

Interactive comment on “Correcting for trace gas absorption when retrieving aerosol optical depth from satellite observations of reflected shortwave radiation” by F. Patadia et al.

Anonymous Referee #1

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Important small detail for consistent climate data records

The paper discusses corrections to mid-visible satellite aerosol retrievals for the impact of the small but non-negligible trace gas absorption. It rightfully states that this wellknown correction remains often at the side of publications on aerosol retrieval and not much detail is provided. The authors make a thorough quantitative assessment of the impact of different absorbing trace gases on window channels used in AOD retrieval. They relate the strength of absorption to typical AOD uncertainties and clearly show that while on global average trace gas absorption in window channels remains within AOD uncertainties, on regional scale it clearly does exceed them. The authors demonstrate the importance of such an accurate trace gas correction for consistent climate data records by assessing the consequences of tiny differences between similar spectral channels of two similar but not identical sensors MODIS and VIIRS which can be used to constitute a long-term AOD record. This fully falls into the scope of AMT and sheds detailed light on such a “small” but still important aspect of long-term climate quality data records. The paper is well written with clear arguments and conclusions supported by substantial data presented in appropriate tables and figures. Title and abstract summarize clearly the content of the paper and its essence. The scientific methods used are state of the art and clearly referenced where suitable.

I therefore recommend to accept the paper after some technical corrections.

[Thank you very much for a thorough review and for suggestions that have improved the paper. We have incorporated all of them.](#)

I have only three more general suggestions and a number of small issues (see detailed comments).

1) I recommend to consistently use the term “atmospheric gas correction” rather than “atmospheric correction” since the latter is commonly used when correcting surface observations for the impact of atmospheric scattering (molecules and aerosols) and trace gas absorption.

[Done](#)

2) Regarding references I suggest to add a few references for leading European aerosol retrieval algorithms, mainly in context of the ESA Climate Change Initiative CCI: a) In addition to the reference to Hollmann et al. 2013 (overview paper of the entire CCI program with its 13 ECVs) a paper on the Aerosol project in CCI should be added: “Popp, et al., Development, Production and Evaluation

of Aerosol Climate Data Records from European Satellite Observations (Aerosol_cci), Remote Sensing, 8, 421; doi:10.3390/rs8050421, 2016” or “de Leeuw, et al., Evaluation of seven European aerosol optical depth retrieval algorithms for climate analysis, Remote Sensing of Environment, 162, 295-315, doi: 10.1016/j.rse.04.023, 2015”

Added

b) When referring to the GCOS requirements, the latest GCOS implementation plan 2016 with its Annex A ECV Product requirement tables (aerosol on page 281) should be added: GCOS-IP 2016: GCOS Implementation Plan 2016. GCOS-200. Available at https://library.wmo.int/opac/doc_num.php?explnum_id=3417

Added

c) As leading algorithm for the AATSR instrument the Swansea algorithm should be added to the references:

“Bevan, et al. (2012), A global dataset of atmospheric aerosol optical depth and surface reflectance from AATSR. Remote Sensing of Environment, 116, 119-210” and

“North, et al., Retrieval of land surface bidirectional reflectance and aerosol opacity from ATSR-2 multiangle imagery, IEEE Trans. Geosci. Remote Sens. 1999, 37, 526–537.

Added

3) Regarding figures, I suggest to increase axis legends of fig. 1, colour bar legends of fig. 6a

Done

Detailed comments:

l. 82: “[MISR ATBD]” – please add http link, date and version to make a unique reference

Done

l. 111: please also add the width of VIIRS channels

Done

l. 178/ 179: two conflicting namings for G_i are made (airmass factor or path length), which could be confusing readers

Corrected

l. 270 ff: here ten gases are discussed, while in l. 169 only 5 are identified as relevant – please harmonize or explain (see also l. 346 and 462/463)

Done

l. 304 ff: I had to read this several times and found it confusing (my first impression was that linear relationship will do then is learned that it will not do) – maybe you can rearrange to start with the clear statement that a quadratic fit is needed

Thanks for pointing this out. Rearranged.

l. 310 ff: Can you state whether the H₂O airmass factors include any effect of multiple scattering (since water vapour prevails in the lower troposphere)?

LBLRTM does not model multiple scattering. The only scattering effect modeled in LBLRTM is Rayleigh extinction.

l. 318: instead of “Gi“ and “gas i” it should be “GO3“ and “O3”
Corrected. Thanks.

l. 332: What is the temporal resolution of the NCEP analysis?
It is 6 hrs. Information in Line 374

l. 386: table 4 contains also another line on Rayleigh OD, which should also be mentioned and brought into perspective in the text
It has been edited out since this information is not relevant to this paper and was inadvertently left in the table

l. 415: correct in the middle: “that that”
Done

l. 486: Another aspect of interest would be whether an overlooked longterm trend in water vapour or ozone concentrations (e.g. by using a static climatology) could create an artificial AOD trend – if you could add a statement on this (whether its relevant or negligible) based on an assumed possible decadal trend in concentrations, this would be very useful?

For H2O and O3, we almost never need to use climatology since we use NCEP data and that can account for trend from these gases. However, for the ‘dry gas’ we use climatology. For the DT algorithm, the important gases in category are CO2 (MODIS) and CH4 (VIIRS).

Over the lifetime of MODIS instrument, CO2 has increased from about 375 to 405 (i.e by ~10%). A 10% change in CO2 amount = $10 * 0.0163 = 0.00163$ CO2 Optical depth at 2.12 μm (MODIS Channel impacted by CO2 absorption). Not accounting for this trend in atmospheric gas correction means that the DT algorithm would retrieve lower AODs and smaller trends. In separate work, we have also estimated the uncertainty in the retrieved AOD due to error in the water vapor data used by the DT algorithm. A 20% error in water vapor content results in AOD uncertainty of ~0.002 (median) , ~0.003 (mean). The uncertainty magnitude was similar for different months of global data we looked at. Therefore, we think that the impact of using climatological values for ‘dry gases’ would have an overall negligible impact on the decadal trends.

