Veloso-Alarcón, M. E., Jansson, P., De Batist, M., Minshull, T. A., Westbrook, G. K., Pälike, H.,
1168	Greinert, J. (2019). Variability of acoustically evidenced methane bubble emissions
1169	offshore Western Svalbard. Geophysical Research Letters, 46(15), 9072-9081.
1170	doi:10.1029/2019gl082750
1171	Wanninkhof, R., Asher, W. E., Ho, D. T., Sweeney, C., & McGillis, W. R. (2009). Advances in
1172	quantifying air-sea gas exchange and environmental forcing. Annual Review of Marine
1173	<i>Science</i> , <i>1</i> (1), 213-244. doi:10.1146/annurev.marine.010908.163742
1174	Washburn, L., Johnson, C., Gotschalk, C. G., & Egland, E. T. (2001). A gas capture buoy for
1175	measuring bubbling gas flux in oceans and lakes. <i>Journal of Atmospheric and Oceanic</i>
1176	<i>Technology</i> , 18, 1411-1420. doi:10.1175/1520-0426

- Weber, D., Marquez, B. A., Taylor, C., Raya, P., Contreras, P., Howard, D., . . . Doezema, L. A.
 (2017). Macroseepage of methane and light alkanes at the La Brea tar pits in Los
 Angeles. *Journal of Atmospheric Chemistry*, 74(3), 339-356. doi:10.1007/s10874-0169346-4
- Weber, T. C., Mayer, L., Jerram, K., Beaudoin, J., Rzhanov, Y., & Lovalvo, D. (2014). Acoustic
 estimates of methane gas flux from the seabed in a 6000 km2 region in the Northern Gulf
 of Mexico. *Geochemistry, Geophysics, Geosystems, 15*(5), 1911-1925.
 doi:10.1002/2014gc005271
- Wiggins, S. M., Leifer, I., Linke, P., & Hildebrand, J. A. (2015). Long-term acoustic monitoring
 at North Sea well site 22/4b. *Journal of Marine and Petroleum Geology*, 68, 776-788.
 doi:10.1016/j.marpetgeo.2015.02.011
- Wilson, D., Leifer, I., & Maillard, E. (2015). Megaplume bubble process visualization by 3D
 multibeam sonar mapping. *Journal of Marine and Petroleum Geology*, 68B, 753-765.
 doi:10.1016/j.marpetgeo.2015.07.007
- Zhao, D., Toba, Y., Suzuki, Y., & Komori, S. (2003). Effect of wind waves on air'sea gas
 exchange: Proposal of an overall CO2 transfer velocity formula as a function of breakingwave parameter. *Tellus B: Chemical and Physical Meteorology*, 55(2), 478-487.
- 1194 doi:10.3402/tellusb.v55i2.16747
- 1195

