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Trends of solar ultraviolet irradiance at Barrow, Alaska, and the effect of measurement uncertainties on trend detection

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Abstract

Spectral ultraviolet (UV) irradiance has been observed near Barrow, Alaska (71° N, 157° W) between 1991 and 2011 with an SUV-100 spectroradiometer. The instrument was historically part of the US. National Science Foundation's UV Monitoring Network

- and is now a component of NSF's Arctic Observing Network. From these measurements, trends in monthly average irradiance and their uncertainties were calculated. The analysis focuses on two quantities, the UV Index (which is affected by atmospheric ozone concentrations) and irradiance at 345 nm (which is virtually insensitive to ozone). Uncertainties of trend estimates depend on variations in the data due to (1) natural vari-
- ¹⁰ ability, (2) systematic and random errors of the measurements, and (3) uncertainties caused by gaps in the time series. Using radiative transfer model calculations, systematic errors of the measurements were detected and corrected. Different correction schemes were tested to quantify the sensitivity of the trend estimates on the treatment of systematic errors. Depending on the correction method, estimates of decadal trends
- changed between 1.5 % and 2.9 %. Uncertainties in the trend estimates caused by error sources (2) and (3) were set into relation with the overall uncertainty of the trend determinations. Results show that these error sources are only relevant for February, March, and April when natural variability is low due to high surface albedo. This method of addressing measurement uncertainties in time series analysis is also applicable to
- $_{20}$ other geophysical parameters. Trend estimates varied between $-14\,\%$ and $+5\,\%$ per decade and were significant (95.45% confidence level) only for the month of October. Depending on the correction method, October trends varied between $-11.4\,\%$ and $-13.7\,\%$ for irradiance at 345 nm and between $-11.7\,\%$ and $-14.1\,\%$ for the UV Index. These large trends are consistent with trends in short-wave (0.3–3.0 μ m) solar
- ²⁵ irradiance measured with pyranometers at NOAA's Barrow Observatory and can be explained by a change in snow cover over the observation period: analysis of pyranometer data indicates that the first day of fall when albedo becomes larger than 0.6 after snow fall, and remains above 0.6 for the rest of the winter, has advanced with a statistically significant trend of 13.6 ± 9.7 days per decade.



1 Introduction

Solar ultraviolet (UV) radiation reaching the Earth's surface affects humans, terrestrial and aquatic ecosystems, and the chemical composition of the troposphere (UNEP, 2010; ACIA, 2005). UV radiation can cause sunburn, cataracts, and skin cancer in humans. Health benefits of UV radiation are principally derived from vitamin D production in the skin, which supports bone health and may decrease the risk of several internal cancers. Arctic inhabitants may experience high UV levels in the summer caused by reflections off of snow but the absence of UV radiation during winter months may result in Vitamin D deficiency (Holick, 2007). Changes in UV radiation (either up or down) can therefore affect health. UV effects on terrestrial ecosystems are often complex and indirect. For example, litter-decomposing fungi are sensitive to UV-B radiation (Gehrke et al., 1995; Moody et al., 1999) and this may affect the recycling of plant

material. Plant exposure to UV-B radiation can change the composition of leaf tissue, which significantly affects the palatability and digestibility of food consumed by herbi-

¹⁵ vores, including reindeer and caribou (Gwynn-Jones, 1999). Furthermore, changes in UV irradiance can reduce the productivity of marine ecosystems with effects on phytoplankton and species higher on the food chain (Hessen, 2001).

UV irradiance at the surface is affected by the solar elevation, stratospheric and tropospheric ozone concentrations, clouds, Rayleigh scattering on air molecules, surface

²⁰ albedo (e.g., snow cover, sea ice), aerosols, absorption by trace gases, and the Sun-Earth distance (WMO, 2007). Long-term changes in any of these parameters can lead to trends in UV radiation. For the Arctic, changes in ozone, albedo (snow and ice), and cloud cover are of particular importance.

Predictions of future ozone concentrations in the Arctic stratosphere have a large ²⁵ uncertainty because of their sensitivity to temperature. For example, heterogeneous reactions on the surfaces of polar stratospheric clouds (PSCs) are responsible for large losses in ozone observed in cold Arctic winters (WMO, 2011). There is a robust linear correlation between the ozone loss and the volume of vortex air with temperatures below -78°C, the temperature threshold below which PSCs start to form (Rex et al.,



2006). Because of the large uncertainty of current Chemistry-Climate Models to predict stratospheric temperatures (for example, many models tend not to capture the low temperatures observed in the Arctic lower stratosphere – WMO, 2011), there is in turn a large uncertainty in the evolution of Arctic spring-time ozone concentrations and surface UV intensities.

The extent of sea ice in the Arctic is currently decreasing rapidly due to climate change (Serreze et al., 2007). Models suggest that ice cover in summer will disappear within the next few decades (Comisco et al., 2008). Reduced surface albedo because of decreases in snow and ice cover will increase the fraction of solar energy absorbed by the Earth's surface. Organisms that were once living below snow and ice will be

- ¹⁰ by the Earth's surface. Organisms that were once living below snow and ice will be exposed to increased doses of UV, but organisms living above the surface will receive lower doses of UV due to the reduced reflectivity (UNEP, 2010). Climate models predict increased cloudiness and precipitation at high latitudes (Meehl et al., 2007), which would generally lead to decreases in UV radiation.
- ¹⁵ Surface UV irradiance can be derived from satellite measurements (e.g., Krotkov et al., 1998, 2001). These data sets have a large uncertainty for high latitudes because of the difficulty in distinguishing between snow and clouds from space (Tanskanen et al., 2007). When snow is misinterpreted as a cloud, the result is reduced below the value calculated for clear sky rather than increased. Particularly at coastal Arctic accurate the product the produc
- ²⁰ locations, ground-based measurements are more accurate than satellite observations, also implying that trend estimates are less prone to error.

In principle, trends in UV radiation can either be inferred from direct measurements (either from ground or space) or reconstructed based on proxy data such as total ozone and sun shine duration (e.g., Lindfors et al., 2003). Trends of summertime daily ery-

themal dose at 12 mid-latitude sites estimated with both methods for the period of 1980–2003 ranged between 2% and 9% per decade (WMO, 2011). Weatherhead et al. (1998) calculated that at least 15 years of measurement are necessary to detect a trend of 5% per decade in solar UV measurements with confidence. Considering that only very few UV data records are longer than 20 years (e.g., Krzycin et al., 2011), the



opportunity to derive meaningful trend estimates from direct measurements has arisen only recently.

The first attempt to estimate trends in solar UV irradiance at Barrow was presented by Gurney (1998). The data analysis was based on UV measurements of the years
1991 to 1995. According to this study, spectral irradiance at 305 nm increased by 3 to 10% per year for all daylit months except June. These trends are much larger than those presented in this paper. It is likely that the large trends calculated by Gurney (1998) were partly a result of the abnormally small total ozone columns in 1992 and 1993, which were caused by high stratospheric aerosol concentrations following the eruption of Mt. Pinatubo in 1991 (WMO, 2003).

The UV radiation climate at Barrow and its influencing factors have been quantified by Bernhard et al. (2007). The authors also performed a preliminary trend analysis of UV-B, UV-A, and visible irradiance based on data of the years 1991–2005. Best estimates of trends varied between -23% and +11% per decade depending on data products and month but were generally not statistically significant. This paper extends this earlier analysis to the period 1991–2011.

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Maintaining a 20+ year data record at a low uncertainty level is a demanding task. Challenges include instrument failures, periods with degraded instrument performance, gaps in operational support, and drifts of calibration standards. All these factors affect trend estimates. Particular attention is given in the following analysis to the effect of

²⁰ trend estimates. Particular attention is given in the following analysis to the effect of measurement uncertainties on the detectability of trends. The method can also be applied to other environmental data sets.

The earlier analysis by Bernhard et al. (2007) showed that trend estimates for noontime and daily dose data are almost identical. Daily doses can only be accurately cal-

²⁵ culated when measurements throughout the day are available. This constraint reduces the number of days available for calculating a monthly average. The trend analysis presented in this paper is based on noontime measurements (22:00 UTC) because the number of noontime observations is generally larger than the number of days when a daily dose is available.



The earlier analysis also demonstrated that trends in UV radiation (and their sign) show large differences from month to month and that annualized trends are difficult to interpret. Trend estimates discussed in this paper are therefore based on monthly average data.