Fig. 7 legend vs. text in l. 427: differences are smaller than 0.08 (figure) or 0.07 (text) – can you harmonize?
Corrected.

Table 4 / Rayleigh OD, 2nd column: correct “Fiter” to “Filter”
Done

AMTD
Interactive
comment

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Interactive comment on “Correcting for trace gas absorption when retrieving aerosol optical depth from satellite observations of reflected shortwave radiation” by F. Patadia et al.

A.C. Povey (Referee)
adam.povey@physics.ox.ac.uk

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This paper outlines a revision of the Dark Target algorithm to better account for the absorption of common atmospheric gases. The radiative transfer code and spectral database have been updated since Collection 5, but the essence of the method is unchanged. A gas’s optical path is modelled as a linear (or, in the case of water, quadratic) function of gas path length. Ten gases were evaluated, and the regression coefficients are reported for both the MODIS and VIIRS sensors. It concludes with a brief reminder that ignoring the difference in spectral response between two instruments can produce non-trivial errors.

Though this paper has a rather small audience, it is well within the remit of this journal and is the sort of work that is often overlooked. Other than a few minor corrections, I recommend it for publication.

[Thank you very much for a thorough review and for suggestions that have improved the paper. We have incorporated all of them.](#)

A few matters that warrant consideration:

L28 Though I appreciate the simplicity of the language, aerosols aren’t necessarily ‘tiny’. Hinds described them as ‘fine’ but I find ‘Aerosols are particles in the atmosphere’ is usually sufficient.

[We have replace ‘tiny’ with ‘fine’.](#)

L285 It is true that these gases are usually well mixed. However, a not-insignificant number of users of aerosol data study emissions from volcanoes and fires. As those emit many of the gases you are studying in significant quantities, do you have any estimate of the magnitude of errors that will result from using climatological concentrations there? A back-of-the-envelope calculation could be quite informative.

[We agree that the climatological concentration of gases is not representative of their](#)

concentrations at sources such as volcanoes, fires, industries etc. We are using climatological optical depth for CO₂, N₂O, O₂ and CH₄. Some of these gases are emitted by fires (CO₂, N₂O, and CH₄) and volcanoes (CO₂, SO₂). These gases are transparent to solar radiation in the shortwave channels used for aerosol retrieval. Tables 2.1, 2.2 show that these gases absorb in the 1.6 μm and the 2.1 μm channels.

CO₂ concentrations (AIRS Maps) indicate ~2% [400 ppm to 408 ppm] increase in fire locations compared to background. In the 2.1 μm channel, used by DT algorithm, 2% increase in CO₂ concentration would translate to ~0.001 CO₂ optical depth, which would not change the total gas optical depth in 2.1 μm channel by a lot.

Similarly, from CH₄ maps from volcano events, there seems to range by ±65 ppbv (~3.5%). For VIIRS, CH₄ optical depth in 2.1 μm channel (0.04914) will change by ~0.002.

So, using climatology for gases other than H₂O and O₃, will impact the atmospheric gas corrections near source regions. However, since these gases absorb in IR, where signal is low, the errors will be small.

L304-9 Could you please provide some quantitative indication of the quality and uncertainty of these fits (e.g. root-mean-square deviation and the maximum error)?

To-Do

This would be particularly instructive for water vapour, where I would appreciate a more scientific justification for using a quadratic fit (or, vice versa, a justification for using linear fits with everything else).

Thank you for this comment. We have revised the text with additional information : The quadratic nature seems to be rooted in the absorption characteristics of water vapor. Water vapor absorbs over a wide range of electromagnetic spectrum. In different regions of the spectrum the molecule absorbs radiation via different absorption mechanism (rotational transitions and intermolecular vibrations in microwave and IR, intramolecular vibrational transitions in IR and electronic transitions in UV). The spectra are also all different for isotopic variants of water (e.g. HDO, D₂O, H₂¹⁸O) and is also a function of pressure and temperature. It is also found that with greater levels of water vapor the increase in water dimer will be nearly quadratic, and increase the direct absorption of sunlight to a greater extent, than water monomer absorption and line broadening. ~~Laboratory measurements show a broad absorption at 405 μm with quadratic dependence on water monomer concentration, enhancement of water dimer absorption at 397–532 nm and the lack of water dimer absorption at 570 nm.~~ The vertical distribution of water vapor in the atmosphere may result in the quadratic nature of absorption/transmission relationship depicted by RT calculations.

L336-8 While I appreciate that in normal operations you can't use the MODIS water vapour product, you've presumably tried using it offline. Could you quantify approximately how much difference it makes to the final product?

We haven't used the MODIS water vapor product. However, we have estimated the uncertainty in the retrieved AOD due to error in the water vapor data used by the DT algorithm. A 20% error in water vapor content results in AOD uncertainty of ~0.002 (median), ~0.003 (mean). The uncertainty magnitude was similar for different months of global data we looked at.

L441-3 This final paragraph begs the obvious question to any algorithm paper: You've proposed something that sounds sensible, but is it actually better than what you did before? Fig. 7 implies you've processed at least a month of data with the new corrections. For that data, does the RMS difference against AERONET collocations improve (or at least not significantly degrade)?

We had tested to ensure that our empirical relations (Eqs 5, 6) can reproduce the transmission calculated from LBLRTM for different water vapor and ozone profiles. At the time of doing this work, few other changes were made in the retrieval algorithm. The

Fig.1 The axes labels are far too small to be legible.
Corrected

Figs.3-4 The axes labels aren't meaningful to someone that hasn't read the text exhaustively. Also, is the y-axis of 3(a) really the log of the log of the transmission factor?

Thank you. Added more description. Yes, y-axis label is correct

Fig.6(a) Could this be the same size as 6(b) to facilitate comparison?
Done

Fig.7 There might be a good reason why not, but could the fractional difference be plotted rather than (or in addition to) the absolute difference? Over the central Pacific changes appear to be 0.01, which is rather significant there.
Done. Additional figure added

The English quality is in the upper quartile of paper's I've reviewed. Though I found the language rather repetitive, it is not my place to nitpick style. However, I do have some grammatical recommendations:

All edits suggested here have been incorporated in the paper. Thank you for the suggestions.