1197NMHCNon Methane Hydro Carbons1198OXGOil and gas1199T/T/CTotal hydrocarbon1200WCSWest Campus Station1201 $C_{ared}(\theta)$ Wind direction-resolved average concentration1202 C_{crit4} Methane concentration1203 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1205 C_{obs} WCS observed concentration1206 C_{scep} Concentration in seep directions1207 C_{sim} WCS simulated concentration1208 $u_{aree}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{max}(\theta)$ Wind direction-resolved maximum u 1211 u_{scep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (botm) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 i Grid cell $i, cating index1221iGrid cell i, cating index1222jGrid cell i, cating index1223K(r, \theta)Wind d$	1196	Table of Nomenclature			
1198O&GOil and gas1199THCTotal hydrocarbon1200WCSWest Campus Station1201 $C_{ave}(\theta)$ Wind direction-resolved average concentration1202 C_{CH4} Methane concentration1203 $C_{mact}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{med}(\theta)$ Wind direction-resolved maximum concentration1205 C_{obs} WCS observed concentration1206 C_{scep} Concentration in seep directions1207 C_{sim}^{s} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved maximum u1210 $u_{mact}(\theta)$ Wind direction-resolved maximum u1211 $u_{meat}(\theta)$ Wind direction-resolved maximum u1212BLBoundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215C'Plume (anomaly) concentrations1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_n Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1221 j Grid cell northing index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction argving, distance correction function to emissions1224 $K(\theta)$ Wind direction-resolved wind speed1233 $\delta(\theta)$ Wind direction angular resolution<	1197	NMHC	Non Methane Hydro Carbons		
1199THCTotal hydrocarbon1200WCSWest Campus Station1201 $C_{gave}(\theta)$ Wind direction-resolved average concentration1202 C_{CH4} Methane concentration1203 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{meel}(\theta)$ Wind direction-resolved median concentration1205 C_{obs} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{5im}^{c} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1211 u_{seep} Wind direction-resolved median u 1212 BL Boundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentrations1218 E_a Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1219 E_{ij} Grid cell northing index1220 E_W Emissions to the water column in the near field1221 i Grid cell northing index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction arrying, distance correction function to emissions1224 i Grid cell northing index1225 r Distance from WCS to cell i, j 1226RResidual in C' after		O&G			
1200WCSWest Campus Station1201 $C_{ave}(\theta)$ Wind direction-resolved average concentration1202 C_{CH4} Methane concentration1203 $C_{max}(\theta)$ Wind direction-resolved median concentration1204 $C_{pred}(\theta)$ Wind direction-resolved median concentration1205 C_{obs} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{sim} WCS simulated concentration1208 $u_{ane}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{med}(\theta)$ Wind direction-resolved maximum u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_a Scabed (bottom) emissions1219 E_{ij} Grid cell northing index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit<	1199	THC			
1201 $C_{ave}(\theta)$ Wind direction-resolved average concentration1202 C_{cH4} Methane concentration1203 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{med}(\theta)$ Wind direction-resolved maximum concentration1205 C_{bbs} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{sim}^{s} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind direction-resolved median u 1212 BL Boundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentrations1218 E_a Scabed (bottom) emissions1219 E_i Grid cell easting index1222 j Grid cell easting index1223 $K(r, \theta)$ Wind direction and distance orrection function to emissions1224 $K(\theta)$ Wind direction and sitance resolved correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind direction angular resolution1234 $\phi(u)$ Wind direction angular resolution1235 $\phi(d,C)$		WCS			
1202 C_{CH4} Methane concentration1203 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{gaed}(\theta)$ Wind direction-resolved median concentration1205 C_{obs} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{sim}^{*} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{max}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction ant time-resolved average concentration1215C'Plume (anomaly) concentrations1217 E_A Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 j Grid cell easting index1221 i Grid cell nothing index1222 j Grid cell nothing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction-resolved wind speed1225 r Distance from WCS to cell i, j 1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1238 t Time12		$C_{ave}(\theta)$			
1203 $C_{max}(\theta)$ Wind direction-resolved maximum concentration1204 $C_{med}(\theta)$ Wind direction-resolved median concentration1205 C_{obs} WCS observed concentration1206 C_{ssep} Concentration in seep directions1207 C'_{sim} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell northing index1222 j Grid cell cordinate system in wind reference frame1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions124 $K(\theta)$ Wind direction angular resoluted125 r Distance from WCS to cell i, j 1223 $K(r, \theta)$ Wind direction angular resolution124 $K(\theta)$ Wind direction angular resoluti			-		
1204 $C_{med}(\theta)$ Wind direction-resolved median concentration1205 C_{obs} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{sim} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved median u 1209 $u_{max}(\theta)$ Wind direction-resolved median u 1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215C'Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 $E_{i,j}$ Grid cell i, j atmospheric emissions1220 Ew Emissions to the water column in the near field1221iGrid cell arsting index1222jGrid cell arsting index1223 $K(r, \theta)$ Wind direction varying, distance correction function to emissions1224 x_i Time1225rDistance from WCS to cell i, j 1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229uWind direction angular resolution1233 $\delta(\theta)$ Model w			Wind direction-resolved maximum concentration		
1205 C_{obs}^{i} WCS observed concentration1206 C_{seep} Concentration in seep directions1207 C_{sinn}^{i} WCS simulated concentration1208 $u_{anex}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{meat}(\theta)$ Wind direction-resolved maximum u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind direction angular resolution1233 $\delta\theta$ Model wind direction angular resolution1234 (ω) Wind direction and vind speed-resolved probability distribution1					
1206 C_{ssep} Concentration in seep directions1207 C_{sim} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u 1209 $u_{max}(\theta)$ Wind direction-resolved median u 1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 $E_{i,j}$ Grid cell i, j atmospheric emissions1220jGrid cell easting index1221iGrid cell easting index1222jGrid cell northing index1223 $K(r, \theta)$ Wind direction varying, distance correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1230 $u(\theta)$ Wind direction angular resolved probability distribution1234 $\phi(u)$ Wind direction and speed-resolved probability distribution1232 Y No		C'obs			
1207 C_{Sim}^{Sim} WCS simulated concentration1208 $u_{ave}(\theta)$ Wind direction-resolved average u1209 $u_{max}(\theta)$ Wind direction-resolved maximum u1210 $u_{med}(\theta)$ Wind direction-resolved median u1211 u_{scep} Wind speed in seep directions1212BLBoundary layer height1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215C'Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217EAAtmospheric emissions1218EBSeabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220EWEmissions to the water column in the near field1221iGrid cell asting index1222jGrid cell onthing index1223K(r, 0)Wind direction varying, distance correction function to emissions1224K(0)Wind direction varying, distance correction function to emissions1225rDistance from WCS to cell i, j 1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1230 $u(\theta)$ Wind direction-resolved wind speed1231 x, y Cartesian coordinate system in wind reference frame1232YNorthing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direc			Concentration in seen directions		
1208 $u_{ave}(\theta)$ Wind direction-resolved maximum u 1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 Ew Emissions to the water column in the near field1221 i Grid cell asting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction varying, distance correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distr			-		
1209 $u_{max}(\theta)$ Wind direction-resolved maximum u 1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Scabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell asting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction varying, distance correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind direction angular resolution1231 x, y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(R$					
