Long-Term Atmospheric Emissions for the Coal Oil Point Natural Marine Hydrocarbon Seep Field, Offshore California

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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 11 Gaussian with respect to wind direction, θ . An inversion model of the seep field, θ -resolved anomaly, THC'(θ), 12 derived atmospheric emissions. The model inversion is for the far field, which was satisfied by gridding the sonar seepage and treating each grid cell as a separate Gaussian plume. This assumption was validated by offshore in situ 13 14 offshore data that showed major seep area plumes were Gaussian. Of plume Total Carbon, TC, (TC=THC+carbon 15 dioxide, CO2,+carbon monoxide) 18% was CO2 and 82% was THC; methane. 85% of THC was CH4. These 16 compositions were similar to the seabed composition, demonstrating efficient vertical plume transport of dissolved 17 seep gases. Air samples also measured atmospheric alkane plume composition. The inversion model used observed winds and derived the three-decade-average (1990-2021) field-wide atmospheric emissions of 83,400±12,000 m3 THC 18 19 d_{t}^{-1} (27 Gg *THC* yr⁻¹ based on 19.6 g mole⁻¹ for *THC*). Based on a 50:50 air to seawater partitioning, this implies 20 seabed emissions of 167,000 m³ THC d⁻¹. Based on atmospheric plume composition, C₁-C₆ alkane emissions were 19, 21 1.3, 2.5, 2.2, 1.1, and 0.15 Gg yr⁻¹, respectively. If CH₄ emissions were dispersed over the ~6.3 km² of 25x25 m² bins with sonar values above noise, we find 5.7 μ M m⁻² s⁻¹. The approach can be extended to derive emissions from other 22 dispersed sources such as landfills, industrial sites, or terrestrial seepage if source locations are constrained spatially. 23

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25 1 Introduction

26 1.1 Seepage and methane

On decadal timescales, the important greenhouse gas, methane, CH₄, affects atmospheric radiative balance far more strongly than carbon dioxide, CO₂ (IPCC, 2007, Fig. 2.21), yet CH₄ has large uncertainties for many sources (IPCC, 2013) and is very sensitive to hydroxyl (OH) concentration, the primary CH₄ loss mechanism (Y. Zhao et al., 2020). Since pre-industrial times, CH₄ emissions have risen by a factor of ~2.5, and after stabilizing in the 1990s and early 2000s, resumed rapid growth since 2007 (Nisbet et al., 2019). The significantly shorter lifetime of CH₄ than CO₂ argues for CH₄ regulatory priority as emission reductions (and changes to the radiative balance) manifest more quickly as atmospheric concentrations decrease (Shindell, Faluvegi, Bell, & Schmidt, 2005). Further impetus for a CH₄ focus Deleted: ay

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is a recent estimate that 40% CH₄ emissions reductions are feasible at no net cost for the oil and gas, O&G, industry
 (IEA, 2020), a major anthropogenic CH₄ source (IPCC, 2014). This is particularly salient given a recent estimate that
 half of recent CH₄ increases are from the O&G industry (Jackson et al., 2020).

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40 For 2008-2017, global CH4 top-down emissions estimates are 576 Tg yr⁻¹; 1 Tg=10¹² g, range 550-594 Tg yr⁻¹, whereas 41 bottom-up approaches find 737 Tg yr⁻¹, range 594-881 Tg yr⁻¹ (Saunois et al., 2020). A significant fraction of this 42 discrepancy arises from uncertainty in OH concentration trends and spatial variability (Y. Zhao et al., 2020). 43 Anthropogenic sources for 2008-2017 were estimated at 336-376 Tg CH4 yr⁻¹ based on bottom-up estimates. Natural 44 sources include wildfires, wetlands, hydrates, and geological seepage, Bottom-up estimates for natural sources are 45 higher than top-down estimates including for geological sources (Saunois et al., 2020). Geological sources (including 46 seepage) are estimated at 63-80 Tg CH4 yr⁻¹ of which marine seepage is estimated to contribute 20-30 Tg CH4 yr⁻¹ 47 (Etiope, Ciotoli, Schwietzke, & Schoell, 2019) or 5-10 Tg CH4 yr⁻¹ (Saunois et al., 2020). For comparison, marine 48 non-geological CH4 emissions are estimated at 4-10 Tg yr¹. The broad range of this estimate arises from the 49 uncertainty in the fraction of seabed emissions that reaches the atmosphere and the uncertainty in overall seabed emissions. Further complexity in assessing geological seepage CH4 emissions arise because both seepage and O&G 50 51 emissions source from the same geological reservoirs (Leifer, 2019) and thus are isotopically similar (Schwietzke et 52 al., 2016).

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54 Seepage is process by which petroleum hydrocarbon gases and fluids in the lithosphere migrate to the hydrosphere 55 and/or atmosphere from a reservoir formation which underlies a capping layer that seals the formation, allowing 56 hydrocarbon accumulation. Thus, seepage requires a migration pathway, typically fracture and/or fault networks, 57 through the capping rock layer(s) (Ciotoli, Procesi, Etiope, Fracassi, & Ventura, 2020) or where the capping layer has 58 eroded away, as an outcropping of the reservoir formation (Abrams, 2005).

60 Marine seepage is widespread in every sea and ocean (Judd & Hovland, 2007) and occurs primarily (but not 61 exclusively) in petroleum systems and mostly in convergent basins (Ciotoli et al., 2020). Quantitative seepage estimates (for global budgets) are limited (though growing); see Leifer (2019) review and below for more recent. 62 Fluxes for individual marine seep vents and seep areas have been reported for the Gulf of Mexico (C. Johansen et al., 63 64 2020; Caroline Johansen, Todd, & MacDonald, 2017; Leifer & MacDonald, 2003; Römer et al., 2019; T. C. Weber et 65 al., 2014), the Black Sea (Greinert, McGinnis, Naudts, Linke, & De Batist, 2010), the southern Baltic Sea (Heyer & 66 Berger, 2000), various sectors of the North Sea (Borges, Champenois, Gypens, Delille, & Harlay, 2016; Leifer, 2015; 67 Römer et al., 2017), offshore Norway (Muyakshin & Sauter, 2010; Sauter et al., 2006) offshore Svalbard in the Norwegian Arctic (Veloso-Alarcón et al., 2019), offshore Pakistan (Römer, Sahling, Pape, Bohrmann, & Spieß, 2012), 68 69 the arctic Laptev Sea (Leifer, Chernykh, Shakhova, & Semiletov, 2017), the East Siberian Arctic Sea (Shakhova et 70 al., 2013), the South China Sea (Di, Feng, Tao, & Chen, 2020), New Zealand's Hikurangi Margin (Higgs et al., 2019), 71 the Cascadia Margin (Riedel et al., 2018), and the Coal Oil Point (COP) marine hydrocarbon seep field, hereafter COP (Deleted: among others

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seep field, in the northern Santa Barbara Channel, offshore Southern California (Hornafius, Quigley, & Luyendyk,
 1999), and for numerous individual vents in the field (Leifer, 2010).

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Most seep emission estimates <u>are</u> snapshot <u>values</u> 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, & Roberts, 2010) to decadal (Fischer, 1978; Leifer, 2019). Additional temporal variability arises from transient emissions – 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.

92 Seepage contributes to oceanographic budgets and to a lesser extent, to atmospheric budgets due to water column 93 losses with significant uncertainty in the partitioning. As a result, uncertainty in the atmospheric contribution is much 94 larger than the (significant) uncertainty in seabed emissions. Seepage partitioning between the atmosphere and ocean 95 - where microbial degradation occurs on timescales inversely related to concentration (Reeburgh et al., 1991), depends 96 primarily on depth (Leifer & Patro, 2002) with little to none of deepsea seabed emissions reaching the atmosphere, 97 e.g., Römer et al. (2019). In contrast, very shallow seepage (meter scale) largely entirely reaches the atmosphere both by direct bubble-meditated transfer and diffusive transport. For intermediate depths, the ocean/atmospheric 98 99 partitioning is complex and depends on depth, bubble flux, bubble size distribution, bubble interfacial conditions, and 100 other characteristics (Leifer & Patro, 2002). Whereas the indirect diffusive flux (proximate and distal) depends on 101 bubble dissolution depth (Leifer & Patro, 2002), vertical turbulence transport in the winter wave-mixed layer (Rehder, Keir, Suess, & Rhein, 1999), microbial oxidation losses, and exchange through the sea-air interface. 102

104 A range of approaches have been used to estimate the sea-air flux. The most common is by measuring the atmospheric 105 and water concentrations and applying air-sea gas exchange theory for the measured wind speeds, e.g., Schmale, 106 Greinert, and Rehder (2005) for Black Sea seepage under weak wind speeds. Sea-air exchange is a diffusive turbulence 107 transfer process that depends on the air-sea concentration difference and the piston velocity, k_T , which depends on gas 108 physical properties, wind speed, u (Liss & Duce, 2005), wave development (D. Zhao, Toba, Suzuki, & Komori, 2003) 109 including wave breaking (Liss & Merlivat, 1986), and surfactant layers at low wind speeds that suppress gas exchange 110 (Frew et al., 2004). k_T increases rapidly and non-linearly with u and has been parameterized by piecewise linear 111 functions (Wanninkhof, Asher, Ho, Sweeney, & McGillis, 2009) or by a cubic function (Nightingale et al., 2000). Air-112 sea gas exchange theory is for (relatively) homogeneous atmospheric and oceanographic fields (concentrations, winds, 113 and wave development), and thus is inappropriate for point-source (bubble-plume) emissions and for the near-field 114 downcurrent plume, which tend to be heterogeneous. 115

116 Another approach uses seabed bubble size measurements or an assumed bubble size distribution to initialize a 117 numerical bubble propagation model to predict direct bubble-mediated atmospheric fluxes (Leifer et al., 2017; Römer Deleted: which

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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.