- "the" should precede the following words: L12 underlying, L14 Moderate, L18 High-resolution, L21 MODIS, L25 gas, L34 characterization, L40 solar, L53 accuracy, L98 context, L106 Earth's, L126 spectral reflectance, L211 coefficients, L311 gas, L326 largest, L381 HITRAN, L395 subsequent, L400 MODIS, L470 nadir.
- L17 There should be a comma after 'paper'.
- L23 AOD biases of up to
- L24 studies are attempting have attempted to create
- L35 'over broad regions' seems redundant to 'global'.
- L41 from the solar radiation interacting interaction with suspended aerosol particles aerosols from the

L51 to apply **in to** new situations. These **latter** include
L55 suggested that **for reducing to reduce** uncertainties
L67 magnitude **as to** pristine AOD, and **is** equal
L76 (well-mixed) **throughout across** the globe
L77 their absorption **would** also **would** lead to
L82 gases to be **too small to bother with negligible** [MISR ATBD]
L105 from blue **through to** the shortwave
L116 ‘observed’ seems redundant to ‘as measured by the satellite’.
L123 been made **for about** the surface
L141 each gas **were was** calculated
L195 The expression for G should be typeset as math not text.
L252 This heading should be bold.
L256 database [] **for calculating to calculate** transmittance
L267 This heading should be bold.
L323 accordance **to with** absorption
L329 w should be typeset as math not text.
L347 SO₂, **and** other trace
L348 day-to-day should be hyphenated.
L355 different gases **is** different
L357 case **of with** a small amount
L371 ‘match’ seems to be the wrong word. I think you mean ‘be consistent with’ or ‘can be used with’.
Fig.1 are overlaid **for visualizing to visualize** their positioning **in** atmospheric ‘window’
region regions where

All above edits have been incorporated in the paper. Thank you for the suggestions.

There are also a few thoughts I would like the authors to be aware of but which are unreasonable to expect a revision:

L86 Though aerosol retrievals don’t often discuss gas correction, sea surface temperature studies do because of their more stringent accuracy requirements, such as doi:10.1016/j.rse.2010.10.016. For aerosol in particular, §2.3.3.3 of the thesis of Haiyan Huang (<http://eodg.atm.ox.ac.uk/eodg/theses/Huang.pdf> and <https://ora.ox.ac.uk/objects/uuid:16e444e6-5da9-43da-a122-c50c7e6a2412>) presents a sensitivity study of TOA brightness temperature from AATSR to a variety of gases. I am curious if the authors have ever considered the importance of species such as F12 and CFCs, which Dr. Huang found to be rather important?

Dr Huang found these gases to be important in the Infrared channels. The DT algorithm doesn’t use these channels for aerosol retrievals. For the channels used in DT algorithm, we have looked at all those gases (in HITRAN database) that might have some absorption in these channels. Only the ones listed in the paper are the important ones.

L257 I agree with Dr. Gordon that you should have used a more recent version of HITRAN.

We agree. The only reason for using the earlier version of HITRAN was because at the time when modifications were made to our gas tables (2012 – 2013), HITRAN2008 was the latest version. In the next version of DT algorithm, we will update our calculations with latest HITRAN version.

L281 Many studies desire a representative set of atmospheric profiles. I appreciate that you've cited someone for making that decision. However, researchers have done statistically robust selections for the minimally representative set. For example, §3 of <https://www.ecmwf.int/sites/default/files/elibrary/2008/11040-generation-rttov-regression-coefficients-iasi-and-airis-using-new-profile-training-set-and-new.pdf>.

Thank you for pointing to this reference. We have added the relevant reference to the paper.

Interactive comment on “Correcting for trace gas absorption when retrieving aerosol optical depth from satellite observations of reflected shortwave radiation” by F. Patadia et al.

I. E. Gordon

igordon@cfa.harvard.edu

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Why HITRAN2008 is used? There are substantial improvements in the quality and extent of the spectroscopic data for atmospheric gases in HITRAN2016 (Gordon et al, J. Quant. Spectrosc. Radiat. Transf. (2017) 203, 3-69. doi:10.1016/j.jqsrt.2017.06.038) and even HITRAN2012 (Rothman et al, J. Quant. Spectrosc. Radiat. Transf. 130, 4-50 (2013) doi:10.1016/j.jqsrt.2013.07.002). In fact many trace gases did not even have data in the shortwave regions in HITRAN2008. In addition substantial improvements in the quality of the spectroscopic data and its completeness for main absorbers, including water vapor and ozone were carried out in more recent editions. HITRAN2016 data is available at www.hitran.org

Thank you for your comment. We agree with you. The only reason for using HITRAN2008 in this study is because at the time when modifications were made to our gas tables (2012 – 2013), HITRAN2008 was the latest version. In the next version of DT algorithm, we will update our calculations with the latest HITRAN version.

1 **Correcting for trace gas absorption when retrieving aerosol optical**
2 **depth from satellite observations of reflected shortwave radiation**

3 Falguni Patadia^{1,2}, Robert C. Levy², Shana Mattoo^{2,3}

4 ¹GESTAR/Morgan State University, Columbia, MD, USA

5 ²NASA Goddard Space Flight Center, Greenbelt, MD, USA

6 ³SSAI, Lanham, MD, USA

7

8

9 **Abstract**

10 Retrieving aerosol optical depth (AOD) from top-of-atmosphere (TOA) satellite-
11 measured radiance requires separating the aerosol signal from the total observed signal. Total
12 TOA radiance includes signal from the underlying surface and from atmospheric constituents
13 such as aerosols, clouds and gases. Multispectral retrieval algorithms, such as the dark-target
14 (DT) algorithm that operates upon the Moderate Resolution Imaging Spectroradiometer
15 (MODIS, onboard Terra and Aqua satellites) and Visible Infrared Imaging Radiometer Suite
16 (VIIRS, onboard Suomi-NPP) sensors, use wavelength bands in “window” regions. However,
17 while small, the gas absorptions in these bands are non-negligible and require correction. In this
18 paper, we use the High-resolution TRANsmission (HITRAN) database and Line-by-Line
19 Radiative Transfer Model (LBLRTM) to derive consistent gas corrections for both MODIS and
20 VIIRS wavelength bands. Absorptions from H₂O, CO₂ and O₃ are considered, as well as other
21 trace gases. Even though MODIS and VIIRS bands are “similar”, they are different enough that
22 applying MODIS specific gas corrections to VIIRS observations results in an underestimate of
23 global mean AOD (by 0.01), but with much larger regional AOD biases of up to 0.07. As recent
24 studies are attempting to create a long-term data record by joining multiple satellite datasets,
25 including MODIS and VIIRS, the consistency of gas correction becomes even more crucial.