1210 $u_{med}(\theta)$ Wind direction-resolved median u 1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 $E_{i,j}$ Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell asting index1222 j Grid cell casting index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\partial \theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 <t< td=""><td></td><td></td><td></td></t<>					
1211 u_{seep} Wind speed in seep directions1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell asting index1222 j Grid cell casting index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\phi)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient128 t Time1299 u Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 $f(\theta, C)$ Wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta, C)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta, C)$ Normalized atmospheric emissio					
1212 BL Boundary layer height1213 C Concentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell easting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1230 $u(\theta)$ Wind direction-resolved wind speed1231 x_{3y} Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\partial \theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and one concentration-resolved probability distribution1235 $\phi(\theta,C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(d,C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta,C)$ <t< td=""><td></td><td></td><td></td></t<>					
1213CConcentration1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215C'Plume (anomaly) concentration1216Ci-C6Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_w Emissions to the water column in the near field1221iGrid cell easting index1222jGrid cell easting index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225rDistance from WCS to cell i, j 1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228tTime1230 $u(\theta)$ Wind direction-resolved wind speed1231 x, y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(d, C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta, C)$ Normalized atmospheric emissions probability1236 $\phi(d, C)$					
1214 $C(t, \theta)$ Wind direction and time-resolved average concentration1215 C' Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_4 Atmospheric emissions1218 E_8 Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell orthing index1222 j Grid cell onrthing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x, y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\theta, U)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\omega)$ Sonar return probability distribution1236 $\phi(\theta, U)$					
1215C'Plume (anomaly) concentration1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell easting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x, y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\mathcal{R}C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\mathcal{R}L)$ Normalized atmospheric emissions probability1233 $\delta\theta$ Model wind direction and wind speed-resolved probability distribution1234 $\phi(\omega)$ Sonar return probability distribution1235 $\phi(\mathcal{R}L)$ <td></td> <td></td> <td></td>					
1216 C_1 - C_6 Methane to hexane concentrations1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell easting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\partial, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\partial, C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(\alpha)$ Sonar return probability distribution1236 $\phi(\partial, C)$ Normalized atmospheric emissions probability1231 $\phi(\omega)$ Normalized sonar return probability distribution1235 <td></td> <td></td> <td></td>					
1217 E_A Atmospheric emissions1218 E_B Seabed (bottom) emissions1219 E_{ij} Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell asting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi_n(E_A)$ Normalized atmospheric emissions probability1237 $\phi(\omega)$ Sonar return probability distribution1236 $\phi_n(E_A)$ Normalized atmospheric emissions probability1237 $\phi(\omega)$ Sonar return1241 ω Sonar return1242 ψ Wind direction1243 ζ Relative					
1218 $E_{\mathcal{B}}$ Seabed (bottom) emissions1219 $E_{i,j}$ Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell easting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(Q,C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(R, C)$ Wind direction and wind speed-resolved probability distribution1235 $\phi(Q,C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi_n(E_A)$ Normalized atmospheric emissions probability1237 $\phi_n(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized sonar return probability distribution1236 $\phi(R, C)$ Wind direction1237 $\phi_n(\omega)$ Normalized sonar return probability distribution </td <td></td> <td></td> <td></td>					
1219 $E_{i,j}$ Grid cell i, j atmospheric emissions1220 E_W Emissions to the water column in the near field1221 i Grid cell easting index1222 j Grid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions			•		
1220 E_W Emissions to the water column in the near field1221iGrid cell easting index1222jGrid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind direction-resolved wind speed1230 $u(\theta)$ Wind direction angular resolution1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1221iGrid cell easting index1222jGrid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta, C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta, C)$ Normalized atmospheric emissions probability1237 $\phi(\omega)$ Normalized sonar return probability distribution1241 ω Sonar return1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions		-			
1222jGrid cell northing index1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta,C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta,C)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1223 $K(r, \theta)$ Wind direction and distance-resolved correction function to emissions1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind direction and concentration-resolved probability distribution1235 $\phi(\theta,C)$ Wind direction and wind speed-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1224 $K(\theta)$ Wind direction varying, distance correction function to emissions1225 r Distance from WCS to cell i, j 1226 R Residual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228 t Time1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions			•		
1225rDistance from WCS to cell i, j 1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228tTime1229uWind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232YNorthing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1226RResidual in C' after Gaussian functional fit1227 R^2 Correlation coefficient1228tTime1229uWind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232YNorthing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1227 R^2 Correlation coefficient1228tTime1229uWind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232YNorthing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1228tTime1229uWind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232YNorthing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1229 u Wind speed1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1230 $u(\theta)$ Wind direction-resolved wind speed1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1231 x,y Cartesian coordinate system in wind reference frame1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions			-		
1232 Y Northing offset of WCS1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1233 $\delta\theta$ Model wind direction angular resolution1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1234 $\phi(u)$ Wind probability distribution1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions		-	•		
1235 $\phi(\theta,C)$ Wind direction and concentration-resolved probability distribution1236 $\phi(\theta,u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions			-		
1236 $\phi(\theta, u)$ Wind direction and wind speed-resolved probability distribution1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions			· ·		
1237 $\phi(\omega)$ Sonar return probability distribution1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1238 $\phi_n(E_A)$ Normalized atmospheric emissions probability1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1239 $\phi_n(\omega)$ Normalized sonar return probability distribution1240 θ Wind direction1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions					
1240 θ Wind direction 1241 ω Sonar return 1242 ψ Wind veering 1243 ζ Relative inshore and offshore emissions	1238	$\phi_n(E_A)$	Normalized atmospheric emissions probability		
1241 ω Sonar return1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions	1239	$\phi_n(\omega)$	Normalized sonar return probability distribution		
1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions	1240	θ	Wind direction		
1242 ψ Wind veering1243 ζ Relative inshore and offshore emissions	1241	ω	Sonar return		
1243 ζ Relative inshore and offshore emissions					
		,			
		-			