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An alternate approach is to derive atmospheric emissions by plume inversion. Leifer, Luyendyk, Boles, and Clark (2006) derived emissions for a blowout from Shane Seep in the COP seep field by a plume inversion. This neglected the portion that dissolves during bubble rise and drifts downcurrent, out of the bubble plume's vicinity before sea-air gas transfer into the atmosphere. Note dissolved gas evasion in the plume vicinity contributes to the inversion emissions estimate.

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132 1.2 Study motivation

133 In this study, we present a novel approach for assessing nearshore seepage atmospheric emissions - air quality station data modeling, specifically using a Gaussian plume inversion model. This model requires that source locations are 134 135 mapped, spatially stable, and lie within a fairly constrained distance range band. These conditions are met for the COP 136 seep field, which is near the West Campus air quality Station (WCS). COP seep field lies in shallow coastal waters of 137 northern Santa Barbara Channel, CA. Spatial constraint is provided by geological structures, such as faults, that 138 constrain emission locations. The Gaussian plume model assumes a far-field source, whereas WCS is in the nearfield 139 of the extensive COP seep field. To satisfy the far field criterion, the source was gridded and each grid cell's emissions 140 treated as a distinct (distant) Gaussian plume. This characterization was validated in an offshore survey of several 141 focused COP seep field seepage areas, which were well-modeled as Gaussian plumes. 142

143 Thus, this study demonstrates an approach to deriving emissions from air quality station data for an area source such 144 as natural marine seepage. This approach could be used to derive emissions from other dispersed sources such as

145 landfills, industrial sites, or natural terrestrial seepage where the source locations are, constrained spatially.

147 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.

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- 153 As seep bubbles rise, they dissolve, losing gas to the surrounding water at a rate that decreases with time. Smaller and
- 154 more soluble gases dissolve faster than larger and less soluble gases, i.e., fractionation (Leifer & Patro, 2002).
- 155 Additionally, larger bubbles transport their contents upwards more efficiently than smaller bubbles (Leifer et al.,
- 156 2006). Sufficiently large bubbles reach the sea surface with a significant fraction of their seabed CH₄ from depths of

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even hundreds of meters (Solomon, Kastner, MacDonald, & Leifer, 2009). There are synergies, too with higher plume 165 fluxes driving a stronger upwelling flow that transports plume fluids with dissolved gases upwards towards the surface where air-sea gas exchange drives evasion (Leifer, Jeuthe, Gjøsund, & Johansen, 2009). Another synergy arises from 166 elevated dissolved plume CH4 concentration (Leifer, 2010; Leifer et al., 2006), which slows dissolution. Also, bubbles 167 168 are oil-coated, which slows dissolution. 169

170 Moreover, gases in bubbles that dissolve in the wave-mixed layer (or reach it by the upwelling flow) then diffuse to 171 the air-sea interface due to wave and wind turbulence. Note, microbial degradation removes a portion of the dissolved

172 CH4, which therefore never reaches the air-sea interface. Thus, there are two timescales that govern the fraction that

173 evades - the microbial degradation timescale, which increases as concentrations decrease, and the diffusion timescale,

174 which decreases with increasing wind speed. As a result, there is a dissolved plume that drifts downcurrent, with

175 evasion from this drifting plume creating a linear-source atmospheric plume. Note, dissolved plume concentrations

176 slowly decrease with time (downcurrent distance) from sea-air gas exchange losses, microbial oxidation, and dispersion, leading to a decreased atmospheric plume.

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179 1.4 Atmospheric Gaussian plumes

180 Strong focused atmospheric plumes are created from seep plume bubbles bursting at the sea surface and from dissolved

181 gas evasion within the bubble surfacing footprint. This evasion is enhanced by water-side turbulence from rising and

182 bursting bubbles (Leifer et al., 2015). Atmospheric plume evolution is described by the Gaussian plume model (Hanna,

183 Briggs, & Hosker Jr., 1982), which relates downwind concentrations to wind transport and turbulence dispersion and

184 is the basis of the inversion calculation (see Supp. Sec. S1 for details).

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186 1.5 Setting

187 1.5.1 The Coal Oil Point seep field

188 The COP seep field (Fig. 1) is one of the largest seep fields in the world, with estimated 1995-1996 seabed total 189 <u>hydrocarbon</u>, *THC*, emissions, E_B , of $1.5 \times 10^5 \pm 2 \times 10^4$ m³ d⁻¹ (Hornafius et al., 1999). Hereafter emissions and 190 concentrations are for THC unless noted. Clark, Washburn, Hornafius, and Luyendyk (2000) estimated that half the 191 COP seep field E_B reach the atmosphere in the near field. This is due to shallowness, bubble oiliness, high plume 192 bubble densities, and turbulence mixing within the wave mixed layer. 193 194 Geological structures play a critical role in the spatial distribution of seepage (Leifer, Kamerling, Luyendyk, & Wilson,

195 2010), which lies along several trends in waters from a few meters to ~85 m deep. These trends follow geologic 196 structures including anticlines, synclines, and faults in the reservoir formation, the Monterey Formation and overlying

197 Sisquoc Formation. Faults and fractures associated with damage zones provide migration pathways with seepage

198 scattered non-uniformly along the trends, including focused seep areas that are highly active, localized, and often are

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210 associated with crossing faults and fractures (Leifer et al., 2010). Seepage in these areas typically surrounds a focus

211 and decreases with distance, primarily along linear trends (Leifer, Boles, Luyendyk, & Clark, 2004). See Supp. Table 212

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S3 for informal names and locations of selected focused seep areas.

#### 1.5.2 Coal Oil Point seep field emissions and composition 214

215 COP seep field sources from the South Ellwood oil field whose primary source rock is Monterey Formation, which is 216 immature to marginally mature. Petroleum gases from marine organic materials have relatively higher proportion of 217 ethane, propane, butane, etc., relative to methane as compared to petroleum gases from terrestrial organic materials. 218 The wet gas fraction (C2-C5/C1-C5) indicates a thermogenic origin of greater than 0.05 (Abrams, 2017). Of the 219 saturated alkanes, the alkenes (olefins) are of biological origin. Additionally, the ethane/ethene ratio and 220 propane/propene ratios can be indicators of seep gas biogenic modification with values above 1000 indicating purely 221 thermogenic origin (Abrams, 2017; Bernard, Brooks, & Zumberge, 2001).

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

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

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

alkanes, seabed bubbles are 90.4% CH4 and 8.6% NMHC. CO2 rapidly escapes the bubbles and is negligible (<1%) 226 227 at the sea surface. At the sea surface, CH4 in bubbles is ~90% with NMHC making up the remaining 10%, neglecting

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

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

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#### 231 1.5.3 Northern Santa Barbara Channel climate

232 Diurnal and seasonal wind cycles are important to the atmospheric transport of COP seep field emissions. The Santa 233 Barbara climate is Mediterranean with a dry season and a wet seasons when storms occur infrequently (Dorman & 234 Winant, 2000). The semi-permanent eastern Pacific high-pressure system plays a dominant controlling role in weather 235 in the Santa Barbara coastal plain. This high-pressure system drives light winds and strong temperature inversions that 236 act as a lid that restricts convective mixing to lower altitudes. The coastal California boundary layer is shallow, from 237 0 to 800 m thick (Edinger, 1959); generally 240-300 m around Santa Barbara (Dorman & Winant, 2000). Additionally, 238 coastal mountains provide physical barriers to transport (Lu, Turco, & Jacobson, 1997).

239 240 As a coastal environment, the land/sea breeze is important to overall wind-flow patterns with weak offshore night 241 winds and stronger onshore afternoon winds (Dorman & Winant, 2000). In coastal Santa Barbara, warming on 242 mountaintops and more interior arid lands relative to cooler marine temperatures drives the sea breeze. Downslope 243 nocturnal flows warm nocturnal surface temperatures, moderating the coastal diurnal temperature cycle (Hughes, Hall,

244 & Fovell, 2007). Formatted: Font: Italic

Formatted: Font: Italic Formatted: Font: Italic 246 Typical morning winds are calm and offshore and often accompanied by a cloud-filled marine boundary layer, 50-

247 150 m thick (Lu et al., 1997). The marine layer usually (but not always) "burns off" mid-morning after which

- 248 temperatures rise, the boundary layer thickens and winds shift clockwise from offshore to eventually prevailing
- 249 westerlies aligned with the coastal mountains. Midday through late afternoon and even evening, winds strengthen,

often leading to whitecapping before the boundary layer collapses and winds return to the nocturnal pattern.