- The uncertainty of a trend estimate depends on natural variability, the measurement uncertainty, and data gaps. All three sources of variability will be discussed in this paper. In general, it is not possible to unambiguously attribute an "outlier" in a measured time series to one of three mechanisms. For example, consider the following thought experiment involving three idealized experiments:
- 10 **1.** An ideal instrument with zero uncertainty is measuring temperature that is normally distributed about a constant value with a known standard deviation of x %.
 - 2. A real instrument with a relative standard error of x % is measuring a constant temperature.
 - 3. The real instrument of (2) measures the variable temperature of (1).
- Experiments (1), (2), and (3) are repeated many times and a trend is estimated for every "run." For Experiments (1) and (2), the mean trend will be zero and the standard deviations (or uncertainties) calculated from the many individual trend estimates will be identical. Variability in measurements – caused either by real fluctuations in temperature (Experiment (1)) or by uncertainties in the instrument (Experiment (2)) –
- are therefore indistinguishable. For Experiment (3), the uncertainty of the trend estimate will be $\sqrt{2}$ times of that of Experiment (1) or (2). In most circumstances, a priori knowledge of the instrument's measurement uncertainty does not allow to reduce the uncertainty of the trend estimate for the "real-world" Experiment (3). However, the trend uncertainty of Experiment (3) can be set in perspective to the hypothetical trend
- ²⁵ uncertainty of Experiment (2). For example, comparing the two uncertainties can help decide whether the uncertainty of the measurement apparatus may seriously affect the ability to detect statistically significant trends in temperature. Comparing the trend



uncertainties derived from real-world measurements to that caused by measurement uncertainties will be an important part of this paper.

2 Data set

The trend analysis presented here is based on measurements of global (sun and sky) spectral irradiance performed between January 1991 and April 2011 with a high-resolution SUV-100 spectroradiometer. The instrument was historically part of the US National Science Foundation's UV Monitoring Network and is now a component of NSF's Arctic Observing Network. The instrument is installed into the roof of the Uk-peagvik lñupiat Corporation building (71°19′29″ N, 156°40′45″ W, 8 m a.s.l.), which is
located approximately 5.5 km northeast of the village of Barrow, Alaska, approximately 300 m inland from the Chukchi Sea, and 10 km away from Point Barrow, the northernmost point of Alaska. The land south and east of the system is flat tundra, which is snow-covered roughly between October and June (Stone et al., 2002). Annual cycles in sea ice and snow cover cause a large difference of surface albedo between summer

- and winter. General weather conditions and the radiation climate have been characterized by Maykut and Church (1973). Dutton at al. (2004) found that the annual average frequency of cloud occurrence has increased at Barrow from about 76% of day time in 1976 to about 82% in 2001. Over the same period, effective cloud transmission decreased significantly from 0.64 to 0.61. These trends in cloud characteristics led to a downward trend in annual average short wave (0.2, 2.0 µm) solar irradiance at the
- $_{20}\,$ a downward trend in annual average short-wave (0.3–3.0 $\mu m)$ solar irradiance at the surface (Dutton et al., 2006).

The data set analyzed here has been referred to as "Version 2 data of the National Science Foundation's Ultraviolet Spectral Irradiance Monitoring Network." It has been described in detail by Bernhard et al. (2007). Measurements have been corrected for

the instrument's cosine response error. Version 2 data are complemented with results of a radiative transfer model (Mayer and Kylling, 2005), which takes into account solar zenith angle, total ozone, vertical profiles of temperature and ozone, surface pressure,



 NO_2 absorption, and effective surface albedo. The model implementation has also been described by Bernhard et al. (2007).

Trends were estimated from two quantities retrieved from the Version 2 spectra: spectral irradiance integrated over the wavelength band of 342.5 to 347.5 nm (hereinafter called "irradiance at 345 nm" or E345), and the UV Index, which is a measure of the effectiveness of UV radiation to cause sunburn in human skin (WMO, 1998; WHO, 2002). The UV Index was calculated by weighting the measured spectra (provided in units of (μ W cm⁻² nm⁻¹)) with the action spectrum for erythema (McKinlay and Diffey, 1987) and multiplying the result by 0.4 cm² μ W⁻¹. Trends were estimated on measurenents performed at 22:00 UTC, which is the measurement closest in time to local solar noon at Barrow. Measurements at this time are available for the entire period. Monthly averages were calculated from the daily noontime measurements and trend estimates are based on these monthly means, denoted $\overline{E}(y_i, m)$, where y_i is year and m is month (m = 1, 2, ..., 12).

15 3 Trend analysis

A linear regression model was used to estimate monthly trends and their uncertainty. Using this model, the measured monthly mean irradiance $\bar{E}(y_i, m)$ can be written as:

$$\bar{E}(y_i,m) = a(m) + b(m)y_i + \varepsilon(y_i,m),$$

where a(m) and b(m) are the regression constant and slope, respectively, and $\varepsilon(y_i, m)$ are the residuals. For simplicity, the argument *m* is omitted in the following. The values of the parameters *a* and *b* were calculated by minimizing the merit function $\sum_{i=1}^{n} (\bar{E}(y_i) - a - by_i)^2$, where *n* is the number of years considered. The residuals $\varepsilon(y_i)$ can be written as:

$$\varepsilon(y_i) = \varepsilon_{\mathsf{S}}(y_i) + \varepsilon_{\mathsf{U}}(y_i) + \varepsilon_{\mathsf{G}}(y_i),$$

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(1)

(2)

where $\varepsilon_{\rm S}(y_i)$ is the component resulting from natural causes such as year-to-year variations in the atmospheric transmission. $\varepsilon_{\rm U}(y_i)$ is resulting from the measurement uncertainty, expressed as a "standard uncertainty" (ISO, 1993) and denoted $u_{\rm U}(\bar{E}(y_i))$. $\varepsilon_{\rm G}(y_i)$ results from the uncertainty in the calculation of a monthly average if there are gaps in the data series.

The uncertainty of trend estimates is determined by the variance of the measured values $\bar{E}(y_i)$ and can generally not be reduced by knowing $u_U(\bar{E}(y_i))$. For example, let us assume that there is an exact linear relationship between the year and the actual monthly mean irradiance (that is, $\varepsilon_S(y_i) = 0$ for all y_i). The values of the two coefficients *a* and *b* are determined by linear regression from the measurements $\bar{E}(y_i)$, which are affected by the uncertainty $u_U(\bar{E}(y_i))$. For this hypothetical case, the uncertainties of the estimated values for *a* and *b* do generally not depend on whether or not the uncertainty $u_U(\bar{E}(y_i))$ is known a priori. Exceptions from this rule may apply under certain circumstances. For example, Hicke at al. (2008) used a parametric bootstrap technique approach is used here: I calculate the uncertainty of *b* (denoted u(b)), from the variability of the measured data; use $u_U(\bar{E}(y_i))$ to calculate a theoretical uncertainty of *b* (denoted $u_{(b)}$), which results from the hypothetical case that $\varepsilon_S(y_i)$ and $\varepsilon_G(y_i)$ are

zero for all y_i ; and finally compare $u_U(b)$ with u(b). Before a regression is attempted it is imperative to remove k

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²⁰ Before a regression is attempted, it is imperative to remove known systematic errors from the daily irradiances that are used to calculate the monthly mean irradiances and systematic errors caused by data gaps. The determination of these errors is also subject to uncertainty. Different methods to correct systematic errors in the measurements have been explored and are discussed in Sect. 3.1. The uncertainty $u_U(\bar{E}(y_i))$

is quantified in Sect. 3.2. The correction for data gaps and the standard uncertainty of this correction, denoted $u_{G}(\bar{E}(y_{i}))$, are described in Sect. 3.3. The different correction schemes result in several different data sets for $\bar{E}(y_{i})$. Regressions were performed on all data sets to test the sensitivity of the derived trends on the choice of the correction method.



Data of all months were tested for autocorrelation using the Durbin-Watson Test (Draper and Smith, 1998). No autocorrelation was indicated. "d"-values of the Durbin-Watson statistics for the different months range between 1.34 and 2.50. The median is 2.0, i.e. the ideal value for a data set that is not autocorrelated. Autocorrelation was therefore not considered when calculating the uncertainties of regression slopes.