26

27 1. Introduction

28 Aerosols are ~~tiny-fine~~ particles in the atmosphere that scatter and/or absorb incoming
29 solar insolation, and because of this are active players in Earth's energy budget [IPCC, 2013]. In
30 addition aerosols affect cloud and precipitation processes [Denman et al., 2007; Boucher et al.,
31 2013], and they degrade air quality, contributing to increased morbidity and mortality rates
32 world-wide [Lim et al. 2012]. For these reasons characterizing and monitoring aerosol
33 distributions has become a global priority [Boucher et al., 2013].

34 Satellite aerosol remote sensing allows for the characterization and monitoring of
35 aerosols ~~over broad regions and~~ globally [Lenoble et al., 2013]. Different aerosol remote sensing
36 schemes are applied, depending on the information received by the different satellite sensors
37 [McCormick et al., 1979; Herman et al., 1997; Stowe et al., 1997; Tanré et al., 1997; Kaufman et
38 al., 1997a; Torres et al., 1998; Veeffkind et al., 1998; Higurashi and Nakajima, 1999; Deuzé et
39 al., 1999; Knapp et al., 2002; Martonchik et al., 1998; Liu et al., 2005; Kahn et al., 2001; North
40 et al., 1999, Bevan et al., 2012]. In terms of passive satellite sensors that measure the solar
41 radiation reflected by the Earth-atmosphere system, aerosol remote sensing methods must isolate
42 the information obtained from the interaction of solar radiation ~~interacting with~~ ~~suspended~~
43 ~~aerosol-partieles~~ aerosols from the information obtained from all other interactions: reflectance
44 from the surface, scattering from atmospheric molecules and clouds, absorption by atmospheric
45 gases, etc. [Vermote et al, 1997]. Thus, characterizing and removing these other sources of
46 information in the satellite signal becomes a fundamental part of the process.

47 Some of the interactions requiring removal continue to receive considerable attention as
48 new sensors are deployed and new aerosol remote sensing algorithms are derived. These include
49 characterizing the contribution from the surface and masking clouds [Hutchison et al., 2008; Shi

50 *et al.*, 2014;]. Other interactions received much less attention, as these are considered to be well-
51 understood and simple to apply ~~in-to~~ new situations. These latter ones include molecular
52 scattering and gaseous absorption [*Tanré et al.*, 1992; *Vermote et al.*, 1997]. However, the
53 requirements on the accuracy of aerosol remote sensing products become tighter as instrument
54 capabilities, calibration and retrieval methods improve. For example, *Hollman et al.*, (2013) has
55 recently suggested that ~~for-to reducing-reduce~~ uncertainties on climate, aerosol optical depth
56 (AOD) should be monitored to an accuracy on the order of $\pm(0.03 + 10\%$; e.g. GCOS, 2011,
57 [GCOS-IP, 2016](#)). The Atmospheric Clouds and ocean Ecosystems (ACE) white paper called for
58 an accuracy of $\pm(0.02 + 10\%)$ [*Starr et al.*, 2010]. To meet such tight criteria, all aspects of
59 traditional aerosol remote sensing methods require re-examination with the objective to reduce
60 uncertainties in the final retrieval, and to assure continuity as the aerosol climate data record is
61 passed from one sensor to the next [[Popp et al.](#), 2016].

62 In this paper we focus on gaseous absorption. Aerosol retrieval algorithms (*Vermote et*
63 *al.*, 1997) tend to use satellite observations taken in wavelength regions where gas absorptions
64 are small. However, while gas absorption is small in these “window” bands, it is not zero. For
65 example, for the 20 nm-wide Moderate Resolution Imaging Spectroradiometer (MODIS) band
66 near 0.55 μm , in the middle of the Chappius region, there is absorption due to ozone. For a US
67 1976 Standard Atmosphere (US76, 1976), with total column ozone of 344 Dobson Units (DU),
68 the gas absorption optical depth (τ^{GAS}) is about 0.03 in this band. This is of similar magnitude as
69 to pristine AOD (~ 0.05), and is equal to the required measurement accuracy (GCOS; 2011,
70 [GCOS-IP, 2016](#)). Water vapor, measured as precipitable water vapor (PW or w), absorbs as well
71 and introduces even greater uncertainty. For example, the w of the US76 standard atmosphere is
72 a modest 1.4 cm, which translates to τ^{GAS} of about 0.025 in the MODIS 2.11 μm band or a τ^{GAS}

73 of 0.05 for a similar-wavelength Visible Infrared Imaging Radiometer Suite (VIIRS) band
74 centered near 2.25 μm . The major difficulty with ozone and water vapor is that the total column
75 burden of these gases varies spatially and temporally over the globe [Heggin et al., 2014]. Other
76 trace gases, including carbon dioxide and methane, also absorb shortwave radiation in
77 wavelength specific regions. While these gases are more evenly distributed (well-mixed)
78 ~~throughout across~~ the globe, failing to correct for their absorption ~~also~~ would also lead to errors
79 in aerosol retrieval.