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#### 252 2 Methods

#### 253 2.1 West Campus Station data

254 WCS data includes wind speed, u, and direction,  $\theta$  by a vane anemometer (010C,020C, Met One, Grants Pass, OR)

and *THC* concentration, *C*, by a Flame Ionization <u>Detector (51i-LT, Thermo Scientific, MA)</u>. 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

- 259 routines as well as standard routines and functions in MATLAB (MathWorks, MA).
- 260

261 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.

263 Data since 2008 were smoothed by nearest-neighbor averaging, yielding 3-minute time resolution. Data prior to 2008

264 were hourly and were not smoothed. Wind data were nearest-neighbor averaged after decomposing into north and east

- 265 components, followed by recalculation of u and  $\theta$ .
- 266

#### 267 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, CA, purchased 2015, Sigma Aldritch, St Louis, MO).

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- 276 Data are real time integrated and visualized in Google Earth on a portable computer (Spectre360, HP) using custom
- 277 software, written in MATLAB (MathWorks, MA), that is, described elsewhere (Leifer, Melton, Fischer, et al., 2018;
- 278 Leifer, Melton, Manish, & Leen, 2014; Leifer, Melton, Tratt, et al., 2018; Leifer et al., 2016). Real-time visualization

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facilitates adaptive surveys, wherein the survey route is modified based on real-time data to improve outcomes (Thompson et al., 2015) - in this case to facilitate plume tracking and to ensure transects were near orthogonal to the wind.

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Accurate, absolute winds are calculated from relative winds after accounting for vessel motion and filtering for nonphysical 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.

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291 Whole air samples were collected in evacuated 2-liter stainless steel canisters, which were filled gently over ~1 minute

292 from ~1 m above the sea surface. The filled canisters were analyzed in the Rowland/Blake laboratory at the University

293 of California, Irvine for carbon monoxide, CO, CH4, and C2-C7 organic compounds. Samples were analysed by a gas

- 294 chromatography multi-column/detector analytical system utilizing flame ionization detection.
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### 296 2.3 Seep plume emissions model

297 The plume inversion model is a three-step process (Leifer, Melton, Fischer, et al., 2018; Leifer, Melton, Tratt, et al., 298 2018; Leifer et al., 2016). Emissions from focused seep areas were derived from offshore data by first fitting Gaussian 299 function(s) to orthogonal transect C' data, termed the data model. C' is relative to C outside the plume, derived by linear interpolation across the plume transect. The data model is derived by error minimization using a least-squares 300 301 linear-regression analysis (Curve fitting toolbox, MathWorks, MA). Next, the Gaussian plume model (Eqn. S1; Supp. 302 Figs. S1 and S2) is fit to the data model. Transect data are collected close to orthogonal to the wind direction and are 303 projected in the wind direction onto an orthogonal plane. See Leifer, Melton, Tratt, et al. (2018) for a validation study 304 of the plume inversion model by comparison with remote sensing-derived emissions (which are largely insensitive to 305 transport). The study found in situ and remote-sensing derived emissions agreed within 11%.

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### 307 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 during an offshore survey that collected meteorology and *in situ* concentration data downwind of several active seep areas.

The definition of <u>an area source versus a point</u> 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.

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320	The area source was based on a Sept. 2005 sonar survey sonar return, $\omega$ , map (Fig. 1), see Leifer et al. (2010) for sonar	
321	survey details, data were re-analysed for this study. Simulations used sonar data gridded at a hybrid 22/56-m in a UTM	
322	coordinate system, with origin at WCS. Specifically, gaps in the 22-m map were filled from the 56-m map (Supp. Fig.	
323	S3). The probability distribution of $\omega$ was used to identify the noise level (Supp. Fig. S4) as in Leifer et al. (2010).	
324		
325	The model calculates a Gaussian plume for $E_{i,j}$ for grid cell <i>i</i> and <i>j</i> , for each grid cell with $\omega$ above noise. $C'_{sim}(x, y)_{i,j}$	
326	and for the observed wind-direction, $\theta$ , resolved wind, $u(\theta)$ , for a typical Santa Barbara channel boundary layer,	Deleted: are
327	BL=250 m. The initial $E_{i,j}$ was calculated by scaling such that the integrated sonar return, $\int \omega(x, y)$ , scales to	Deleted: (0)
328	$E_A=1.5 \times 10^5 \text{ m}^3 \text{ dy}^{-1}$ , i.e., $E_B$ from Hornafius et al. (1999). The Gaussian plume is calculated in a Cartesian coordinate	Deleted: in
329	system (Supp. Fig. S5A), rotated to $\theta$ , and the interpolated linearly to double the spatial resolution. Then, the rotated	Deleted: and Deleted: <i>E</i> _{<i>l</i>,j}
330	plume is regridded to UTM coordinates using the ffgrid.m function (Supp. Fig. S5B). Interpolation removes gaps in	Deleted: Elij
331	the regridded plume map. Then, the regridded plume is renormalized to ensure total mass is conserved before and	
332	after these operations. Rotated regridded plumes are translated to the seep field grid and added, yielding $C'_{sim}(x, y)$ ,	
333	which is the simulated seep field plume anomaly (Supp. Fig. S5C).	
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335	The model scans $\theta$ for the seep directions (110°< $\theta$ <330°) and calculates the simulated plume anomaly, $C'_{sim}(\theta)$ at	
336	WCS, which is compared with the observed WCS concentration, $C'_{obs}(\theta)$ . Hereafter, $C_{obs}$ and $C_{Sim}$ and their	
337	anomalies refer to values at WCS. $C'_{obs}(\theta)$ is defined.	Deleted: :
338	$C_{obs}(\theta) = C_{obs}(\theta) - \min(C_{obs}(\theta)) $ ⁽¹⁾	
339	with the minimum typically from the west in a direction with no known seepage. Specifically, $C'_{Obs}(\theta)$ was calculated	
340	by subtracting the minimum in the annualized observed $C'_{Obs}(t, \theta)$ each year, t, after applying a 7-year running	
341	average.	
342		
343	Emissions from suburban communities, light industry, and commercial centers enhance $C'_{Obs}(\theta)$ for the north <u>west</u> to	
344	<u>north</u> east ( $\sim 330, 30^{\circ}$ ), These terrestrial emissions were removed by fitting a Gaussian function to $C'_{Obs}(\theta)$ for	Deleted: 35
345	$330^{\circ} < \theta < 30^{\circ}$ with the residual yielding $C'_{obs}(\theta)$ . This only affected $C'_{obs}(\theta)$ for overlapping directions corresponding	Deleted: 70
346	to the fields' eastern edge.	Deleted: se
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348	Simulations were run at angular resolutions of 2°. Higher angular resolution produced small artifacts for the 22/56-m	(
349	sonar grid while the 11-m sonar grid was overly sparse due to the distance between sonar tracks (Supp. Fig. S3A).	
350		
351	The source is the $\rho$ map in units of decibels, whereas emissions are in units of moles m ⁻² s ⁻² . Given that the relationship	Deleted: ma
352	between $\omega$ and bubble density (emissions) is complex and non-linear (Leifer et al., 2017), there is poor agreement	Formatted:
353	between $C'_{sim}(\theta)$ and $C'_{obs}(\theta)$ . Thus, a correction function, $K(\theta)$ , is applied to emissions for each grid cell along each	Formatted: Deleted: , E
354	$\underline{\theta}_{k,i}(\theta)_{i,i}(\theta)_{i,j}$ and the model rerun. $K(\theta)$ is defined,	Deleted: , E
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370	$K(\theta) = \frac{C'_{obs}(\theta)}{C'_{sim}(\theta)} $ (2)		
371	Initially, $K(\theta) = 1$ , but in subsequent iterations, $K(\theta)$ is scaled as in Eqn. 2, Because $K(\theta)$ weights closer seeps more		<b>Deleted:</b> to adjust $E_A$ in cells along $\theta$
372	than more distant seeps, a distance-varying correction function, $K(r, \theta)$ , was calculated such that,		
373	$\int_{r=0}^{r=\infty} E_A(r,\theta) = \int_{r=0}^{r=\infty} K(r,\theta) E_A(r,\theta) dr $ (3)		
374	where r is distance from WCS. $E_A$ from simulations for different northwards shifts of WCS were fit with a polynomial		Deleted: Simulations
375	to derive the function form of $K(r)$ , Accounting for off-axis plume contributions requires several iterations to achieve		Deleted: that
376	Convergence, which was defined,	$\langle \rangle \rangle$	Deleted: shifted
377	$Convergence = \sum C'_{Sim}(\theta) \sum C'_{Obs}(\theta) / \sum C'_{Obs}(\theta)^{\perp} $ (4)		<b>Deleted:</b> northwards showed $E_A$ varied <b>Deleted:</b> nearly linearly with distance
378	Iterations continued to <i>Convergence</i> of 1% or better – typically 4 to 5 iterations. Simulations suggest wind veering,		
379	$\psi$ , was important, which was implemented by calculating $C'(\theta)$ and assigning it to $C'(\theta + \psi)$ .		
380			
381	3 Results		
382	3.1 Offshore in situ surveys		
383	An offshore COP seep field survey measured in situ C _{CH4} and u on 28 May 2016. Data were collected from the Santa		
384	Barbara harbor (~7.5 km east of the seep field, Fig. 2A; Supp. Fig. S6) to offshore Naples, several kilometers west of		
385	the seep field. Overall winds were easterly with an onshore component near Campus Point flowing onto UCSB and a		
386	broad (6-km wide) offshore flow west of COP that shifts to along coast near Naples (Fig. 2A, white arrows). Observed		
387	winds veered ~10° from east to the west sides of the seep field, roughly comparable to the shift in coastline orientation.		
388			
389	Plumes are apparent downwind of major seeps, with the largest plume associated with the Trilogy Seep (Fig 2B).		
390	Strong plumes also are evident downwind of the La Goleta Seep and Patch Seep. Notably, the Seep Tent Seep plume		
391	was very weak. The Seep Tent Seep was the dominant seep area in the COP seep field from its appearance in June		
392	1973 (Boles, Clark, Leifer, & Washburn, 2001) until recent years.		
393			
394	Additionally, the offshore survey identified focused plumes from beyond the extent of the seep field's 2005-sonar		
395	map. Specifically in the Goleta Bay, which has been noted (Jordan et al., 2020), and offshore Haskell and Sands		
396	Beaches, an area with abandoned oil wells, and off Naples Point (Fig. 2A, red arrow).		Deleted: areas
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398	Plume alkane C' were determined by the difference between two "background" air samples collected immediately		
399	outside the plume and three Trilogy Seep plume air samples. CH4 was 88.5% of THC, with ethane, propane, and		
400	butane at 3.1%, 4.2%, and 2.8%, respectively, with pentane, hexane, and heptane at 1.11, 0.13, and 0.04%, respectively		