The uncertainty of the regression slope u(b) is calculated based on the propagationof-error principle for uncorrelated variables (e.g., Press et al., 1986):

$$u(b) = \left[\sum_{i=1}^{n} \left[u(\bar{E}(y_i))\right]^2 \left(\frac{\partial b}{\partial \bar{E}(y_i)}\right)^2\right]^{0.5}.$$

The uncertainty $u(\overline{E}(y_i))$ is assumed to be independent of *i* and calculated as the sample standard deviation of the residuals for two degrees of freedom:

$$u(\bar{E}(y_i)) \equiv u(\bar{E}) = \sqrt{(n-2)^{-1} \sum_{i=1}^{n} [\varepsilon(y_i)]^2}.$$

To calculate the hypothetical uncertainty $u_U(b)$ of a regression slope resulting from measurement uncertainties, the term $u(\bar{E}(y_i))$ in Eq. (3) has to be replaced by $u_U(\bar{E}(y_i))$:

$${}_{15} \quad u_{\mathsf{U}}(b) = \left[\sum_{i=1}^{n} \left[u_{\mathsf{U}}(\bar{E}(y_i))\right]^2 \left(\frac{\partial b}{\partial \bar{E}(y_i)}\right)^2\right]^{0.5}.$$
(5)

Similarly, the standard uncertainty of $\bar{E}(y_i)$ related to data gaps is denoted $u_G(\bar{E}(y_i))$, and the hypothetical uncertainty of the regression slope resulting from $u_G(\bar{E}(y_i))$ is denoted $u_G(b)$. To calculate $u_G(b)$, the term $u(\bar{E}(y_i))$ in Eq. (3) has to be replaced by $u_G(\bar{E}(y_i))$:

$${}_{20} \quad u_{\rm G}(b) = \left[\sum_{i=1}^{n} \left[u_{\rm G}(\bar{E}(y_i))\right]^2 \left(\frac{\partial b}{\partial \bar{E}(y_i)}\right)^2\right]^{0.5}.$$
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(3)

(4)

(6)

Trends are calculated at a confidence level of 95.45%. (I chose a level of 95.45% rather than the level of 95.0% that is often used because 95.45% is the percentage of values within $\pm 2\sigma$ of a normal distribution.)

Percentual decadal trends T were calculated relative to the year 2000: $T[\%] = 1000 \times$

- $_{5}$ $b/(a+2000 \times b)$. The uncertainty of the trend, denoted u(T), is $u(T)[\%] = 1000 \times u(b) \times t(n-2,0.9545/2)/(a+2000 \times b)$, where t(n-2,0.9545/2) is the value of Student's t-distribution for *n* samples and a confidence level of 95.45%. Trend uncertainties associated with the uncertainties $u_{U}(b)$ and $u_{G}(b)$ were calculated similarly and are denoted $u_{U}(T)$ and $u_{G}(T)$.
- ¹⁰ Using Student's t-test for determining the significance of trends is only appropriate if residuals $\varepsilon(y_i)$ are normally distributed. Normality was tested with the Anderson-Darling test (Stevens, 1974). The test's A²-statistic was smaller than 0.7 for all months, both for E345 and UV Index data. This suggests that the null-hypothesis that data are normally distributed cannot be rejected (p < 0.05). The t-test is therefore suitable.
- ¹⁵ The propagation-of-error principle suggests that

 $(u(T))^2 = (u_{\rm S}(T))^2 + (u_{\rm U}(T))^2 + (u_{\rm G}(T))^2,$

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where $u_{\rm S}(T)$ is the uncertainty in the trend estimate resulting from natural variability alone. This quantity indicates the uncertainty of the trend that could be expected from measurements of an ideal instrument. Real trends must be outside the range of $T \pm u_{\rm S}(T)$ such that they can be detected with an ideal instrument with confidence. The uncertainty of $u_{\rm S}(T)$ is therefore a principle limit for trend detection given by natural variability.

The ratios of $R_U(T) \equiv (u_U(T))^2 / (u(T))^2$ and $R_G(T) \equiv (u_G(T))^2 / (u(T))^2$ were also calculated to judge the contribution of the measurement and gap variances to the overall trend variance $(u(T))^2$.



(7)

3.1 Detection and correction of systematic errors

Comparing measured UV spectra with spectra calculated with a radiative transfer model is a useful method to assess the quality of the measurements and to detect significant errors in measurements that may change over time (Bernhard et al., 2004,

- 5 2008). The method is most accurate for periods when the state of the atmosphere and the surface albedo are well defined. At Barrow, these conditions are usually met during clear-sky periods in the summer when aerosol concentrations are small and the surface is free of snow. For example, the surface albedo of tundra during the snow-free months of June through September can be assumed to be constant at about 5 % in the
- ¹⁰ UV (Blumthaler and Ambach, 1988). Assuming that there are no trends in unknown atmospheric absorbers, the ratio of the measured and modeled spectral irradiance during cloud-free period in the summer should be similar for every year, and variations of this ratio over time may indicate a drift in the measurements. (There may still be a systematic bias between measurement and model due to invariant systematic errors
- ¹⁵ in either the measurements or the calculations. However, such a bias would not affect long-term trend estimates).

Model spectra that are part of the Version 2 data set use effective surface albedo (Lenoble et al., 2004) and total ozone column as input parameters. Both parameters are calculated from the measurements (Bernhard et al., 2004). Version 2 data are

- therefore not independent from model results. The ratio of measurement and model is therefore a less-useful quality control tool for months when the ground is covered by snow or at wavelengths that are affected by ozone absorption. Fortunately many systematic errors, such as the drift in the output of calibration lamps, have only a modest wavelength dependence. Temporal variations in the ratio of measurement and model at 345 nm, where ozone absorption is negligible, are therefore also good indicators for
 - instrument drifts below 340 nm where ozone absorption is important.

The analysis is based on spectra measured at solar zenith angles (SZA) smaller than 80°. The limitation has two reasons. First, measurements and model calculations are more challenging at large SZAs and the value of using model calculations as a



quality control tool is reduced at large SZAs. Second, trend estimates presented in this paper are based on noontime measurements when the SZA smaller than 80° (with the exception of data measured before 26-February or after 15-October when SZA is larger than 80°). Some Version 2 spectra are flagged for inferior quality. These ⁵ measurements were also excluded from the data analysis.

In theory, the consistency of measurements over time could also be tested by comparing clear-sky irradiances measured at the same SZA during different years. In practice, this method is problematic as clear-sky conditions occur in different periods every year.

- ¹⁰ The method of using model spectra to quality-control measurements is similar to that presented by Bernhard et al. (2007). Only a summary of the implementation is provided here. First, clear-sky periods are determined based on temporal variability using the method by Bernhard et al. (2008). Second, spectra measured during these periods for SZA < 80° are ratioed to the associated "clear-sky" model spectra of the Version
- ¹⁵ 2 data set. In the second step of the analysis, the medians of these "ratio spectra" are calculated on a wavelength-by-wavelength basis from all ratio spectra within preset sample intervals. These sample intervals include entire years (e.g., 1991, 1992, ..., 2011), months (e.g., January 1991, February 1991, ..., April 2011), and low-albedo summer periods (e.g., 1 July–30 September 1991, 1 July–30 September 1992, ...,
- ²⁰ 1 July–30 September 2010). Lastly, these "median-ratio-spectra" are averaged over the wavelength range 340–350 nm. The resulting "q-ratios" are denoted $q_{annual}(y)$, $q_{monthly}(y,m)$, and $q_{summer}(y)$, respectively, where y is year (y = 1991, 1992, ..., 2011) and m is month (m = 1, 2, ..., 12). The three quantities are shown in Fig. 1. The q-ratios vary between 0.79 and 1.08 with the majority of values being between 0.94 and 1.02.
- The medians of the q-ratios are 0.970, 0.973, and 0.963, respectively, and are denoted \bar{q}_{annual} , $\bar{q}_{monthly}$, and \bar{q}_{summer} .

The expanded uncertainty of measurements of Version 2 UV data is 6.0% (Sect. 3.2). To contrast this value with the variation of the q-ratios, Fig. 1 also includes lines at 1±0.060 (yellow lines) and $\bar{q}_{monthly} \pm 0.060$ (orange lines). Most q-ratios fall



within the two limits but some are outside, requiring further investigation. I consulted Network Operations Reports (available at: http://uv.biospherical.com), measurements of the GUV filter radiometer that is collocated with the SUV-100 spectroradiometer (Bernhard et al., 2003), and records of the SUV-100's internal temperature for clues

- for the observed outliers. A detailed assessment of this information is available at: http://uv.biospherical.com/Version2/PaperBAR/Increased_uncertainty.pdf. Periods with increased uncertainty that should not be used for trend analysis are also identified in this document, and data of these periods were excluded from the trend analysis of this paper.
- For some months, $q_{\text{monthly}}(y,m)$ is based on less than 10 spectra, indicating long periods of persistent cloudiness. The remaining spectra are likely also somewhat contaminated by clouds and the usefulness of these spectra for QC purposes is questionable. The following discussion focuses on months with at least 10 spectra (solid blue symbols in Fig. 1).
- ¹⁵ The low values of $q_{\text{monthly}}(1991,7)$ and $q_{\text{monthly}}(1991,8)$ can be explained by the effect of aerosols from the Mt. Pinatubo eruption of May 1991, which may not have been correctly addressed by the model. The low value of $q_{\text{monthly}}(1994,4)$ can likely be attributed to overheating of the instrument.