80 Different aerosol retrieval algorithms respond to the challenge of gaseous correction
81 differently. Some include all gaseous absorbers and account for the variability of water vapor
82 and ozone [Levy et al., 2013; 2015], while others use a fixed ozone concentration [e.g Thomas et
83 al., 2010; Sayer et al., 2012], and others correct for some gases, but consider the effect of other
84 gases to be ~~too small to bother with~~ negligible [MISR ATBD 09, 2008 :
85 <https://eosps0.gsfc.nasa.gov/sites/default/files/atbd/atbd-misr-09.pdf>]. Few include methane
86 [Levy et al., 2013; 2015]. How does a less complete gaseous correction scheme affect the global
87 retrieval of AOD? How sensitive are gaseous absorption schemes to slight shifts in spectral
88 bands from instrument to instrument? While all operational aerosol retrieval algorithms employ
89 gaseous correction schemes in their retrieval and describe these schemes, more or less, within the
90 “gray literature” of internal documentation, there are few recent articles in the peer-reviewed
91 literature that openly describe the process and quantify the impact of the subtle choices made
92 during algorithm development.

93 In this paper we re-examine gaseous correction as it is applied in the traditional MODIS
94 Dark Target (DT) aerosol retrieval [Levy et al., 2013], and as that retrieval algorithm is ported to
95 the new VIIRS data [Levy et al., 2015]. In Section 2 we discuss the absorption of radiation by

96 atmospheric gases within the MODIS and VIIRS bands used for the DT aerosol retrieval. We
97 introduce the relationship of gas abundance to its transmittance spectra, which is the theoretical
98 basis for gas corrections in DT AOD retrievals. The atmospheric gas correction methodology is
99 detailed in Section 3. The impact of the updated atmospheric gas corrections applied to the
100 Collection 6 MODIS AOD is also briefed in Section 3. In Section 4 we discuss the importance of
101 accurate atmospheric gas corrections in the context of DT AOD retrievals from the VIIRS
102 instrument. The study is summarized and concluded in Section 5.

103

104 **2. The DT approach to aerosol retrieval and gas correction**

105 **2.1 The DT aerosol algorithm and wavelength bands**

106 As explained in detail by Levy et al. (2013; 2015) and references therein, the dark-target
107 (DT) aerosol algorithm uses seven channels (or bands) covering the solar reflective spectral
108 region from blue through-to the shortwave infrared (SWIR) to characterize aerosols, clouds and
109 the Earth's surface. These bands were specifically chosen to correspond to the spectral window
110 regions of minimal gas absorption. On MODIS, these bands include B1, B2, B3, B4, B5, B6 and
111 B7, which are each 20-50 nm in width and centered near 0.65, 0.86, 0.47, 0.55, 1.24, 1.63 and
112 2.11 μm , respectively. On VIIRS, the DT algorithm uses bands M3, M4, M5, M7, M8, M10 and
113 M11, which are the “moderate resolution” or M-bands with 20 and 60 nm bandwidths, ~~that and~~
114 are centered near 0.48, 0.55, 0.67, 0.86, 1.24, 1.60 and 2.26 μm , respectively.

115 The DT algorithm is actually two algorithms, one applied to MODIS- or VIIRS-measured
116 reflectance over land surfaces and the other to measured reflectance over ocean (Levy et al.,
117 2013; 2015). Both the land/ocean algorithms employ a single atmospheric gas correction method

118 before any retrieval is performed. DT uses a LUT approach in which atmospherically corrected
119 observed top-of-atmosphere (TOA) reflectance (~~as measured by the satellite~~) is compared with
120 simulated reflectance. The simulations are calculated by radiative transfer codes, and account for
121 multiple scattering and absorption effects of a combined surface (land or water), molecular
122 (Rayleigh), and aerosol scene, but do not account for gaseous absorption. These simulations also
123 account for the angular dependence of the scattered radiation, through use of a pseudo-spherical
124 approximation (e.g. Ahmad and Fraser, 1991). The DT retrieval operates on regions of pixels for
125 which cloud pixels, glint pixels, and other unsuitable pixels have been masked out. Thus, the DT
126 aerosol retrieval is performed for cloud-free sky, and assumptions have been made ~~for~~ about the
127 surface reflectance properties and atmospheric constituents. The LUT is interpolated as a
128 function of observing geometry (solar and view zenith and azimuth angles), and then searched to
129 determine which aerosol conditions provide the spectral reflectance that best “matches” the
130 spectral reflectance observed by the satellite. The reported solution (retrieved spectral AOD) is
131 some function of the solutions that meet sufficient criteria for matching the observations. For the
132 DT algorithm, expected uncertainty for retrieved AOD at 0.55 μm (as compared to global
133 network of sun photometers) is $\pm(0.05 + 15\%)$ over land, and $\pm(0.03 + 10\%)$ over ocean [*Levy et*
134 *al.*, 2013].

135 These LUTs are created as if the atmosphere is composed only of aerosol and scattering
136 (Rayleigh) molecules. The gas absorption is assumed to be zero. This is because of the large
137 spatial/seasonal variability of two of the primary absorbers: ozone and water vapor. Ozone can
138 range from 100 to 500 DU around the globe [*Hegglin et al.* 2014] and water vapor varies by an
139 order of magnitude from the wet tropics to the dry poles. It would be cumbersome and

140 computationally inefficient to add two or more new indices to the LUT and cover the dynamic
141 range of each gas in the LUT calculation.

142 While gas absorption in these window bands may be small, they are not zero, as
143 described above. Figure 1 shows the TOA transmission spectra (black lines) in the 0.4 – 2.5 μm
144 spectral range in the presence of major gases, including H_2O , O_3 , CO_2 , CH_4 , O_2 , N_2O , and CO .
145 The transmission spectra of each gas ~~were~~was calculated using the Line-by-Line Radiative
146 Transfer Model (LBLRTM) code [Clough *et al.*, 1992, 2005] for a nadir viewing geometry and
147 for the US 1976 Standard Atmosphere (US76; 1976). A transmittance of 1.0 indicates that the
148 atmosphere is transparent to incoming solar radiation (insolation) i.e. it is not absorbed in the
149 atmosphere. Overlaid on Fig. 1 are the spectral response functions of the seven MODIS channels
150 (blue curves) and seven VIIRS channels (red curves) used in the DT retrievals. As can be seen
151 from Fig. 1, depending on the wavelength, the atmosphere can be totally transparent to a certain
152 gas and partially opaque to another. For example, in the MODIS 0.62-0.67 μm band (B1), H_2O ,
153 O_3 , and O_2 , absorb radiation while CO_2 , N_2O , CO , and CH_4 do not. In the 1.230 – 1.250 μm band
154 (B5), O_2 , H_2O and CO_2 are major absorbers while other gases are not. Absorption bands of the
155 major atmospheric gases are listed in Table 1.