401 (Table 1). Mean THC molecular weight is 19.6 g mole-1 based on a composition weighting. Branched alkanes were

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410	detected, with 2-methylpentane and 3-methylpentane comprising 0.21%, each, as well as simple aromatics, e.g.,		
411	benzene and toluene, with concentrations of $\frac{78}{28}$ and $\frac{18}{28}$ ppm, respectively.	 Deleted: 0.078	
412		 Deleted: 0.180	
413	The observed wet gas fraction, $\sum_{n=2}^{5} C_n / \sum_{n=1}^{5} C_n$ was 0.11 indicating a thermogenic origin, i.e., greater than 0.05	 Deleted: -	_
414	(Abrams, 2017) - and thus derived from marine organic materials. Although the olefins ethene and ethyne were		
415	detectable at 0.02% and 0.004%, respectively, butene was not detected. These olefins primarily derive from microbial		
416	processes (Abrams, 2017), thus, the ethane/ethyne ratio of 6200 strongly indicates a thermogenic source (Bernard et	 Deleted: also	
417	al., 2001). Plume atmospheric CO ₂ was elevated by 12 ppm, thus CO ₂ was 18% of Total Carbon, $TC$ , – defined $TC$ =		
418	THC+CO2+carbon monoxide, CO. CO was elevated minimally in the plume, just 2 ppb. Given that CO2 completely	 Deleted: as	
419	dissolves from bubbles well before reaching the sea surface (Clark et al., 2010), this demonstrates efficient vertical		
420	transport of dissolved seep gases to the sea surface.		
421			
422	Plumes for the Trilogy Seeps, La Goleta Seep, and Seep Tent Seep were inverse Gaussian plume modeled to derive		
423	emissions for each plume. For the Trilogy Seeps, the average $u$ across the plume was 5.9 m s ⁻¹ , insolation was full		
424	sun, and the source height was set at 25 m based on Trilogy's atmospheric plume being buoyant. Model surface		
425	concentrations for Trilogy B plume are shown in Fig 2B. The other two seeps are far less intense and used a 1-m	 Deleted: 2A	_
426	source height.		
427			
428	E for Trilogy A was 1.28 Gg CH4 yr ⁻¹ (5600 m ³ CH4 d ⁻¹ ₄ ), whereas Trilogy B and C contributed 0.06 and 0.07 Gg CH4	 Deleted: y	
429	yr ⁻¹ , respectively, for a total of 6200 CH4 m ³ d ¹ / _q . Note, plume origins and the sonar seep bubble plume locations do	 Deleted: y	
430	not precisely match because the sonar map is for near the seabed, and currents deflect the bubble surfacing location,		
431	up to ~40 m. La Goleta Seep released 4000 m ³ CH ₄ $d_{\psi}^{-1}$ and the Seep Tent Seep released 310 m ³ CH ₄ $d_{\psi}^{-1}$ with almost	 Deleted: y	_
432	no surface bubble expression. For comparison, Clark et al. (2010) used a flux buoy, which measures near surface	 Deleted: ay	
433	bubble fluxes, and found Trilogy Seep emissions of 5500 and 4200 m ³ THC d _v ⁻¹ and 930 m ³ THC d _v ⁻¹ for La Goleta	 Deleted: y	
434	Seep in 2005 and 5700 m3 THC dr1 for the Seep Tent Seep in 2002. During the cruise, surface bubble plumes were	 Deleted: ay	
435	not observed for the Seep Tent Seep, although its bubble plume had been a perennial and dominant feature since its	 Deleted: y	_
436	appearance. Note, Clark et al. (2010) reported THC in near sea surface bubbles was 91% CH4.		
437			

# 438 3.2 West Campus Station

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

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

Significant daily, seasonal, and interannual variations are apparent in the day-averaged *u* and *C* (Fig. 3). The calmest
season is late summer to fall, whereas spring is the windiest and most variable due to synoptic systems (Fig. 3A).
Winds have strengthened since a minimum in 1995-1996, moreso for the seep directions with stronger winds becoming
more frequent and moreso for summer than winter (Supp. Figs. S8, S9).

470 Trends in *C* reflect trends in both seep field emissions and ambient *C*. *C* is higher in fall and spring (Fig. 3B). Given 471 that stronger winds decrease *C* through dilution, this suggests the seasonal variation in *C* underestimates the seasonal 472 variation in emissions. Several studies have shown increased emissions under higher wave regimes (storminess), 473 reviewed in Leifer (2019) and proposed from wave pumping. Storms increase evasion from higher wave turbulence 474 and breaking-wave bubbles, which sparge dissolved CH4 and other trace gases down to the seabed in shallow (<100 475 m) waters (Shakhova, Semiletov, Salyuk, et al., 2010). Note, *u*, *θ*, and *C*' correlate with time of day. For example,

- 476 north generally reflects weak, offshore nocturnal winds with no seep contribution.
- 477

469

#### 478 3.2.2 Spatial heterogeneity

479 Calculating the angular-resolved average C,  $C_{ave}(\theta)$ , for the complete dataset with respect to  $\theta$  shows the highest C480 from the main seep field direction (155-250°, Fig. 4). For the seep directions,  $C_{ave}(\theta)$  was poorly fit by a single 481 Gaussian function but well fit ( $R^2 = 0.997$ ) by two Gaussian functions with peaks at 178° and 198° corresponding to 482 the Seep Tent and Trilogy Seeps' directions, respectively (Figs. 4A, 4B). Notably, the fit residual showed a linear 483 increasing trend,  $dC_{ave}(\theta)/d\theta$ , of 0.17 ppb degree⁻¹ from 180 to 210° (Supp. Fig. S9B) consistent with evasion from 484 a dissolved downcurrent plume that drifts west-northwest along the coast (Leifer, 2019).

485

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 beyond the dissolved plume's approximate shoreward edge. Figs. 4A and 4B show  $C_{Obs}(\theta)$  before removal of terrestrial emissions, which do not overlap in any significant manner with seep field

490 491 emissions.

- 492 There is a strong, focused peak in  $C_{max}(\theta)$  at  $\theta \sim 190^\circ$ , close to the Seep Tent Seep direction (Figs. 4E, 4F), which is 493 fairly isolated on the <u>offshore seep trend</u> (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 on the Red Mountain Fault trend.

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499 This strongly suggests that the seep field extends further to the west-northwest than current maps. These data cannot 500 be explained by dissolved plume outgassing, which would affect  $C_{ave}(\theta)$  but not  $C_{max}(\theta)$ . 501 502  $C(\theta)$  enhancements for non-seep directions (Fig. 4A, 4B) show a peak at ~35°, corresponding to the direction of a 503 commercial center amid suburban development. This could result from terrestrial seepage and natural gas pipeline 504 leakage and/or THC emissions from communities and traffic. 505 506 Neglecting the synoptic system, topographic forcing from the east-west Santa Ynez Mtn. range means that the 507 strongest winds are the prevailing westerlies (Fig. 4C, 4D). North winds (320-15°) largely are weak as are winds from 508 due south; however, the sea breeze strengthens winds rapidly away from due south.  $\theta$  peaks in the maximum winds 509 (1-minute sustained),  $u_{max}(\theta)$ , correspond to the west and east peaks in  $u_{ave}(\theta)$  with strengths to 16 m s⁻¹. Interestingly, 510 there also are strong north (0-30°) winds or downslope flow, termed sundowner winds, a highly localized and infrequent phenomenon. The overlap of  $u_{med}(\theta)$  and  $u_{ave}(\theta)$  shows winds largely are normally distributed. 511 512 513 The median C,  $C_{med}(\theta)$ , and average C,  $C_{ave}(\theta)$ , have similar shapes, albeit with lower values at all  $\theta$  (Fig. 4A), 514 indicating C is not normally distributed. This is shown in the wind direction-resolved wind speed probability distribution,  $\phi(\theta, u)$  (Fig. 5A), defined such that 515 516  $\int \phi(\theta, u) du = 1, \quad \int \phi(\theta, C) dC = 1_{-}$ (5) 517  $\phi(\theta, u)$  is very narrow (y-axis) for the northeast (~45°) where winds are largely weak and broad for the east-southeast (70-135°) and the prevailing westerlies (250-280°). The east-southeast distribution skews to the south (stronger winds 518 519 extend further from the south - offshore), whereas the prevailing westerly wind distribution skews to the northeast (as 520 does the coastline). 521 522 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  $\theta$  extending further to the west than the seep field extent (240°) and then decreasing more abruptly 523 than the decrease to the east. This asymmetry is expected given the seep field's asymmetric orientation relative to 524 525 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. 526 527

The  $\theta$ -resolved maximum  $C(\theta)$ ,  $C_{max}(\theta)$ , remains elevated through ~270°, far west of the  $C_{ave}(\theta)$  peak at ~200°.