 $q_{\text{monthly}}(2003,7)$ is based on data from the 9th, 20th, and 21th of July only. All other days were cloudy. Solar heating on those days increased the temperature of the instrument's monochromator beyond the capacity of the instrument's thermoelectric cooler. The instrument has a negative temperature coefficient: increased temperatures result in decreased instrument responsivity. Solar measurements are therefore biased low, leading to $q_{\text{monthly}}(2003,7) = 0.91$, which is 6.5% below \bar{q}_{monthly} . It is not reasonable to correct monthly data upward by this amount, as only three clear-sky days are affected by the temperature effect.

Data from 14 May–31 May 2005 are not available due to a defective instrument shutter; data of May 2005 were not used for trend analysis. The calibration for July 2005 is uncertain and the low value of $q_{\text{monthly}}(2005,7)$ suggest that measurements



are indeed biased low.

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Calibration scans could not be performed between mid-July and November 2007. The calibration of the instrument for this period therefore has an increased uncertainty. The low values of $q_{monthly}(2007,8)$ and $q_{monthly}(2007,9)$ are likely caused by a systematic error in the measurements. The instrument's collector was not cleaned in August 2008, likely leading to reduced responsivity and a low value of $q_{monthly}(2008,8)$. The instrument was not calibrated in the fall of 2009; data of this period were also not used.

The study of ancillary material could not explain outliers in $q_{\text{monthly}}(y,m)$ for May 1992, March 1993, August 1999, June 2002, and June 2010, indicating that either the measurements of these months are affected by a source of uncertainty that has not been addressed in the uncertainty budget or that the model calculation is biased (e.g., the albedo used in the model might be too small or too large).

 $q_{summer}(y)$ tends to be larger between 1992 and 2001 compared to 2002–2010. This step change may be explained by the modification of the instrument's cosine collector at the beginning of 2001 (Bernhard et al., 2007). The differences in the angular response of the old and new collector were address by the cosine-error correction, but may not have been removed completely. Results presented in Sect. 4 indicate that the step change has a noticeable effect on trend estimates.

Data were corrected by scaling measurements using the results of the measurementto-model comparison. For example, in order to utilize the $q_{annual}(y)$ values for the correction, all measurements performed in year y were multiplied with the correction factor $C_{annual}(y) \equiv \bar{q}_{annual}/q_{annual}(y)$. Likewise, the values of $q_{summer}(y)$ and $q_{monthly}(y,m)$ were used by multiplying the noontime measurements with $C_{summer}(y) \equiv \bar{q}_{summer}/q_{summer}(y)$ and $C_{monthly}(y,m) \equiv \bar{q}_{monthly}/q_{monthly}(y,m)$, respectively. If q-ratios for a certain month were not available (for example because the number of clear-sky days was not sufficient), the correction factor was set to 1. For reasons explained above, $C_{summer}(y)$ should be the most accurate correction factor, however, the factor may not be the most appropriate for years when the systematic errors changed between spring and summer.



3.2 Measurement uncertainty

The uncertainty of spectra measured with the SUV-100 spectroradiometer at Barrow has been discussed by Bernhard et al. (2007). For a SZA of 45°, the expanded relative uncertainty (coverage factor k = 2, equal to a confidence interval of 95.45%) of

⁵ erythemal irradiance varies between 5.8% and 6.2% and depends only little on year. The corresponding range for SZA = 80° is 5.8%–8.8%. Uncertainties are dominated by "type B" uncertainties (ISO, 1993), which do not change with averaging. Based on these considerations, the relative standard uncertainty $u_U(\bar{E}_i)/\bar{E}_i$ of the monthly means was set to 3% for all years and months.

3.3 Correction for gaps in data set

The monthly mean irradiances \bar{E}_i were calculated by averaging all available noontime measurements of a given month. If a measurement of a single day was missing, but measurements of the previous and subsequent days were available, the value for the missing day was calculated as average of the measurements of the two adjacent days.

- No uncertainty was attributed to this procedure. Months with more than 10 missing days were not used for the trend analysis. There will be a bias in the monthly average if periods with missing days are not equally distributed. For example, solar radiation tends to increase during months in the spring because the noontime SZA decreases. If measurements are missing at the beginning of a month, the monthly average will
- ²⁰ be biased high. To correct for this effect, for every day of the year, the average was calculated from measurements of all years, resulting in a climatological "mean cycle" of UV radiation at Barrow. In a second step, two averages were calculated from this mean cycle, one that uses all values of a given month, denoted A(m), and one (denoted $A^*(y,m)$) that is based only on those days of this mean cycle that are available in
- the data set for year *y* and month *m* that needs adjustment. Monthly averages are corrected by multiplication with the "gap correction factor" $C_{\rm G}(y,m) \equiv A(m)/A^*(y,m)$. For example, suppose the monthly mean of March 2001 is to be corrected. In this month,



measurements of the first five days are missing. First, the average of all available noontime measurements of this month is calculated, resulting in the "uncorrected monthly average" for March 2001, denoted $\bar{E}_u(2001,3)$. Second, the mean cycle is averaged over March (days 60 to 90), resulting in A(3). Third, the mean cycle is averaged over days 65 to 90 (i.e., the first five days of March are omitted), resulting in $A^*(2001,3)$. Lastly, the "corrected monthly average" for March 2001 ($\bar{E}_c(2001,3)$), is determined:

$$\bar{E}_{c}(2001,3) = \bar{E}_{u}(2001,3) \frac{A(3)}{A^{*}(2001,3)} = \bar{E}_{u}(2001,3)C_{G}(2001,3).$$
 (8)

The correction mostly takes into account climatological variations in SZA and ozone (e.g., the mean Dobson-Brewer circulation), which repeat every year. Other factors, such as year-to-year variations in total ozone are not taken into account. The relative standard uncertainty associated with this correction, $u_{G,rel}(y,m)$, is calculated as

$$u_{\rm G,rel}(y,m) = \frac{1}{\sqrt{3}} |C_{\rm G}(y,m) - 1|.$$
(9)

The calculation of $u_{\rm G}(y,m)$ is based on the assumption that the probability distribution function of $\bar{E}_{\rm c}(y,m)$ is rectangular; that is, the probability that the true value of the monthly average, $\bar{E}_{\rm t}(y,m)$, lies within the interval $[\bar{E}_{\rm c}(y,m)/C_{\rm G}(y,m),\bar{E}_{\rm c}(y,m) \times C_{\rm G}(y,m))$ is constant and is zero outside this interval (ISO, 1993). This is a conservative estimate of the uncertainty.

4 Results

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Trends in monthly mean noontime UV radiation $\overline{E}(y_i)$ were determined for E345 and the UV Index. Four different data sets were considered for each quantity based on different corrections for systematic errors. These are: no correction (Method 1); and corrections using the correction factors C_{annual} (Method 2), C_{summer} (Method 3), and C_{monthly} (Method 4). The "gap correction" (Sect. 3.3) was applied to all data sets.

Figure 2 shows the result of the time series analysis for the monthly mean noontime irradiance at 345 nm. Data have been corrected with Method 4. The figure is divided into nine panels for the months of February through October. Each panel indicates the monthly means $\overline{E}(y_i)$, the measurement uncertainty $u_U(\overline{E}(y_i))$, the trend sestimate T (expressed in percent per decade), the uncertainty of the trend u(T), the correlation coefficient (R^2), the uncertainty of the trend that can be explained with the measurement uncertainty $u_U(T)$, and the trend uncertainty caused by gaps $u_G(T)$. The hyperbolic confidence bands associated with each of the three uncertainties are also indicated and were calculated according to Draper and Smith (1998).

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- The following can be concluded from the results presented in Fig. 2:
 - Trend estimates range between -14% (October) and +3% (May) per decade.
 - The trend uncertainty ranges between 3% (March and April) and 13% (September). The uncertainty is much smaller in the spring than in the fall. This can be explained with the smaller natural variability in the spring when the albedo is large and variability due to cloud attenuation is greatly reduced (Ricchiazzi et al., 1995; Nichol et al., 2003).
 - Trends are not significant at the 95.45 % confidence level for any month, except for October, where the trend is -14 ± 12 % per decade.
 - $u_U(T)$ is 3% for all months. Because of the small natural variability in spring, $u_U(T)$ has a much larger effect on the detectability of trends during the first part of the year compared to fall, when the natural variability outweighs the effect of the measurement uncertainty.
 - $u_{\rm G}(T)$ ranges between 0 % and 3 %, and is an important factor for the detectability of trends only for the month of February.
- ²⁵ Trend calculations were repeated using the data sets corrected with Methods 1, 2 and 3, to quantify the sensitivity of the trend estimates on the treatment of systematic errors. Results of the trend estimates *T* and the associated uncertainty u(T) are shown in



Fig. 3. The difference between maximum and minimum trend estimate for each month varies between 1.6% (February) and 2.9% (June). Results of Methods 1, 2, and 4 are generally very similar, but trends determined by Method 3 are 1.7% larger on average. This is caused by the step-change of $C_{summer}(y)$, with lower values for the years 1992–2001 than for the 2002–2010 period. The difference between the maximum and minimum uncertainty of the trend estimates (error bars of Fig. 3) varies between 0.3% (July and September) and 1.8% (June).