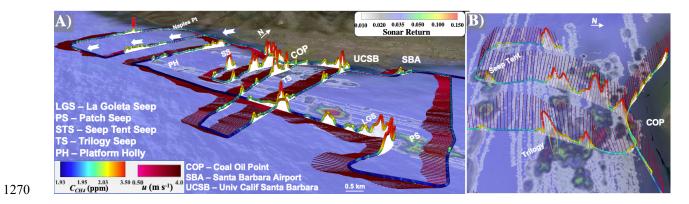
1247	Gas	Fraction	Emissions	Emissions
1248		(%)	$(m^3 dy^{-1})$	$(Mg yr^{-1})$
1249	CH ₄	88.5	73,900	19,300
1250	C_2H_6	3.1	2,590	1270
1251	C_3H_8	4.2	3,510	2520
1252	C4H10	2.76	2,300	2180
1253	C5H12	1.11	930	1090
1254	$C_{6}H_{14}$	0.13	110	150
1255	C_6H_6	4.4x10 ⁻⁵	4.0	4.7
1256	C7H16	0.04	33	55
1257	C_7H_8	1.0x10 ⁻⁵	1.0	1.3
1258	NMHC	11.3	9470	7280
1259				
1260	$C_1 - C_7^*$	85	83,400	26,600
1261	CO ₂	15	18,000	12,900
1262	* C_1 - C_7 = THC			
1263				