#### 528 3.2.3 Seep field diurnal emissions cycle

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- C and u for the seep field direction,  $u_{seep}$ , and  $C_{seep}$ , respectively, follow diurnal patterns that are not the same as the
- 530 overall diurnal pattern due to the wind direction constraint and because  $C_{seep}$  depends on  $u_{seep}$ . The dependency arises
- because higher u dilutes emissions, decreasing C, but higher u also increases dissolved plume evasion and bubble-
- 532 mediated emissions from higher swell (after a delay for wave build-up). Diurnal winds in coastal regions feature a

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sift between weak nocturnal offshore winds that veer to onshore winds in the morning - the sea breeze circulation.

This was explored in time and direction segregated u and C and seep direction averaged  $u_{seep_t}$  and  $C_{seep}$  for 90-270° (Fig. 6). Data were segregated by  $\theta$  for pre- and post-2008 (when station improvements facilitated better wind characterization, particularly for night winds, which are seldom from the seep field direction, see Supp. Fig. S10 for 1991-2007).  $u(\theta,t)$  and  $C(\theta,t)$  were 2D Gaussian kernel smoothed with a 1-bin standard deviation (contours based on

- a 3-bin standard deviation) by the imgaussfilt.m algorithm (MATLAB, MathWorks, MA) after interpolating thecalibration data gap 24:00-01:00.
- 543

Early morning (01:00–03:00) *useep* 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),
- 547 with both increasing slightly through midday (~12:00).  $C_{seep}$  follows an afternoon trend of an overall decrease to a
- 548 minimum at ~20:00 before increasing into the late evening.
- 549

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
- 555 consistent with wave development time, transport time, and sparging of the downcurrent plume; however,
- 556 interpretation based on these patterns largely is speculative.
- 557

#### 558 3.3 Overall seep field emissions

#### 559 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 (Supp. Fig. S3). Simulations used moderate insolation to derive the turbulence parameters and stability class (Supp. Fig. S2), a 250-m *BL*, typical of 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,  $K(r, \theta)$ , was linear (Supp. Fig. S12).

- 568 Simulations could not reproduce observations in the Platform Holly direction ( $\theta$ =238°). Thus, a source was added for
- 569 the platform area, which improved simulation-observational agreement in this wind direction. Since significant seep
- 570 bubbles plumes generally are not observed in the platform's vicinity, these emissions could arise from incomplete
- 571 combustion during flaring.

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576 The model-derived,  $E_A$ , for 1990-2020 was  $83_{\pm}400$  m³ d⁻¹ (Fig. 7). Using a composition-weighted <u>THC</u> molecular 577 mass of 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 (Table 1). Given that CH4 is 73% of THC, non-methane hydrocarbon (NMHC: C2-C7) emissions are 9,600 578 579 m³ d¹ and gases emissions of 6.0 Gg yr¹. For reference, Santa Barbara County 2018 Reactive Organic Carbon (ROC) 580 emissions are listed at ~27 tons dr¹ (9.9 Gg yr⁻¹) (ourair.org/emissions-inventory, SBAPCD). For our analysis NMHC 581 and ROC are the same. The largest NMHC was propane with emissions of 3510 m³ d⁻¹, followed by ethane at 2590 582 m³ d⁻¹. The NMHC components of *THC* are conservative (do not react significantly) on the typical transport timescales 583 from the seep field to WCS (20-30 minutes). 584 585 Seabed emissions,  $E_B$ , are necessarily significantly greater than  $E_A$  as  $E_A$  misses the fraction of emissions that remain 586 in the water column,  $E_W$ , at least in the field's near downcurrent. There are two notes, the model  $E_A$  includes evasion 587 from the dissolved plume in the area covered by the seep field sonar map. Secondly, the model does not include  $E_A$ 588 from the dissolved fraction that evades beyond the seep field extent. For the seep field area and near downcurrent area, 589 Clark et al. (2000) estimated a 50:50 air/water partitioning based on a field study, implying  $\mathcal{E}_{B_{\varphi}} = 167,000 \text{ m}^3 \text{ d}_{\varphi}^{-1}$  for 590 <u>1990-20204</u>54 Gg yr¹). A comparison of  $E_A$  versus  $\omega$  showed a very steep increase with  $\omega$  for  $E_A = 1-10$  g s⁻¹ m⁻² with 591 rollover at @~0.022 (Supp. Fig. S13), which was approximately the noise level (Supp. Fig. S4). 592 593 Insights were provided by how the model partitioned emissions between different seep areas (Fig. 7). Particularly 594 notable is the model's treatment of the Trilogy Seep area - the second strongest seep area after the Seep Tent Seep 595 during the study period. The model re-assigned Trilogy Seep emissions to seepage to the west, representing Trilogy 596 Seep emissions as unrealistically weaker than other, smaller seeps, such as IV Super Seep. One likely contributor to 597 this re-assignment is wind veering (Supp. Fig. S14). Also suggesting wind veering is the model's assignment of strong 598 emissions to the field's eastern and western edges despite weak sonar returns. In a comparison of the Seep Tent Seep 599 and La Goleta Seep areas, the model emphasized the Seep Tent Seep whereas La Goleta Seep emissions were shifted

to inshore seepage. This re-partitioning was greatly reduced for a +10° wind veer, which also lessened the 600 601 strengthening of emissions from the field's western edge relative to sonar. Given the lack of field data between the

- 602 seep field and WCS on wind veering, further wind veering analysis was not conducted.
- 603

#### 604 3.3.2 Seep field sector emissions

605 To investigate sub-field scale emissions, the seep field was segregated into three sectors: inshore, offshore east, and

- 606 offshore west (Fig. 1). Based on integrating sonar return,  $\omega$ , inshore seepage contributes 40% of the field's  $\omega$  with the
- 607 offshore seep trend split between 9% for the west and 51% for the east. Supporting this comparison is the similarity
- 608 in the normalized sonar return probability distribution,  $\phi_n(\omega)$ , for the inshore seeps and offshore east seeps (Fig. 8). In
- 609 contrast,  $\phi_n(\omega)$  for offshore west seepage differed dramatically despite the similarity in geology along the anticline
- 610 underlying the offshore seep trend (Leifer et al., 2010). This likely results in part from the interaction between

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629 distribution,  $\phi_n(E_A)$ , for the inshore and offshore seeps are similar over most of the range (except the weakest,  $E_A < 0.02$ 630 g s⁻¹), significant differences are evident between offshore east and west seepage. Offshore east seepage is more 631 dispersed and favors weaker seepage compared to offshore west seepage, 632 633 The weakest seepage ( $\omega < 0.02$ ) contributes negligibly to overall sonar return and had no notable inshore-offshore  $\phi_n(\omega)$ 634 difference (Fig. 8). The largest difference is between the strongest seepage ( $\omega > 0.5$ ) for the inshore and offshore seeps. 635 Specifically, there is a strong peak at  $\omega$ -0.45 and nothing stronger for the inshore seeps, whereas offshore  $\phi_n(\omega)$ 636 continued to  $\omega$ -0.7. The  $E_A$  probability distribution,  $\phi_n(E_A)$ , for the strongest inshore seepage was similar to  $\phi_n(E_A)$  for 637 strong offshore seepage. However, this masked a significant east-west offshore seepage difference. Specifically,

migration and production from Platform Holly. Although the normalized atmospheric emissions probability

stong on shore seepage. However, this masked a significant east west on shore seepage directeder. 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.

641 The <u>similarities of these</u> distributions suggest that <u>the</u> controlling geological structures (fractures, fault damage zones, 642 chimneys, <u>etc.</u>) are the same for inshore seepage and offshore east seepage, with the primary difference for the 643 strongest seepage in these two sectors which are of similar strength – the inshore Trilogy Seeps provide focused 644 emissions, whereas the offshore east La Goleta Seeps are comparatively dispersed and far oilier.