156 Note that there are also wavelength regions that are nearly opaque because of gas
157 absorption. For example, Fig. 1 shows the well-known water vapor absorption within the
158 wavelength region near 1.38 μm . Because of the strong absorption, the 1.38 μm band cannot be
159 used for aerosol retrieval. Yet, this band is very useful for detecting cirrus clouds that would
160 otherwise contaminate a cloud-free aerosol retrieval (Gao *et al.*, 2002). This special case of using
161 absorption information is not discussed further in this paper.

2.2 Derivation of a gas absorption correction

Because the LUT is calculated without gas absorption, an alternative technique must be substituted to account for the effect of the gases in each wavelength. If not, then when the algorithm attempts to match the measured TOA reflectances to the LUT-calculated reflectances the LUT values will be brighter than the measured values for the same amount of aerosol. In the most straightforward sense retrieved AOD, dominated by scattering, will be systematically too low because the retrieval will be searching for a less bright TOA reflectance in the LUT, with less aerosol, to match the observed values. The algorithm deals with this mismatch between measured and LUT reflectance caused by the missing gas absorption in the LUT values by adjusting *the measured* TOA reflectances in each wavelength band, in effect brightening the measurements to better match the values in the LUT.

Figure 1 shows that ~~five-six~~ gases (H₂O, O₃, CO₂, N₂O, O₂ and CH₄) have absorption lines that fall within the wavelength bands used for the DT aerosol retrieval. Because each window band spans tens of nanometers, every DT channel is affected by at least one gas where the transmittance is less than 1.0.

We have introduced two measures, gas opacity and transmissivity corresponding to the gas absorption optical depth and transmittance. The two parameters are related via,

$$T_{\lambda}^i = \exp(-G^i \tau_{\lambda}^i) \quad \dots \dots \dots (1)$$

where T_{λ}^i is the downward transmittance for a particular wavelength band or λ , and for a particular absorbing gas “ i ”, and where τ_{λ}^i is the gas optical depth designated for the particular gas and wavelength and G^i is the airmass factor (or the atmospheric path length i.e. the slant path through the atmosphere) for gas i . Equation (1) shows that transmission of light is a function of

184 the ~~atmospheric path length~~airmass factor (G^i) and the gas optical depth (τ_λ^i), and that
185 transmissivity decreases with increasing air mass and increasing gas concentration.

186

187 **2.2.1 Gas optical depth**

188 The gas optical depth, τ_λ^i , represents the spectral integral over the wavelength band, and
189 if the gas concentration was uniform along the path (column), then τ_λ^i would be directly
190 proportional to the loading of gas i in the column. Some gases are indeed well-mixed in the
191 atmosphere, but water vapor and ozone are not. These important absorbers exhibit distinctive
192 vertical profiles, as will be discussed in Section 3.1. Note that each individual gas has its own
193 particular absorption efficiency based on its characteristic absorption cross section, and that for
194 the same column concentrations τ_λ^i will be different for different gases. In the absence of a long
195 slant path, and for small gas optical depths ($\tau_\lambda^i \ll 1.0$), transmission can be estimated by ~~$T_\lambda^i \sim 1 - G^i \tau_\lambda^i$~~

196 ~~$T_\lambda^i \sim 1 - G^i \tau_\lambda^i$~~

197

198 **2.2.2 Airmass factor**

199 The airmass factor, G can be approximated as $G=1/\cos Z$ where Z is the zenith angle, for a
200 homogenous (exponential decay) atmosphere, and for small values (near nadir) of a zenith angles
201 Z . This is the flat earth approximation. As Z increases beyond 60° , the air mass factor is more
202 accurately described by spherical shell geometry towards the horizon [Gueymard, 1995], i.e.:

$$G = \sqrt{(r \cos Z)^2 + 2r + 1} - r \cos Z \quad \dots \dots (2)$$

203 where, $r = R_E / H_{atm}$; R_E = radius of Earth (6371km) and H_{atm} = effective scale height of the
204 atmosphere (approx 9km). This expression accounts for Earth's sphericity and atmospheric

205 refraction. Differences in computing G are small for $Z < 70^\circ$, but increase to 10% as $Z = 84^\circ$ (the
 206 maximum zenith angle allowed within the DT algorithm).

207 Yet, there are complications. When atmospheric constituents are well-mixed and their
 208 concentrations are nearly proportional to altitude within the atmosphere, Eq (2) is sufficient.
 209 However, water vapor (concentrated near the surface) and ozone (concentrated in the
 210 stratosphere) are not well-mixed in the vertical, having different scale heights. In this layered
 211 situation (rather than continuous), there are empirical formulas (e.g. *Kasten and Young, 1989*)
 212 that provide slight improvements to the calculation of G assuming spherical geometry. For
 213 example, *Gueymard, [1995]* derived the empirical formula

$$G^i = (\cos Z + a_{i,1} Z^{a_{i,2}} * (a_{i,3} - Z)^{a_{i,4}})^{-1} \dots \dots \dots (3)$$

214 where $a_{i,j}$ are the coefficients ($j=1,4$) for gas type i . Thus, G^i varies with gas type and specific
 215 profile within the atmosphere. The values of coefficients $a_{i,j}$ can be found in Table 4.1 of
 216 *Gueymard, [1995]*.