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Figure 4 is similar to Fig. 2, but shows the result of the time series analysis for the UV Index. The following can be concluded from the results presented in Fig. 4:

- Trend estimates range between -14 % (October) and +5 % (August) per decade.
 - The trend uncertainty ranges between 4 % (February) and 13 % (September), is smaller in the spring than in the fall, but tends to be larger than the trend uncertainty calculated for irradiance at 345 nm.
 - Only the trend for October is significant; it is -14 ± 13 % per decade.
- $u_U(T)$ is 3 % for all months. $u_G(T)$ ranges between 0 % and 4 %, and is an important factor for the detectability of trends only for the month of February.

Trend calculations were again repeated using the data sets corrected with Methods 1, 2 and 3 and results are shown in Fig. 5. The difference between maximum and minimum trend estimate for each month varies between 1.5% (February) and 2.9% (June). Trends estimated with Method 3 are 1.7% larger on average. The difference between the maximum and minimum uncertainty of the trend estimates varies between 0.2% (September) and 1.1% (April and June). These results are very similar to those obtained for the trend in irradiance at 345 nm.

Table 1 provides a compilation of all relevant parameters of the trend analysis. ²⁵ Monthly trends were generally calculated based on data of 17 to 19 years. $u_U(T)$ ranges between 2.5% and 3.0%. $u_G(T)$ exceeds 3% in February and varies between 0.1 and 1.5% for the other months. For irradiance at 345 nm, the ratio of the variance



caused by the measurement uncertainty to the observed variance, $R_{II}(T)$, is larger than 50% in February, March, and April; varies between 20% and 33% in May (depending on the correction method); is between 10% and 20% for May-August; and below 10 % in September and October. For the UV Index, ratios of $R_{\rm H}(T)$ are generally 5 smaller than those calculated for irradiance at 345 nm because monthly variations of the UV Index are also influenced by ozone variations. The contribution of $u_{11}(T)$ to the overall variability is therefore smaller. The ratio of the variance caused by gaps to the observed variance, $R_{\rm G}(T)$, exceeds 60 % in February and is below 10 % for the other months, indicating again that the contribution of gaps in the time series is unimportant for trend detection. For February, the sum of $(u_{II}(T))^2$ and $(u_{G}(T))^2$ is larger than 10 $(u(T))^2$. $(u_S(T))^2$, calculated with Eq. (7), becomes therefore negative, which is an impossible result. I believe that the uncertainty attributed to data gaps was unrealistically large for February. For all months but February, the difference between u(T) and $u_{s}(T)$ varies between 0.3% (October) and 1.4% (April), suggesting again that reducing the measurement uncertainty would be mostly beneficial for measurements in the spring, 15 but would have little effect on the ability to detect trends for low-albedo months.

5 Discussion

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The most striking result of the trend analysis is the large and statistically significant downward trend in UV irradiance of about 14% per decade for October. I hypothesize that this trend is caused by changes in snow cover and the resulting changes in surface albedo.

Surface albedo increases downwelling irradiance because a fraction of photons reflected upward by the surface are scattered downward by either air molecules (clear sky case) or cloud droplets. Model calculations for a SZA of 80° (noontime SZA at

Barrow on 15 October) indicate that an increase in albedo from 5 % to 80 % (a typical value for Barrow during winter – Bernhard et al., 2007) will increase spectral irradiance at 345 nm by 38 % for clear skies and by a factor of 2.97 for overcast skies (cloud optical state).



depth of 50). For clear skies, high surface albedo has a larger effect on UV than visible irradiance due to the wavelength dependence of Rayleigh scattering. However, this wavelength-dependence is largely reduced for overcast situations. According to my model calculations, a change in albedo from 5% to 80% will increase global spectral
 ⁵ irradiance at 600 nm by a factor of 2.89, which is only slightly below the factor of 2.97 calculated for 345 nm.

To determine whether changes in snow cover have occurred during the last 20 years, measurements of upwelling and downwelling short-wave irradiance measured by pyranometers at the Barrow Observatory of NOAA's Global Monitoring Division were analyzed. The facility is located 2 km east of the UV spectroradiometer. Data are part of the Baseline Surface Radiation Network (BSRN) and are available for the years 1992– 2009. Albedo was calculated by dividing upwelling with downwelling irradiance. Snow cover usually leads to a distinct change in albedo from below 0.2 (no snow) to above 0.7 (snow). I determined the first day of fall when albedo becomes larger than 0.6 and remains above 0.6 for the rest of the winter. Results shown in Fig. 6 indicate that this day has advanced considerably during the last 20 years with a statistically significant trend of 13.6 ± 9.7 days per decade. A similar analysis using an albedo value of 0.4 as threshold resulted in an almost identical trend of 12.7 ± 8.6 days per decade.

These results strongly support the hypothesis that the observed downward trend of UV irradiance is caused by the advancement of the day of persistent snow cover.

To corroborate this conclusion, the trend of downwelling short-wave irradiance from the BSRN data set was also calculated. For October, the trend of the monthly mean noontime irradiance was $-12.5\% \pm 13.9\%$ per decade, which is slightly smaller than the trend calculated for the UV irradiance. This result is consistent with the weak wavelength dependence of the albedo effect in the presence of clouds discussed earlier.

²⁵ length dependence of the albedo effect in the presence of clouds discussed earlier. The October trend for irradiance in the visible (400–600 nm) calculated from the Version 2 spectra of the SUV-100 is $-14.4\% \pm 16.6\%$.

The day of the year when albedo drops below 0.3 at the end of winter was also calculated. The time of snow melt varied between day 145 (25 May) and 167 (16



June). A similar analysis has been conducted by Stone et al. (2002) for the period 1941–2000. An update of the analysis spanning the years 1941–2009 is available at: http://www.esrl.noaa.gov/gmd/grad/snomelt.html. Snow melt now occurs 10 days earlier than at the beginning of this 68-year period. Most of the advance has occurred

- since the mid 1970s and is coinciding with a major shift in atmospheric circulation that occurred in the North Pacific beginning in 1976 (Stone et al., 2005). The trend estimate for the period 1992–2009 is -1.4±5.9 days per decade. This trend is an order of magnitude smaller than that calculated for the onset of snow cover in the fall discussed above, and not statistically significant. The absence of a trend in UV irradiance for June (Table 1) is consistent with the lack of a clear change in the timing of snow melt over
- ¹⁰ (Table 1) is consistent with the lack of a clear change in the timing of snow melt over the last two decades.

It is of interest to compare trend estimates for the UV Index with trends in total ozone. Total ozone was retrieved consistently for all years from measured UV spectra according to Bernhard et al. (2003). The algorithm takes seasonal variation in the vertical

¹⁵ ozone and temperature profiles into account and has been validated against measurements of a Dobson photometer operated by NOAA at Barrow and observations from the Total Ozone Mapping Spectrometer (TOMS) on NASA's Earth Probe satellite (Bernhard et al., 2003). Figure 7 shows trends of total ozone calculated with the same method as that applied to E345 and the UV Index. Trends are not statistically significant ²⁰ with the exception for the trend for July, which is 3.6 ± 3.3 % (Table 1).