646 Although,  $\omega$  is not  $E_A$ ,  $E_A$  followed the 40:60 partition in  $\omega$  between inshore and offshore seepage. Interestingly, the 647  $E_A$  partitioning between the offshore east and offshore west differed significantly from sonar partitioning with 21% of  $E_A$  from offshore west and 38% from offshore east. This greatly accentuated the  $E_A$  Seep Tent Seep area. In part, this 648 649 arises from a diurnal cycle bias - WCS observes the offshore west seeps for afternoon/evening westerly winds, which 650 are stronger, whereas WCS observes the offshore east seeps when winds are weaker, earlier in the day (Fig. 6B). 651 Winds increase bubble emissions from wave hydrostatic pumping and dissolved gas evasion. Also potentially 652 contributing is saturation of  $\omega$  at very high bubble-density bubble plumes, such as the Seep Tent Seep and Trilogy 653 Seep (Leifer et al., 2017). Saturation would imply an under-estimate of  $\omega$  for the strongest seep area emissions which 654 are for the west offshore seepage, altering the west:east  $\omega$  ratio (9%:51%).

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#### 656 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,  $\zeta$ . Sensitivity study details are presented in Supp. Sec. S7.4. **Deleted:** and compared to  $\phi_n(\omega)$ 

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668 The contribution to uncertainty from  $\delta\theta$ , C',  $\psi$ , and spatial offsets within the seep trends were minimal – just a few 669 percent or less. Moderate uncertainty was identified for BL and  $\zeta$ . For example, for BL ranging from 150 to 350 m, 670 mean  $E_A$  uncertainty was 6%. Although u has strong sensitivity, combined with BL it does not as u counters BL – 671 lower u corresponds to higher BL. There still is uncertainty, though in the value of BL, which is not measured. 672 Assessing uncertainty in  $\zeta$  was more challenging as there is no verification data on variability in the  $E_A$  partitioning 673 between the inshore and offshore seep trends. The mean  $E_A$  uncertainty for -50% < $\zeta$ <50% is 11.5% from a polynomial 674 fit. Still, the consistency in seepage location between sonar surveys spanning decades (Leifer, 2019) suggests only 675 modest changes in  $\zeta$  over the multi-decade time period of model averaging. Total uncertainty was taken as 15% based 676 on the sum of uncertainty in *BL* and  $\zeta$ , each averaged to the nearest 5%.

#### 678 3.4 Ellwood Field emissions

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679  $C(\theta)$  increases to the northeast with a peak at 290-320° corresponding to the direction towards abandoned wells off 680 Haskell Beach (Fig. 10). Emissions from this area – either from natural seepage or leaking wells – were noted in the 681 offshore survey data near Haskell Beach (Fig. 2A). Additionally,  $C_{max}(\theta)$  shows a 22-ppm peak in in this direction 682 well above  $C_{ave}(\theta)$  (Fig. 4F), consistent with transient releases from natural seep and/or abandoned well emissions. 683

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.

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. Deleted: :

#### 698 4 Discussion

### 699 4.1 Atmospheric seep field observations

#### 700 4.1.1. Air quality station

701 A range of approaches are available to evaluate marine seepage CH4 emissions: in situ approaches including direct 702 capture (Leifer, 2015; Washburn, Johnson, Gotschalk, & Egland, 2001), fluid flow measurements (Leifer & Boles, 703 2005), video (Leifer, 2015), and remote sensing approaches that include active acoustics, i.e., sonar (Hornafius et al., 704 1999), dissolved in situ (Marinaro et al., 2006), and passive acoustics (Wiggins et al., 2015). Remote sensing is the 705 best approach for long-term monitoring to capture shifts in emissions between vents. To date, only sonar remote 706 sensing has provided quantitative seep plume (seabed) emissions. Notably, sonar ranges are up to a few hundred 707 meters, far less than the size scales of many seep fields, whereas high power, demands typically require a cabled 708 observatory for long-term observations.

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## 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

712 example, sonar surveys generally are scheduled during summer when seas are calmer and winds more predictable and

# 713 when seepage is weakest (Fig. 3); however, not during storms when emissions likely are enhanced.

714 715 The approach derived atmospheric trace gas emissions for a dispersed area source constrained by sonar seepage maps 716 from long-term air quality and meteorology data. This approach can be extended to terrestrial seepage if the source 717 can be constrained spatially (due to geology); although nearby anthropogenic sources may complicate emissions 718 assessments. Other terrestrial sources such as landfills, O&G production fields, or industrial sites – if spatially 719 constrained – could be addressed by this approach, particularly if isolated from other confounding sources. The use 720 of cavity enhanced absorption spectrometers that can speciate gases like CH₄ and C₂H₆ that_could enable

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#### 725 4.1.2 *In situ* atmospheric surface surveys

could include in situ aqueous chemical sensors, current measurements.

726 Atmospheric emissions were assessed for three seep areas - zone of focused seepage - by an atmospheric in situ survey

discrimination some confounding sources as well as better characterization of emissions. Although an onshore station

can address nearshore seepage, further offshore seepage could be addressed by a moored station. Moored stations also

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

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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.

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#### 747 4.2 Seep field emissions

#### 748 4.2.1 Total emissions

749 To date, only two estimates of COP seep field seabed emissions,  $E_{B}$ , have been published. Hornafius et al. (1999) 750 estimated  $E_B=1.5 \times 10^5 \text{ m}^3 \text{ dy}^{-1}$  (64 Gg yr⁻¹) based on sonar surveys covering 18 km² from Nov. 1994 – Sep. 1996, 751 collected during the summer to late fall seasons. This value excluded Seep Tent collection. A 4.1 km² sonar survey in 752 Aug.-Sep. 2016 estimated E_B=24,000 m³ dy⁻¹ (Padilla, Loranger, Kinnaman, Valentine, & Weber, 2019), significantly 753 lower, which in part arises from field subsampling, but also could arise from long-term changes; however, neither 754 study addressed temporal variability. The sonar surveys occurred in summer and fall when seepage activity is at a 755 minimum, whereas winter and early spring feature much higher activity associated with large transient events and 756 storms (Bradley et al., 2010).

758 Hornafius et al. (1999) used an engineered bubble plume to calibrate emissions, an approach also used in Leifer et al. 759 (2017). Due to technology limitations at the time, the strongest seepage was clipped or saturated, i.e., underestimated, 760 and the survey did not cover shallow seepage. Thus, the Hornafius et al. (1999) emissions estimate is a lower limit for 761 summer/fall emissions. The Padilla et al. (2019) survey was calibrated by an inverted seep flux buoy suspended at 23 762 m. This differs significantly from the seep flux buoy measurement approach reported in, Washburn et al. (2001), which 763 are collected in surface drift mode. Surface drift mode ensures a horizontal orientation for the buoy and an absence of lateral velocity difference between the capture device and currents - either of which decreases capture efficiency from 764 765 100%, biasing derived emissions low. Further, the Padilla et al. (2019) survey was calibrated 1 month after the sonar 766 surveys, whereas the 1995 engineered plume calibration by Hornafius et al. (1999) was contemporaneous. The 767 Hornafius et al. (1999) approach accounts in part for dissolution between the seabed and survey depth window - it 768 used air rather than CH4, which dissolves slower than CH4 Dissolution losses for CH4 between the seabed and the 769 depth window can be addressed by a numerical bubble model (Leifer et al., 2017).

The Gaussian plume model-derived  $E_A$  was  $8.3 \times 10^4$  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. Deleted: s

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(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.

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## 782 4.2.2 Methane and non-methane hydrocarbon emissions

783 Analysis of atmospheric samples provided a picture of the complexity of atmospheric emissions that arises from the 784 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 785 786 bubble transport (Leifer & Clark, 2002). Thus, bubble-mediated transport enhances larger alkanes relative to smaller 787 alkanes leaving more of the smaller alkanes in the water column. For strong seeps, bubble plumes are associated with 788 strong upwelling flows (Leifer et al., 2009), which transport dissolved gases to the sea surface where they outgas. 789 Additionally, oil (as droplets and bubble coatings) enhances alkane transport due to slower dissolution and diffusion 790 of larger alkanes through oil. 791

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.

798 COP seep field seabed emissions are orders of magnitude greater than typically reported for other seep areas, e.g., summary Römer et al. (2017) where emissions for 12 different seep areas including sites in the North Sea, Pacific 799 northwest, Gulf of Mexico, etc., were 2-480 tons yr-1, multiple orders of magnitude less than COP seep field seabed 800 801 emissions. Römer et al. (2017) used a bubble model for Dogger Bank seepage in the North Sea to estimate emissions 802 for observed atmospheric CH₄ plumes. The model estimated direct atmospheric bubble-mediated emissions of 21.7 803 ton yr⁻¹, 20% of seabed emissions. For the Tommelieten Seeps (in 70-m water) Schneider von Deimling et al. (2011) estimated 4% of the 0.024 Gg CH₄ yr⁻¹ seabed emissions, i.e., ~1 Mg CH₄ yr⁻¹ reached the atmosphere by bubble-804 mediated transfer. Schneider von Deimling et al. (2011) used a bubble model based on an assumed bubble size and 805 806 neglected diffusive flux. These diffusive fluxes include bubble dissolution into the wave mixed laver in the local area. A few studies have directly measured atmospheric fluxes by an air-sea gas transfer model. For example, Schmale, 807 808 Beaubien, Rehder, Greinert, and Lonmbardi (2010) found seep air fluxes of 0.96-2.32 nmol m⁻² s⁻¹, much higher than 809 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⁻ 810 811 ² s⁻¹ for the shallow South China Sea based on an air-sea gas transfer model. If we disperse COP seep field atmospheric 812 emissions of  $1.15 \times 10^9$  M yr⁻¹ over the ~6.3 km² of 25x25 m² bins with emissions, we find 5.7  $\mu$ M m⁻² s⁻¹, three orders 813 of magnitude greater.