1246	Table 1. A	Atmospheric plume o	composition and m	odel atmospheric emiss	sions.
1247	Cas	Ensation	- Englishing	Enviraina	



1266

Figure 1: Sonar return, ω , map after Leifer et al. (2010). Purple star marks West Campus Station (WCS). Seep names are informal (Table S3), font size corresponds to strength. E-W arrow segregates east and west offshore seepage. Data keys on panels. Inset shows S. California, red dot marks COP seep field. California inset map from Freeworldmaps (2020).



- Figure 2: A) COP seep field methane, C_{CH4}, and wind, u, data for 28 May 2016. White arrows show canyon offshore flow. Red
- 1271 1272 1273 1274 1275 arrows show unmapped seepage to the west of the COP seep field. B) C_{CH4} and u showing Gaussian plume model for Trilogy Seep. Sonar return, ω , map in background. Data key and seep name key on panel. Displayed in GoogleEarth environment. See Supp.
- Fig. S6 for overhead view.



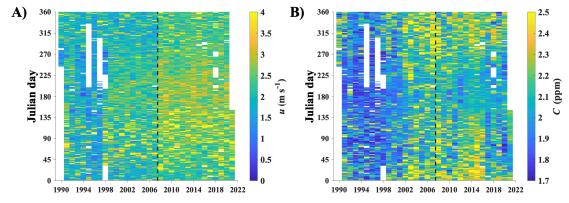
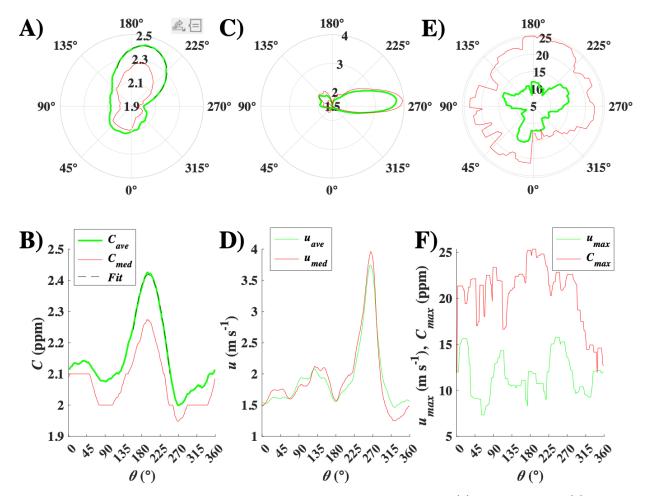


Figure 3: A) Daily mean wind speed, u, and **B**) concentration, C. Data key on figure. WCS upgrade on Jan 2008 is shown by a dashed black line. Supp. Fig. S4 shows raw dataset.