217 As long as the total gas optical depth is small ($\sum_i \tau_\lambda^i \ll 1.0$), the total transmission of all
 218 trace gases is well-approximated by the product of each individual gas (i.e.):

$$T_\lambda^{GAS} = \prod_i T_\lambda^i = \exp\left(\sum_i -G^i \tau_\lambda^i\right) \dots \dots \dots (4a)$$

219 The total gas transmissivity defined in Equation (4a) for each wavelength band quantifies the
 220 degree to which the measured reflectance will be diminished due to gaseous absorption. In order
 221 to match the measured reflectances to those calculated for the LUT, these diminished
 222 reflectances have to be “corrected” or brightened. This correction factor is simply the inverse
 223 transmissivity, \tilde{T} ,

$$\tilde{T}_\lambda^{GAS} = 1/T_\lambda^{GAS} = \exp\left(\sum_i G^i \tau_\lambda^i\right), \quad \dots \dots \dots (4b)$$

224 which when multiplied with the measured reflectance restores the amount of light absorbed by
 225 gases along the one-way path of transmission. Or, given a measured radiance, L_M , the corrected
 226 (brightened) radiance L , is simply, $L = L_M * \tilde{T}$.

227 When observing from a ground-based sun photometer (e.g. AERONET), the correction is
 228 straightforward, because the path of transmission traverses the depth of the atmosphere only
 229 once. The problem is more complicated for satellite remote sensing, because a satellite measures
 230 radiation that has traveled downwards through the atmosphere and then back up to space. We
 231 have to calculate a two-way correction factor and G must account for the Z angles of both
 232 downward (the solar zenith angle) and upward paths (view zenith). As Z gets large, the vertical
 233 profile of the gas (layering) becomes more important.

234 There are two parameters determining the transmission, T_λ^{GAS} , and therefore the
 235 correction factor, \tilde{T}_λ^{GAS} , and these are G^i and τ_λ^i . The goal, then, is to parameterize equation 4(a)
 236 or 4(b) i.e. the relationship between atmospheric transmission of gas and $G^i \tau_\lambda^i$; taking into
 237 consideration the varying gas concentrations and their vertical profiles through the atmosphere,
 238 around the globe. Furthermore, the parameterization will be developed to link \tilde{T}_λ^{GAS} directly to
 239 column measures of the gases instead of to the optical depth. This allows the algorithm to
 240 bypass calculations of optical depth from inputs of precipitable water vapor (w in cm) and ozone
 241 (O in Du), and instead use the inputs directly.

242

243 **3. Use of LBLRTM to derive gas absorption parameterization**

244

245 To develop an empirical relationship between atmospheric gas transmission, the airmass
246 factor (G^{\dagger}) and its optical depth (τ_{λ}^{\dagger}), we require a radiative transfer (RT) code that can
247 accurately simulate the gaseous absorption and transmission process in the atmosphere. Among
248 other things, the RT code requires these two pieces of information: (a) the absorption cross-
249 sections and concentration of gas constituent in spectral bands of interest and (b) accurate high-
250 resolution information of the absorption spectra of the relevant gases. The MODIS and/or VIIRS
251 channels widths are on order of 20-50 nm. We require a high-resolution database to capture the
252 fine absorption lines within these bandwidths. To address (a) and (b), we use the Line-By-Line
253 Radiative Transfer Model (LBLRTM) to parameterize equations 4(a) and 4(b) instead of a
254 MODTRAN based RT code. The following section provides details of LBLRTM.

255

256 **3.1 LBLRTM description**

257 The Line-By-Line Radiative Transfer Model (LBLRTM) is known to be an accurate and
258 flexible radiative transfer model that can be used over the full spectral range from ultraviolet to
259 microwave [Clough et al., 2005]. It uses the High-resolution TRANsmission (HITRAN)
260 molecular absorption database [Rothman et al., 2009] ~~for to calculating~~ calculate transmittance
261 and radiance of molecular species. The HITRAN2008 database contains over 2,713,000 lines for
262 39 different molecules. The spectral resolution of the data is different in different spectral
263 regions and for different species [see Rothman et al., 2009]. For example, for water vapor
264 absorption in the Near IR region, the line resolution is 0.001 cm^{-1} [2.5 – 3.4 μm]. The LBLRTM
265 has been extensively validated against atmospheric radiance spectra [e.g. Turner et al., 2003;
266 Shephard et al., 2009; Alvarado et al., 2013]. Use of the HITRAN database and other attributes
267 of LBLRTM provide spectral radiance calculations with accuracies that are consistent with

268 validation data. Limiting errors are, in general, attributable to line parameters and line shape.
269 Algorithmic accuracy of LBLRTM is approximately 0.5% and is about five times less than the
270 error associated with line parameters [Clough *et al.*, 2005].

271 **3.2 LBLRTM calculations for MODIS and VIIRS**

272 The LBLRTM model was run for many scenarios representing different combinations of
273 gas vertical profiles, gas concentrations and air mass factors for each type of gas and each of the
274 wavelength bands of interest. Transmissions of the ten important atmospheric gases, viz. H₂O,
275 O₃, O₂, N₂O, NO₂, NO, SO₂, CO₂, CO, and CH₄, that affect either the MODIS or the VIIRS
276 spectral bands [Levy *et al.*, 2013] were calculated. However, only H₂O, O₃, CO₂, N₂O, O₂ and
277 CH₄ were found to have some absorption in the wavelength bands used for the DT aerosol
278 retrieval (Tables 2.1, 2.2). The results link transmission, T_{λ}^i , or gas correction factor, \tilde{T}_{λ}^i , to gas
279 path length, $G^H_2O w$ or $G^O_3 O$, for water vapor (H₂O) and ozone (O₃), respectively, where w is the
280 precipitable water vapor in cm and O is ozone column loading in DU. Values for w and O are
281 input into the algorithm from ancillary data. The other gases are considered to be well-mixed and
282 not varying spatially or temporally, and therefore, are not dependent on input ancillary data. The
283 final parameterization will be curve fits through the scatter of the model results.