815 Recent estimates of total global geo-CH₄ sources from a bottom-up approach are 45 Tg yr⁻¹ with submarine seepage

816 contributing 7 Tg yr⁻¹ (Etiope & Schwietzke, 2019), implying COP seep field contributes  $\sim 0.27\%$  of global submarine 817 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 estimate, if 818 819 accurate, would imply the COP seep field contributes an astounding 1% of global seep emissions (submarine and 820 aerial) and is difficult to reconcile with the COP seep field and other top seepage estimates. For example, CH4 atmospheric emissions for the Lusi hydrothermal system of 0.1 Tg yr⁻¹ (Mazzini et al., 2021), a hotspot in the Laptev 821 822 Sea of 0.9 Tg yr⁻¹ into shallow seas (Shakhova, Semiletov, Leifer, et al., 2010), and for the East Siberian Arctic Sea 823 using eddy covariance of 3.0 Tg yr⁻¹(Thornton et al., 2020). Thus, COP seep field emissions either play a significant 824 role in global seep emissions or indicate that geo-gas emissions are less tightly constrained.

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COP seep field  $C_2H_6$  emissions were 1.27 Gg  $C_2H_6$  yr⁻¹. For reference, this is 11% of the 11.4 Gg  $C_2H_6$  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 contribute 2-4 Tg  $C_2H_6$  yr⁻¹ (Etiope & Ciccioli, 2009), and from ice cores, 2.2-3.5 Tg yr⁻¹ {Nicewonger, 2016 #3816}.

- This suggests the COP seep field contributes 0.03-0.06% of global seep emissions.
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833 Seep THC was 4.2% propane, implying emissions of 2.5 Gg C₃H₈ yr⁻¹. Global propane emissions are 10.5 Tg yr⁻¹ 834 (Pozzer et al., 2010), with 1-2 Tg yr⁻¹ estimated for seeps (Etiope & Ciccioli, 2009). This suggests the COP seep field 835 contributes 0.05-0.1% of the global seep budget. Oceans are estimated to contribute 0.35 Tg C₃H₈ yr⁻¹ (Pozzer et al., 836 2010), less than geological seepage contribution.

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Based on an evaluation of the COP seep field emissions with respect to global seep ethane and propane emissions,
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 geoCH₄ emissions, e.g., 1.6 Tg yr⁻¹ (1.15%) (Hmiel et al., 2020).

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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, <u>with</u> combined C₂-C₅ emissions <u>of</u> 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 (D. 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.

854 Both benzene and toluene were detected with estimated emissions of 8300 and 2300 kg yr⁻¹, respectively. These

855 emissions likely are underestimates, potentially significantly, due to neglecting the oil slick evaporation contribution.

856 Both gases are of significant health concerns, as are alkanes like pentane and hexane.

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# 858 4.3 Downcurrent emissions

The seep field concentration,  $C'(\theta)$ , anomaly was centered at  $\theta$ -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**).

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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, 866 867 extending as far as ~280° from WCS due to the coastline shift from northwest to west around Haskell Beach (Fig. 10),  $\sim$ 30° beyond the seep field's sonar mapped western edge (Fig. 1). As prevailing winds are westerlies (paralleling the 868 869 coastal mountains), downcurrent plume evasions decrease with distance due to dispersion and also as surface waters 870 become depleted by evasion. Evasion increases non-linearly with u, particularly for winds that include wave breaking 871 (Nightingale et al., 2000); however, higher winds also dilute emissions. Note, there are no mapped seeps in this area. 872 873 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.

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## 879 4.4 Focused seep area emissions

880 Trilogy Seep area emissions were estimated at 6,200 m3 CH4 dy-1 in May 2016. For comparison, Clark et al. (2010) 881 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 882 surface bubble fluxes in Sept. 2005. Note, the plume inversion approach also includes outgassing of near surface 883 waters that have enhanced  $C_{CH4}$  from plume dissolution, which the flux buoy approach does not include. Although 884 Clark et al. (2010) found surface bubbles had undetectable CO2, the atmospheric plume's CO2 to CH4 concentration 885 ratio was comparable to the seabed bubble concentration ratio. This demonstrates significant upwelling flow transport of seabed water to the sea surface where dissolved gases evade near where the bubble plume surfaces. This near-plume 886 887 evasion contributes to the atmospheric plume. Note, these emissions neglect downcurrent emissions. A 50:50 Deleted: 2

889	atmosphere:ocean partitioning suggests 2016 Trilogy Seep emissions were ~40% lower than in 2005 - a difference	
890	within the difference between the two 2005 Trilogy Seep measurements Clark et al. (2010).	
891		
892	In contrast, agreement was very poor for the Seep Tent Seep, for which Clark et al. (2010) mapped emissions of 5700	
893	m3 day-1 (5000 m3 CH4 day-1) in Nov. 2002 whereas this study found 310 m3 CH4 dy-1. This discrepancy was readily	
894	apparent with almost no visible surface bubble expression in May 2016, whereas the Seep Tent Seep has been a	
895	perennial feature since its appearance. The absence of more than a few scattered bubbles at the sea surface (the boil	
896	in 2000 was driven by a 1-2 m s ⁻¹ upwelling - Leifer, Clark, and Chen (2000) - indicates that most emissions are from	
897	evasion. A buoyancy plume associated with the rising oil (thick oil slicks surface above the Seep Tents) as well as	
898	CH4 dissolved in the oil likely are transporting the observed, focused CH4 emissions.	
899		
900	This is remarkable given that the seep field's geofluid migration "center" in recent decades has been the Seep Tent	
901	Seep (Bradley et al., 2010), which was the largest seep in the field in 2010 (Clark et al., 2010). The Seep Tent Seep	
902	consists of emissions not captured by the Seep Tents - two large (33-m square) steel capture tents on the seafloor. For	
903	reference, the Seep Tents captured ~16,800 m ³ gas dy ⁻¹ in the early 2000s (Boles et al., 2001). Bradley et al. (2010)	
904	found in WCS data that when overall seep field emissions decreased to a minimum in 1995, they were focused on the	
905	Seep Tent Seep direction. Note, the Seep Tent Seep was observed first in 1970 as a boil visible from 1.6-km distant.	
906	The Seep Tent seep was tented in Sept. 1982 (Boles et al., 2001).	
907		
908	Underlying these observations are several factors. First, the Seep Tent Seep is modern - since 1978 - as it was not	
909	mapped in a 1953 seep survey (Leifer, 2019). At the time it was first reported as a sea boil visible over a kilometer	
910	distant (Boles et al., 2001). Since installation, overall Seep Tent production has diminished (Boles et al., 2001) by a	
911	factor of 3 from 1984 to 1995. Some fraction of this trend could have resulted from the expansion of active seepage	
912	beyond the seep tents. Perhaps more significantly, the Seep Tent Seep lies over one of the Platform Holly wells (Leifer	
913	et al., 2010; Fig. 3C), creating the potential of linkage between well production (including stimulation) and Seep Tent	

- 914 production and thus Seep Tent seepage (the uncaptured portion).
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#### 916 **4.6 Diurnal trend and bias**

917 The diurnal wind patterns typical of the coastal Pacific marine environment are weak offshore (northerly) night winds that shift to from the east in the morning and then swing to from the south. In afternoon they strengthen and shift to 918 prevailing westerlies, continuing to late in the evening (Bradley et al., 2010). Note, WCS seep emissions require winds 919 to "probe or scan" across the seep field, and thus miss the strong afternoon prevailing winds when emissions are 920 921 expected to be higher. This is because higher wind speeds increase sea-air gas emissions of dissolved near-surface 922 gases (Nightingale et al., 2000) and increase emissions from higher hydrostatic pressure fluctuation driven by wave height (Leifer & Boles, 2005). Given that prevailing winds are westerlies, higher afternoon emissions will generally 923 924 (but not always) drift eastwards, missing WCS.

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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.

934 The diurnal trend for C from the seep direction followed the diurnal wind cycle, increasing by  $\sim 20$  ppb and peaking

 $\sim$ 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*₄.

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.

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#### 946 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  $CH_{4_e}$  C₂H_{6_e} and C₃H₈. 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 speeds based on the wind speed probability distribution with respect to wind direction,  $\phi(u_e \theta)$ , e.g., Fig. 5.