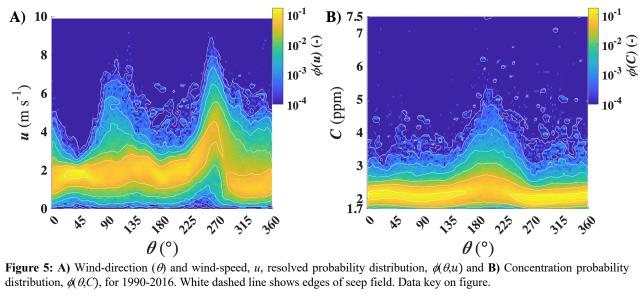
1284

Figure 4: A, B) Concentration, C, versus wind-direction, θ , 1990-2021 for average, $C_{ave}(\theta)$, and median, $C_{med}(\theta)$, and fit to $C_{ave}(\theta)$ for $155 < \theta < 250^{\circ}$. Data key on panel B. C, D) Wind speed, u, average, $u_{ave}(\theta)$, and median, $u_{med}(\theta)$, Data key on panel

D. and E, F) Maximum C, $C_{max}(\theta)$, and wind speed, $u_{max}(\theta)$. Data key on panel F. Polar plot oriented as at WCS facing the

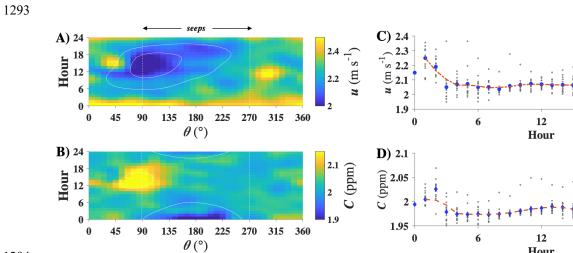
1287

COP seep field.

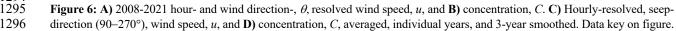


1290 1291









¹²⁹⁶ 1297 Midnight data missing due to daily calibration.