The relationship between the percental change in the UV Index and the associated percental changes in total ozone is often expressed with Radiation Amplification Factors (RAF) (Booth and Madronich, 1994). For SZAs between 0° and 50°, RAF is about 1.1, meaning that a 1% decrease in total ozone causes a 1.1% increase in the UV Index. RAF depends somewhat on solar zenith angle and total ozone and is often smaller than 1 for the large SZAs prevailing at high latitudes (Micheletti et al., 2007; WMO, 2011). Trends in the UV Index, denoted $\hat{T}_{UVI}(m)$, were estimated from trends in total ozone ($T_{O_2}(m)$) and trends in E345 ($T_{E345}(m)$) using the relationship

 $\hat{T}_{\text{UVI}}(m) = T_{\text{E345}}(m) - \text{RAF}(\overline{\text{SZA}}(m), \overline{\text{O}_3}(m)) \times T_{\text{O}_3}(m),$ 26638



(10)

where $\overline{SZA}(m)$ and $\overline{O_3}(m)$ are the average noontime SZA and average total ozone column for month *m*, respectively. RAF($\overline{SZA}(m), \overline{O_3}(m)$) is the associated Radiation Amplification Factor. Equation (10) assumes that trends in the UV Index caused by factors other than ozone (e.g., clouds and albedo) can be characterized with $T_{E345}(m)$. In Table 2, the estimated trend of the UV Index, $\hat{T}_{UVI}(m)$, is compared with UV Index trends determined from the regression analysis of Sect. 4, $T_{UVI}(m)$. The difference between $T_{UVI}(m)$ and $\hat{T}_{UVI}(m)$ is smaller than ±1.3%, except for August when the difference is 2.6%. These differences are well within the uncertainty of the trend analysis

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- and confirm that the trends of $T_{E345}(m)$, $T_{UVI}(m)$, $T_{O_3}(m)$ are self-consistent. Trends in the UV Index controlled for trends in E345 (i.e., $T_{UVI}(m) - T_{E345}(m)$) anti-correlate with trends in ozone $\overline{O_3}(m)$ for the months of February through August, as would be expected. UV Indices in September and October do not anti-correlate with trends in ozone, which may be explained with the dominance of cloud and albedo effects for these months.
- Results presented in Sect. 4 have shown that trend estimates depend only weakly on the correction method. I also calculated trends without gap correction and varied the minimum numbers of days required per month between 15 and 25. As expected, trend uncertainties without gap correction were larger. Requiring a minimum of 25 days per months reduced the number of years available for the trend analysis considerably,
- ²⁰ and I found that 20 days is the best compromise between the two competing desires of accurately calculating a monthly average while having as many years as possible available for the trend analysis.

The correction factors of Methods 2–4 were determined from clear sky data. Some systematic errors of the instrument such as the cosine error depend on sky condition.

²⁵ The application of correction factors established for clear skies to cloudy conditions is therefore subject to an uncertainty, which was not considered.



6 Conclusions

Trends in monthly average solar irradiance were calculated from spectral UV measurements performed near Barrow, Alaska, between 1991 and 2011. The analysis focused on two quantities: spectral irradiance integrated over the wavelength band of 342.5

- to 347.5 nm ("irradiance at 345 nm") and the UV Index. An important objective of this analysis was to quantify the effects of measurement uncertainties and data gaps on the ability to detect statistically significant trends in UV radiation. The method can also be applied to similar environmental data sets. The data set was further tested for systematic errors using radiative transfer calculations, and three different correction
- schemes to reduce these errors were explored. Depending on the correction method, estimates of decadal trends changed between 1.5% and 2.9%. Overall, it can be concluded that systematic errors in the measurements do not play a decisive role in limiting the detectability of trends. Measurement uncertainties have the largest effect on UV irradiance during the spring period when natural variability is small because of
- ¹⁵ large surface albedo. Trend estimates varied between -14% and +5% per decade and were significant (95.45% confidence level) only for October. Trends for this month varied between -11.4% and -13.7% for irradiance at 345 nm and between -11.7% and -14.1% for the UV Index. These large negative trends were confirmed with an independent data set of short-wave solar irradiance measured with pyranometers at
- ²⁰ NOAA's Barrow Observatory and can be explained with a change in snow cover over the observation period: analysis of pyranometer data indicates that the first day of fall when albedo becomes larger than 0.6 after snow fall, and remains above 0.6 for the rest of the winter, has advanced with a statistically significant trend of 13.6 ± 9.7 days per decade. Trends of total ozone were compared with trends in UV Index, and the
- two trends were found to be consistent. For ozone, a significant positive trend of 3.6 % was observed for July. Trend estimates for February–June were also positive. These positive trends in total ozone for spring and summer months are consistent with recent reports of the ozone layer's recovery (WMO, 2011), but observations over longer time



periods are required to confirm that these trends are sustainable. Results indicate that factors affected by a warming climate, such as snow cover, may affected the future Arctic UV climate more than changes in stratospheric ozone concentrations.

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References

ACIA: Arctic Climate Impact Assessment, 1042 pp., Cambridge Univ. Press, New York, 2005.

- Bernhard, G., Booth, C. R., and McPeters, R. D.: Calculation of total column ozone from global UV spectra at high latitudes, J. Geophys. Res., 108(D17), 4532, doi:10.1029/2003JD003450, 2003.
 - Bernhard, G., Booth, C. R., and Ehramjian, J. C.: Version 2 data of the National Science Foundation's Ultraviolet Radiation Monitoring Network: South Pole, J. Geophys. Res., 109, D21207, doi:10.1029/2004JD004937, 2004.
- 25
- Bernhard, G., Booth, C. R., Ehramjian, J. C., Stone, R., and Dutton, E. G.: Ultraviolet and visible radiation at Barrow, Alaska: Climatology and influencing factors on the basis of version 2 National Science Foundation network data, J. Geophys. Res., 112, D09101, doi:10.1029/2006JD007865, 2007.



Bernhard, G., Booth, C. R., and Ehramjian, J. C.: Comparison of UV irradiance measurements at Summit, Greenland; Barrow, Alaska; and South Pole, Antarctica, Atmos. Chem. Phys., 8, 4799–4810, doi:10.5194/acp-8-4799-2008, 2008.

Booth, C. R. and Madronich, S.: Radiation amplification factors: Improved formulation accounts

for large increases in ultraviolet radiation associated with Antarctic ozone depletion, in: Ultraviolet Radiation in Antarctica: Measurement and Biological Effects, edited by: Weiler, C. S. and Penhale, P. S., AGU Antarct. Res. Ser., 62, 39–52, 1994.

Blumthaler, M. and Ambach, W.: Solar UVB-albedo of various surfaces, Photochem., Photobiol., 48, 85–88, 1988.

¹⁰ Comiso J. C., Parkinson C. L., Gersten R., and Stock, L.: Accelerated decline in the Arctic sea ice cover, Geophys. Res. Lett., 35, L01703, doi:10.1029/2007GL031972, 2008.

Draper, N. R. and Smith, H.: Applied regression analysis, 3rd edition, 706 pp., John Wiley & Sons Inc., New York, ISBN 0-471-17082-8, 1998.

Dutton, E. G., Farhadi, A., Stone, R. S., Long, C. N., and Nelson, D. W.: Long-term variations in

- the occurrence and effective solar transmission of clouds as determined from surface-based total irradiance observations, J. Geophys. Res., 109, D03204, doi:10.1029/2003JD003568, 2004.
 - Dutton, E. G., Nelson, D. W., Stone, R. S., Longenecker, D., Carbaugh, G., Harris, J. M., and Wendell, J.: Decadal variations in surface solar irradiance as observed in a globally remote network, J. Geophys. Res., 111, D19101, doi:10.1029/2005JD006901, 2006.

 network, J. Geophys. Res., 111, D19101, doi:10.1029/2005JD006901, 2006.
 Gehrke, C., Johanson, U., Callaghan, T. V., Chadwick, D., and Robinson, C. H.: The impact of enhanced ultraviolet-B radiation on litter quality and decomposition processes in Vaccinium leaves from the Subarctic, Oikos, 72, 213–222, 1995.

Gwynn-Jones, D.: Enhanced UV-B radiation and herbivory, in: animal responses to global

change in the North, edited by: Hofgaard, A., Ball, J. P., Danell, K., and Callaghan, T. V., Ecol. Bull., 47, 77–83, 1999.

Gurney, K. R.: Evidence for increasing ultraviolet irradiance at Point Barrow, Alaska, Geophys. Res. Lett., 25, 903–906, 1998.

Hessen, D. O. (ed.): UV Radiation and Arctic Ecosystems, 321 pp., Springer-Verlag, New York, 2001.

30

Hicke, J. A., Slusser, J., Lantz, K., and Pascual, F. G.: Trends and interannual variability in surface UVB radiation over 8 to 11 years observed across the United States, J. Geophys. Res., 113, D21302, doi:10.1029/2008JD009826, 2008.





Holick, M. F.: Vitamin D deficiency, New. Engl. J. Med., 357, 266–281, 2007. International Standards Organization (ISO): Guide to the expression of uncertainty in measurement, 101 pp., ISO, Geneva, Switzerland, 1993.

- Krotkov, N., Bhartia, P., Herman, J., Fioletov, V., and Kerr, J.: Satellite estimation of spectral sur-
- face UV irradiance in the presence of tropospheric aerosols 1. Cloud-free case, J. Geophys. Res., 103(D8), 8779–8793, 1998.
 - Krotkov, N., Herman, J., Bhartia, P., Fioletov, V., and Ahmad, Z.: Satellite estimation of spectral surface UV irradiance 2. Effects of homogeneous clouds and snow, J. Geophys. Res., 106(D11), 11743–11759, 2001.
- Krzyścin, J. W., Sobolewski, P. S., Jarosławski, J., Podgórski, J., and Rajewska-Więch, B.: Erythemal UV observations at Belsk, Poland, in the period 1976-2008: data homogenization, climatology, and trends, Acta Geophys., 59, 155–182, doi:10.2478/s11600-010-0036-3, 2011.