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. Deleted: and Deleted:

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#### 965 5 Conclusions

966 In this study, data from an onshore air quality station located downwind of a large marine seep field was analysed to 967 derive the three-decade-averaged seep field emissions using an inversion model. The modeled emissions were similar 968 to reported emissions; however, this was coincidental given that prior reported emissions were during a period of field 969 quiescence. Highlighting the significance of the COP seep field, ethane and propane emissions suggest the COP seep 970 field contributes 0.04% and 0.12% of the global seep budget, respectively. As a result, COP seep field emissions of 971 19 Gg CH4 yr⁻¹ are consistent with global geo-gas budgets of 45 Tg yr⁻¹, but inconsistent with significantly lower 972 emissions estimated from ice core isotopic data. Additionally, the approach could be adapted to air quality station data 973 for other sources including terrestrial seeps, production fields, etc., if the sources are spatially constrained and isolated 974 from confounding sources. 975 Data availability. All data needed to evaluate the conclusions in the paper are present in the paper and/or the 976 977 Supplementary Materials and/or were submitted to the Mendeley Data Repository, see Leifer, Ira (2020), 978 "Seep_Air_Data", Mendeley Data, V1, http://dx.doi.org/10.17632/znzhxkftm8.1 979 980 Supplement. The supplement contains additional supporting figures and details to complement the manuscript and 981 an interactive map file as a Google Earth archive of the offshore survey data that are presented in Fig. 2. 982 983 Author Contributions. IL Developed and conducted the study, analysed data, and wrote the manuscript. CM analysed 984 data and edited the manuscript. DB analysed air sample data and edited the manuscript. 985 986 Competing interests. The authors declare that they have no conflict of interest. 987 988 Acknowledgements. We would like to gratefully acknowledge the SBCAPCD for providing data from their ongoing monitoring program and the contribution of Marc Moritsch and Joel S. Cordes in particular for help with these data, 989 and Doug Wilson for the processed sonar data. We recognize the skill and participation of vessel captains Jeff Wright 990 and Tony Vultaggio and editorial review by Charlotte Marston, Bubbleology Research International. 991 992 993 Financial Support. This work was supported by Plains All American Pipeline and the Bubbleology Research 994 International, Internal Research and Development (IRAD) fund.

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Gas	THC Fraction	TC* Fraction	Emissions	Emission
	(%)	(%)	$(m^3 dy^{-1})$	(Mg yr ⁻¹ )
CH ₄	88.5	72.8	73,900	19,300
$C_2H_6$	3.10	2.55	2,590	1270
$C_3H_8$	4.18	3.44	3,510	2520
$C_4H_{10}$	2.76	<u>2.27</u>	2,300	2180
C5H12	1.11	0.92	930	1090
$C_6H_{14}$	0.133	0.11	110	150
C6H6	7.8x10 ⁻⁵	6.4x10 ⁻⁵	7.1	8.3
C7H16	0.036	0.030	<u>29.7</u>	<u>45.8</u>
C7H8	1.8x10 ⁻⁵	1.5x10 ⁻⁵	1.8	2.3
NMHC ^{**}	11.5	9.48	9,640	7410
THC ^{***}	85	82.3	83,400	26,600
$CO_2$		17.7	21,600	15200
CO		0.003	3780	2660

1393 Table 1 At . del at . :4: а . .. •

<u>*TC is total carbon and is THC+CO+CO2</u> <u>*NMHC is non methane hydrocarbons and is C₂-C₇</u> <u>***THC is total hydrocarbon and is C₁-C₇</u>

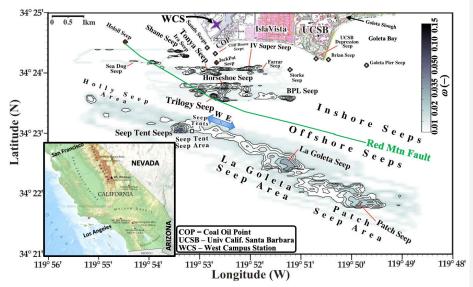
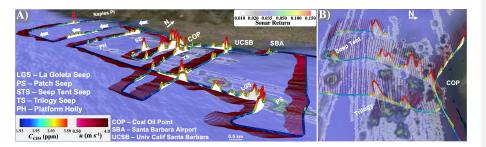


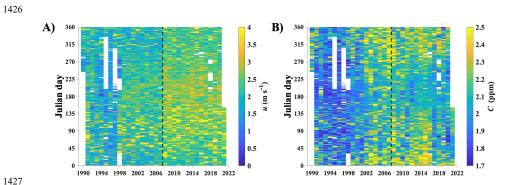


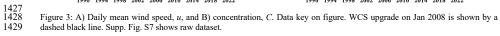
Figure 1: Sonar return,  $\omega$ , map after Leifer et al. (2010). Purple star marks West Campus Station (WCS). Seep names are informal (Table S3), font size corresponds to strength. E-W arrow segregates east and west offshore seepage. Data keys on panels. Inset shows S. California, red dot marks COP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

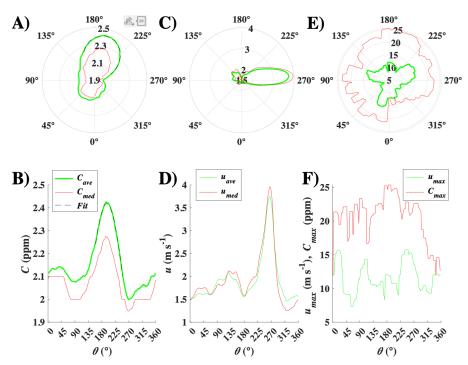
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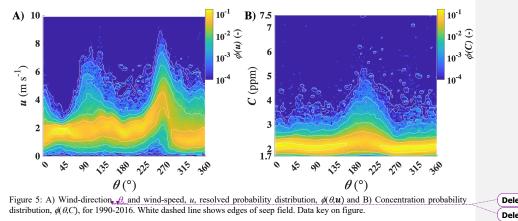
1421 1422 1423 1424 Figure 2: A) Methane,  $C_{CH4}$ , and wind, u, data for 28 May 2016. White arrows show canyon offshore flow. Red arrows show unmapped scepage to the west of the COP seep field. B)  $C_{CH4}$  and u showing Gaussian plume model for Trilogy Seep. Sonar return map shown on sea surface. Data key and seep name key on panel. Displayed in the  $\bigcirc$  GoogleEarth environment. See Supp. Fig. S6 for overhead view.

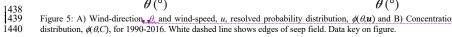




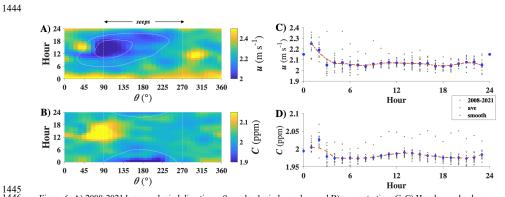


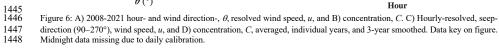
- 1433 1434 1435 1436 Figure 4: A, B) Concentration, C, versus wind-direction,  $\theta$ , 1990-2021 for average,  $C_{ave}(\theta)$ , and median,  $C_{med}(\theta)$ , and fit to  $C_{ave}(\theta)$  for  $155 \le \theta \le 250^\circ$ . Data key on panel B. C, D) Wind speed, u, average,  $u_{ave}(\theta)$ , 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 COP seen field seep field.

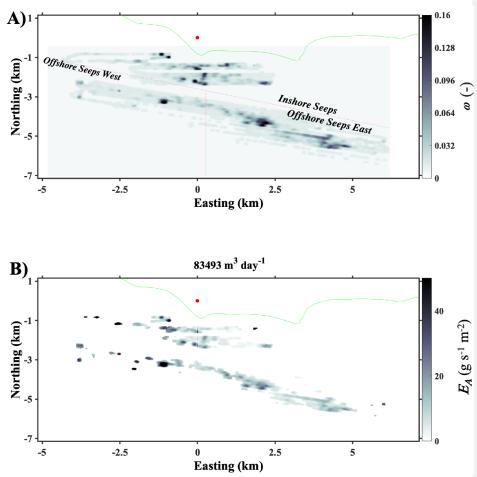




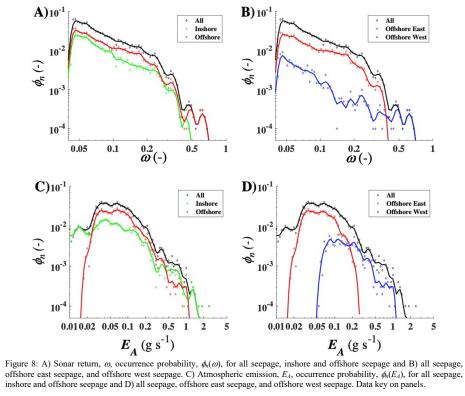
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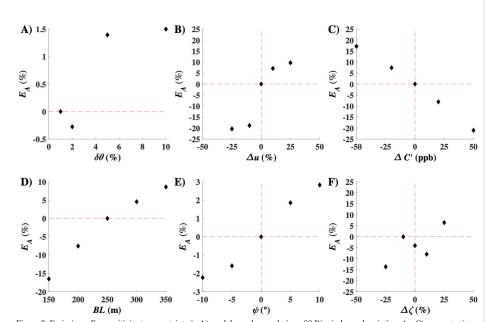




1451 1452 Figure 7: A) Sonar return,  $\omega$ , gridded at 22-m resolution. B) Atmospheric emissions,  $E_A$ . West campus station (red dot) is at coordinate system origin. Green line is coast line.



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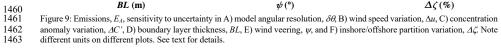




Figure 10: Map of the Goleta Plains oil and gas fields, wells, and the Coal Oil Point (COP) seep field. Grey hatch shows 1995 field extent, green outlines the 1940 field extent is from Leifer (2019). Field locations from Olson (1983). Well data from CDOGGR (2018). Faults from Minor et al. (2009). Seep names are informal. Data keys on panels. Shown in the © Google Earth environment.