18

18

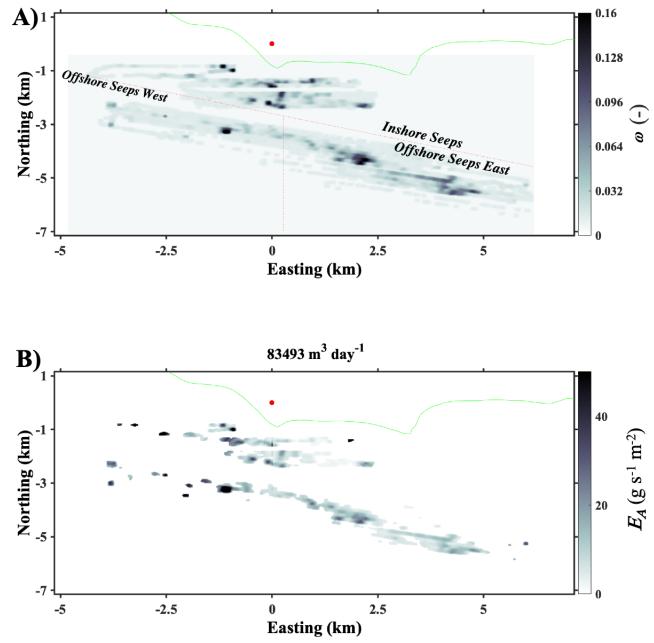
Hour

. ave 24

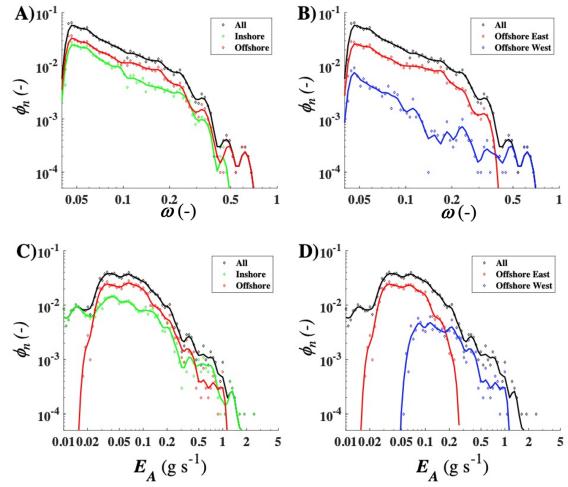
24

2008-2021

smooth



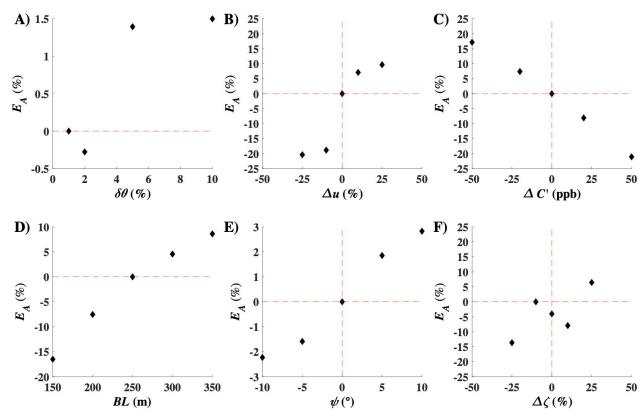
1299Easting (km)1300Figure 7: A) Sonar return, ω, gridded at 22-m resolution. B) Atmospheric emissions, E_A. West campus station (red dot) is at
coordinate system origin. Green line is coast line.



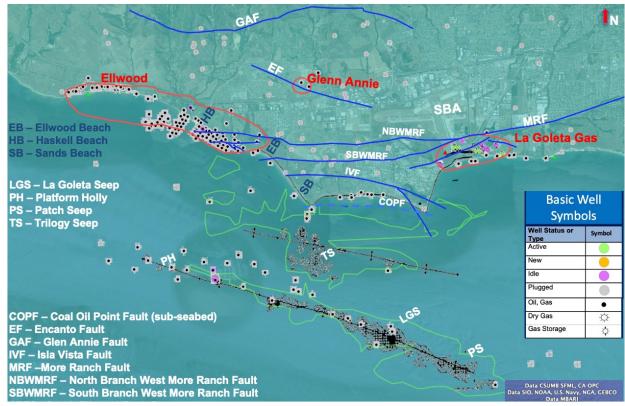
1303 1304

Figure 8: A) Sonar return, ω , occurrence probability, $\phi_n(\omega)$, for all seepage, inshore and offshore seepage and B) all seepage, 1305 1306 offshore east seepage, and offshore west seepage. C) Atmospheric emission, E_A , occurrence probability, $\phi_n(E_A)$, for all seepage, inshore and offshore seepage and D) all seepage, offshore east seepage, and offshore west seepage. Data key on panels.





BL (m) ψ (°) $\Delta \zeta$ (%)1310Figure 9: Emissions, E_A , sensitivity to uncertainty in A) model angular resolution, $\delta\theta$, B) wind speed variation, Δu , C)1311concentration anomaly variation, $\Delta C'$, D) boundary layer thickness, BL, E) wind veering, ψ , and F) inshore/offshore partition1312variation, $\Delta \zeta$. Note different units on different plots. See text for details.



1314 1315 Figure 10: Map of the Goleta Plains oil and gas fields, wells, and the Coal Oil Point (COP) seep field. Grey hatch shows 1995

field extent, green outlines the 1940 field extent is from Leifer (2019). Field locations from Olson (1983). Well data from CDOGGR 1317 (2018). Faults from Minor et al. (2009). Seep names are informal. Data keys on panels. Shown in the Google Earth environment.