Lenoble, J., Kylling, A., and Smolskaia, I.: Impact of snow cover and topography on ul-

- traviolet irradiance at the alpine station of Briançon, J. Geophys. Res., 109, D16209, doi:10.1029/2004JD004523, 2004.
 - Lindfors, A. V., Arola, A., Kaurola, J., Taalas, P., and Svenøe, T.: Long-term erythemal UV doses at Sodankylä estimated using total ozone, sunshine duration, and snow depth, J. Geophys. Res., 108, 4518, doi:10.1029/2002JD003325, 2003.
- Mayer, B. and Kylling, A.: Technical note: The libRadtran software package for radiative transfer calculations – description and examples of use, Atmos. Chem. Phys., 5, 1855–1877, doi:10.5194/acp-5-1855-2005, 2005.
 - Maykut, G. A. and Church, P. E.: Radiation climate at Barrow, Alaska, 1962–66, J. Appl. Meteorol., 12, 620–628, 1973.
- McKinlay, A. F. and Diffey, B. L. (eds.): A reference action spectrum for ultraviolet induced erythema in human skin, CIE J., 6, 17–22, 1987.
 - Meehl, G. A., Stocker, T. F., Collins, W. D., Friedlingstein, P., Gaye, A. T., Gregory, J. M., Kitoh, A., Knutti, R., Murphy, J. M., Noda, A., Raper, S. C. B., Watterson, I. G., Weaver, A. J., and Zhao, Z.-C.: Global Climate Projections, in: Climate Change 2007: The Physical
- Science Basis, Contribution of Working Group I to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change, edited by: Solomon, S., Qin, D., Manning, M., Chen, Z., Marquis, M., Averyt, K. B., Tignor, M., and Miller, H. L., Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, 2007.



- Micheletti, M. I., Piacentini, R. D., and Madronich, S.: Sensitivity of biologically active UV radiation to stratospheric ozone changes: effects of action spectrum shape and wavelength range, Photochem. Photobiol., 78, 456–461, doi:10.1562/0031-8655(2003)0780456SOBAUR2.0.CO2, 2003.
- Moody, S. A., Newsham, K. K., Ayres, P. G., and Paul, N. D.: Variation in the responses of litter and phylloplane fungi to UV-B radiation (290–315 nm), Mycol. Res., 103, 1469–1477, 1999.
 Nichol, S. E., Pfister, G., Bodeker, G. E., McKenzie, R. L., Wood, S. W., and Bernhard, G.: Moderation of cloud reduction of UV in the Antarctic due to high surface albedo, J. Appl. Meteorol., 42, 1174–1183, 2003.
- Press, W. H., Flannery, B. P., Teukolsky, S. A., and Vetterling, W. T.: Numerical recipes: the art of scientific computing, 818 pp., Cambridge University Press, Cambridge, UK, ISBN 0-521-30811-9, 1986.
 - Rex, M., Salawitch, R. J., Deckelmann, H., von der Gathen, P., Harris, N. R. P., Chipperfield, M. P., Naujokat, B., Reimer, E., Allaart, M., Andersen, S. B., Bevilacqua, R., Braathen, G.
- O., Claude, H., Davies, J., De Backer, H., Dier, H., Dorokhov, V., Fast, H., Gerding, M., Godin-Beekmann, S., Hoppel, K., Johnson, B., Kyrö, E., Litynska, Z., Moore, D., Nakane, H., Parrondo, M. C., Risley Jr., A. D., Skrivankova, P., Stübi, R., Viatte, P., Yushkov, V., and Zerefos, C.: Arctic winter 2005: Implications for stratospheric ozone loss and climate change, Geophys. Res. Lett., 33, L23808, doi:10.1029/2006GL026731, 2006.
- 20 Ricchiazzi, P., Gautier, C., and Lubin, D.: Cloud scattering optical depth and local surface albedo in the Antarctic: simultaneous retrieval using ground-based radiometry, J. Geophys. Res., 100, 21091–21104, 1995.
 - Serreze, M. C., Holland, M. M., and Stroeve, J.: Perspectives on the Arctic's shrinking sea-ice cover, Science, 315, 1533–1536, doi:10.1126/science.1139426, 2007.
- Stephens, M. A.:, EDF Statistics for Goodness of Fit and Some Comparisons, J. Am. Stat. Assoc., 69, 730–737, 1974.
 - Stone, R. S., Dutton, E. G., Harris, J. M., and Longenecker D.: Earlier spring snowmelt in northern Alaska as an indicator of climate change, J. Geophys. Res., 107(D10), 4089, doi:10.1029/2000JD000286, 2002.
- ³⁰ Stone, R., Douglas, D., Belchansky, G., and Drobot S.: Correlated declines in Pacific Arctic snow and sea ice cover, Arctic Research of the United States, 19, 18–25, 2005.
 - Tanskanen, A., Lindfors, A., Määttä, A., Krotkov, N., Herman, J., Kaurola, J., Koskela, T., Lakkala, K., Fioletov, V., Bernhard, G., McKenzie, R., Kondo, Y., O'Neill, M., Slaper, H.,



den Outer, P., Bais, A. F., and Tamminen, J.: Validation of daily erythemal doses from Ozone Monitoring Instrument with ground-based UV measurement data, J. Geophys. Res., 112, D24S44, doi:10.1029/2007JD008830, 2007.

United Nations Environment Programme (UNEP): Environmental effects of ozone depletion

and its interactions with climate change: 2010 assessment, 236 pp., UNEP, Nairobi, Kenya, ISBN: ISBN 92-807-2312-X, 2010.

Weatherhead, E. C., Reinsel, G. C., Tiao., G. C., Meng, X.-L., Choi, D., Cheang, W.-K., Keller, T., DeLuisi, J., Wuebbels, D. J., Kerr, J. B., Miller, A. J., Oltmans, S. J., and Frederick, J. E.: Factors affecting the detection of trends: Statistical considerations and applications to

¹⁰ environmental data, J. Geophys. Res., 103(D14), 17149-17161, 1998.

World Health Organisation (WHO): Global Solar UV Index: A Practical Guide, 28 pp., ISBN 92-4-159007-6, Geneva, Switzerland, 2002.

World Meteorology Organisation (WMO): Report of the WMO-WHO meeting of experts on standardization of UV Indices and their dissemination to the public, Global Atmos. Watch

¹⁵ Rep. 127, Geneva, Switzerland, 1998.

World Meteorological Organization (WMO): Scientific assessment of ozone depletion: 2002, Global Ozone Res. Monit. Proj. Rep., 47, 498 pp., Geneva, Switzerland, 2003.

World Meteorological Organization (WMO): Scientific assessment of ozone depletion: 2006, Global Ozone Res. Monit. Proj. Rep., 50, 572 pp., Geneva, Switzerland, 2007.

20 World Meteorology Organisation (WMO): Scientific assessment of ozone depletion: 2010, Global Ozone Res. Monit. Proj. Rep., 52, 516 pp., Geneva, Switzerland, 2011.



Table 1. Statistics for trend analysis. Irradiance at 345 nm is abbreviated "E345;" the UV Index is abbreviated "UVI". Trends are provided as "percental change per decade" [$\% d^{-1}$]. Significant trends are printed bold face.

Quantity	Correction	n	<i>T</i>	u(T)	$u_{U}(T)$	$u_{\rm G}(T)$	$u_{\rm S}(T)$	$R_{\cup}(T)$	$R_{\rm G}(T)$
	Method		[% d ⁻⁺]	[% d ⁻ ']	[% d ⁻ ']	[% d ⁻⁺]	[% d ⁻ ']	[%]	[%]
February									
E345	1	19	-0.6	3.3	2.5	3.2	-	57	92
E345	2	19	-0.3	3.5	2.5	3.2	-	51	80
E345	3	19	1.0	4.1	2.5	3.2	-	39	62
E345	4	19	-0.6	3.3	2.5	3.2	-	57	92
UVI	1	19	-2.6	4.1	2.5	3.6	-	38	78
UVI	2	19	-2.4	4.1	2.5	3.6		38	/5
	3	19	-1.1	4.6	2.6	3.6	1.4	30	51
Ozono	4	19	-2.0	4.1	2.5	3.0	21	30	10
Ozone	I	10	1.4	3.4	-	1.4	3.1	-	10
March									
E345	1	18	0.0	3.2	2.6	0.9	1.7	64	8
E345	2	18	-0.2	3.3	2.6	0.9	1.8	61	7
E345	3	18	1.5	3.4	2.6	0.9	1.9	60	7
E345	4	18	0.6	3.9	2.6	0.9	2.8	44	5
UVI	1	18	-0.9	7.1	2.6	1.2	6.5	13	3
UVI	2	18	-1.0	6.8	2.6	1.2	6.2	15	3
001	3	18	0.6	7.0	2.6	1.2	6.4	14	3
Ozono	4	10	-0.2	7.0	2.6	1.2	6.4	14	3
Ozone	1	19	1.0	0.3		0.2	0.3	-	0
April									
E345	1	19	-1.3	2.8	2.5	0.4	1.2	79	2
E345	2	19	-1.3	3.0	2.5	0.4	1.6	70	2
E345	3	19	0.5	3.5	2.5	0.4	2.3	53	1
E345	4	19	-0.9	3.0	2.5	0.4	1.6	69	2
UVI	1	19	-3.4	5.0	2.5	0.7	4.2	26	2
	2	19	-3.3	5.0	2.5	0.7	4.2	26	2
	3	10	-1.0	3.0	2.5	0.7	3.2	19	2
Ozone	4	21	-2.9	4.0	2.0	0.7	4.0	20	2
May		21	2.0	0.1		0.1	0.1		•
E345	1	17	1.6	5.7	2.8	0.1	5.0	24	0
E345	2	17	1.5	5.5	2.8	0.1	4.7	26	0
E345	3	1/	3.7	4.9	2.8	0.1	4.0	33	0
E345	4	17	3.1	6.3	2.8	0.1	5.6	20	0
	1	17	-0.4	5.9 5.7	∠.8 2.9	0.4	5.2 4 0	∠3 25	0
	2	17	-0.4 1 º	5.7	2.0	0.4	4.9	20	0
	3	17	1.0	6.5	2.9	0.4	5.3	10	0
Ozone	4	19	1.2	4.0	2.0	0.4	4.0	- 15	0
520110			1.0	4.0		0.0	4.0		

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Table 1. Continued.

	June									
ĺ	E345	1	18	-0.3	7.2	2.9	0.5	6.6	16	0
	E345	2	18	0.2	7.4	2.9	0.5	6.8	15	0
	E345	3	18	2.6	6.5	2.9	0.5	5.8	20	1
	E345	4	18	0.1	8.3	2.9	0.5	7.8	12	0
	UVI	1	18	-1.8	6.9	2.9	0.3	6.3	17	0
	UVI	2	18	-1.3	6.7	2.9	0.3	6.1	18	0
		3	18	1.0	0.4	2.9	0.3	5.6	21	0
	Ozone	4	20	-1.4	2.5	2.9	0.3	2.7	15	0
	020110		20	2.5	2.1	_	0.2	2.1	_	0
	July									
	E345	1	19	-2.6	7.8	2.6	0.6	7.4	11	1
	E345	2	19	-2.5	8.1	2.6	0.6	7.6	11	1
	E345	3	19	-0.7	7.9	2.6	0.6	7.4	11	1
	E345	4	19	-1.3	8.0	2.6	0.6	7.5	11	1
		2	19	-0.2	9.1	2.7	0.7	8.7	8	1
		2	19	-0.1	9.0	2.7	0.7	0.0 8.8	9	1
	UVI	4	19	-5.0	9.3	27	0.7	8.9	8	i
	Ozone	1	19	3.6	3.3		0.1	3.3	_	0
1	August									
	FOAF	-	10	0.0	7.0	0.0	0.0	7.0	10	-
	E345	2	19	0.8	7.8	2.8	0.8	7.3	13	1
	E345	2	10	2.6	8.7	2.0	0.0	8.2	10	÷
	E345	4	19	2.0	8.1	2.0	0.0	7.5	12	i
	UVI	1	19	3.6	7.2	2.8	1.3	6.6	15	3
	UVI	2	19	3.7	7.5	2.8	1.3	6.9	14	3
	UVI	3	19	5.4	8.0	2.8	1.3	7.4	12	3
	UVI	4	19	5.3	7.6	2.8	1.3	6.9	13	3
	Ozone	1	19	-0.1	2.9	-	0.1	2.9	-	0
	Septem	ber								
Ì	E345	1	17	-3.1	12.6	3.0	0.5	12.2	6	0
	E345	2	17	-2.9	12.4	3.0	0.5	12.0	6	0
	E345	3	17	-1.4	12.6	3.0	0.5	12.2	6	0
	E345	4	17	-3.0	12.7	3.0	0.5	12.3	5	0
	UVI	1	17	-2.7	13.2	3.0	0.8	12.8	5	0
	UVI	2	17	-2.4	13.0	3.0	0.9	12.7	5	0
	UVI	3	17	-0.9	13.2	3.0	0.8	12.8	5	0
	Ozone	4	1/	-2.5	13.2	3.0	0.8	12.8	5	0
4	Ostabas		10	0.5	4.7		0.1	4.7		0
,	October									
	E345	1	19	-13.3	12.4	2.7	0.8	12.0	5	0
	E345	2	19	-13.4	12.6	2.7	0.8	12.3	5	0
	E345	3	19	-11.4	10.4	2.7	0.8	12.0	6	0
	E345	4	10	-13.7	12.4	2.1	0.0	12.0	5 1	1
		2	19	-13.7	13.0	2.1	1.4	12.0	4	1
	UVI	3	19	-11.7	12.2	2.7	1.4	11.8	5	1
	UVI	4	19	-14.1	13.0	2.7	1.4	12.7	4	1
	Ozone	1	17	-0.8	4.5	-	0.2	4.5	-	0

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Table 2. Comparisons of trends in the UV Index (\hat{T}_{UVI}) estimated from trends in irradiance at 345 nm (T_{E345}) and trends in ozone (T_{O_3}) with trends in the UV Index determined from direct measurements (T_{UVI}).

Month	SZA	$\overline{O_3}$	RAF	Decadal Trend (%)				
	(degree)	(DU)		T _{E345}	T_{O_3}	\hat{T}_{UVI}	T _{UVI}	
February	84	413	0.6	-0.6	1.4	-1.4	-2.6	
March	73	435	0.8	0.6	1.8	-0.8	-0.2	
April	61	424	1	-0.9	2.6	-3.5	-2.9	
May	52	385	1.1	3.1	1.8	1.1	1.2	
June	48	350	1.1	0.1	2.5	-2.7	-1.4	
July	50	323	1.2	-1.3	3.6	-5.6	-5	
August	58	301	1.1	2.6	-0.1	2.7	5.3	
September	69	300	1	-3	0.5	-3.5	-2.5	
October	80	319	0.8	-13.7	-0.8	-13.1	-14.1	

 T_{345E} and T_{UVI} are based on the data set corrected with Method 4. SZA and $\overline{O_3}$ are the average solar zenith angle, and average total ozone column. RAF is the Radiation Amplification Factor.





Fig. 1. Ratios between measurement and model at 345 nm for different averaging intervals ("q-ratios"). Open blue symbols: monthly q-ratios ($q_{monthly}(y,m)$); solid blue symbols: subset of $q_{monthly}(y,m)$ where the median was calculated from at least 10 spectra; pink solid symbol connected by lines: $q_{annual}(y)$; green solid symbol connected by lines: $q_{summer}(y)$. Outliers discussed in the text are indicated by red circles. $q_{monthly}(y,m)$ used for the correction are indicated by a cyan border. Values of $q_{summer}(y)$ not used for the correction have a black core. Broken lines drawn in yellow (orange) indicate 1.0 ($\bar{q}_{monthly}$). Solid lines drawn in yellow (orange) indicate the range of 1±0.060 ($\bar{q}_{monthly} \pm 0.060$).





Fig. 2. Time series analysis of irradiance at 345 nm. Each panel shows results for a different month, starting with February (top left) and ending with October (bottom right). Black symbols connected by black lines indicate the monthly means $\bar{E}(y_i)$ of measured noontime spectral irradiance at 345 nm. Data have been corrected for systematic error using Method 4. Error bars indicate the measurement uncertainty $u_U(\bar{E}(y_i))$. Trend estimates are indicated by black lines. Their 95.45% confidence bands are shown in pink shading. The lower half of each plot gives the square of the correlation coefficient (R^2), the estimate of the decadal trend *T* (black number), the uncertainty of the trend estimate u(T) (red number), the uncertainty of the trend estimate $u_U(\bar{T})$ (blue number), and the trend uncertainty caused by gaps $u_G(T)$ (green number). The hypothetical confidence bands associated with the measurement uncertainty and the gap uncertainty are indicated by blue and green lines, respectively.











Fig. 4. Same as Fig. 2, but for the UV Index.





Fig. 5. Same as Fig. 3, but for the UV Index.





Fig. 6. Day of year when surface albedo becomes larger than 0.6 and remains above 0.6 for the rest of the winter. The red line is a regression to the data. The broken line indicates 1 October.





Fig. 7. Same as Fig. 2, but for total ozone. The measurement uncertainty was set to zero.

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