284 As described in Section 2.2, $\tilde{T}_{\lambda}^{GAS}$ will be affected by the vertical distribution of the gases
285 in the column, especially at oblique zenith angles. To account for this effect in building the
286 parameterization we use 52 atmospheric profiles (personal communication, Pubu Ciren, NOAA,
287 Chevallier, 2002) that were obtained from model runs, and characterize different locations and
288 seasons (Figure 2). The columnar gas concentrations differ across the 52 profiles, varying by
289 more than a factor of ten for water vapor, and by 100% for ozone. Except for NO₂, which is

290 highly variable in both horizontal and vertical, the other trace gases tend to be well-mixed
291 throughout the atmosphere. Using radiative transfer calculations, Ahmad et al., (2007) show that
292 NO₂ has largest impact (1%) on TOA reflectance in the blue channels (412 and 443 nm). Other
293 visible channels are impacted to a lesser degree. We will use the term ‘dry gas’ to denote the
294 eight gases that are neither H₂O or O₃, and use the US 1976 Standard Atmosphere (US 76) as a
295 default profile.

296 For H₂O and O₃, and each of their respective profiles, we use LBLRTM to calculate air
297 mass factors and transmissions for 10 values of viewing zenith angle, ranging from 0° – 80°.
298 Transmission is integrated across the wavelength band, and weighted by relative sensor response
299 (RSR) (Barnes et al., 1998; Xiaoxiong et al., 2005) within the band. Because air mass factor (G^i)
300 varies with gas type (on account of the vertical profile), LBLRTM calculates G^i as well as
301 transmission for the given column amount of gas i . For “dry gas”, the integrated RSR weighted
302 transmission is converted to gas optical depth, so dry gas transmission (as a function of air mass
303 factors) is easily computed using Eq (1). The US 1976 Standard Atmosphere (US 76) profiles are
304 used to compute “dry gas” transmission for nadir view.

305 Figure 3 plots the relationship between absorption correction factor, \tilde{T}_λ^{GAS} , and gas path
306 length, $G^{H_2O}w$, for H₂O (panel a) and, $G^{O_3}O$, for O₃ (panel b), for MODIS. Figure 4 plots the
307 same for VIIRS. These correction factors (inverse of transmission) are plotted for each window
308 band, for different combinations of H₂O or O₃ concentrations (w in cm or O in DU) and
309 internally derived air mass factors (G^i) for the given gas type and specific vertical profile. Water
310 vapor, being so variable as well as concentrated near the boundary layer, cannot be explained
311 with a linear relationship. However, fFor water vapor (panels (a) in both figures 3, 4), a near-

312 linear dependence of $\tilde{T}_\lambda^{H_2O}$ to $G^{H_2O}w$ ~~does exists~~ in log-log space. ~~Water vapor, being so variable~~
 313 ~~as well as concentrated near the boundary layer, cannot be explained with a linear relationship.~~
 314 ~~Even Water vapor, being so variable as well as concentrated near the boundary layer, cannot be~~
 315 ~~explained with a linear relationship. Although, even~~ within the log-log space, there is a small
 316 curvature that ~~requires-required~~ a quadratic ~~for the~~ empirical fit. For ozone, however, the log of
 317 our correction factor ($\tilde{T}_\lambda^{O_3}$) is nearly linear as a function of absorption through a slant path
 318 ($G^{O_3}O$). Again, note that G^i is computed by LBLRTM, and represents the curvature and vertical
 319 profile of each gas type.

320 Equation 5 describes the quadratic empirical relationship (seen in Fig.3a and Fig 4a)
 321 between ~~the~~ gas transmission correction factor of water vapor ($\tilde{T}_\lambda^{H_2O}$), its concentration (w) and
 322 air mass factor (G^{H_2O}):

$$\tilde{T}_\lambda^{H_2O} = \exp(\exp(K_{1,\lambda}^{H_2O} + K_{2,\lambda}^{H_2O} \ln(G^{H_2O}w) + K_{3,\lambda}^{H_2O} (\ln(G^{H_2O}w))^2)) \dots \dots \dots (5)$$

324
 325 and Equation 6 describes the near linear relationship for ozone (panels b in both Fig. 3
 326 and Fig. 4).

$$\tilde{T}_\lambda^{O_3} = \exp(K_{1,\lambda}^{O_3} + K_{2,\lambda}^{O_3}(G^{O_3}O)) \dots \dots \dots (6)$$

328 “O” denotes ozone concentration in Eq. 6 and $G^{O_3}G^i$ is the airmass factor for ~~gas-Ozone~~ i
 329 and is computed using equation 3.

330 The regression coefficients $K_{1,\lambda}^{H_2O}$, $K_{1,\lambda}^{H_2O}$, $K_{1,\lambda}^{H_2O}$ and $K_{1,\lambda}^{O_3}$, $K_{2,\lambda}^{O_3}$ (the slopes and intercepts)
331 for H₂O and O₃ are presented for MODIS and VIIRS in Tables 3.1 and Table 3.2. The slope and
332 intercepts are wavelength dependent (lines of different color on Figs. 3 and 4) and in accordance
333 ~~to~~-with absorption characteristics of the gas. For example Table 2.1 shows that water vapor
334 absorption is the highest in MODIS band 7 (B7 = 2.11 μm) and lowest in B3 (0.47 μm).
335 Correspondingly, the slope and intercept for the H₂O regression relation (Table 3.1) indicates
336 largest water vapor correction in B7 and lowest in B3. Similarly, largest correction (and slope)
337 for ozone is in MODIS B4 (0.55 μm) and lowest in B7.

338 To calculate the correction factors for water vapor ($\tilde{T}_\lambda^{H_2O}$) and ozone ($\tilde{T}_\lambda^{O_3}$), Equations (5)
339 and (6) require information on water vapor (~~ww~~) and ozone concentration (~~oo~~). For the DT
340 algorithm, these are provided by an ancillary data set. For the current version (e.g. MODIS
341 Collection 6), ancillary data are acquired from National Center for Environmental Prediction
342 (NCEP) analysis, specifically the “PWAT” and the ozone fields from the 1° X 1° global
343 meteorological analysis (created every six hours – format “gdas.PGrbF00.YYMMDD.HHz”).
344 Note that there are water vapor products derived operationally from MODIS and VIIRS data
345 (e.g. Gao and Goetz, 1990; Kaufman and Gao, 1992). However, the DT aerosol algorithm runs
346 before these other algorithms in the processing chain, causing the internally-derived water vapor
347 to be unavailable to the aerosol algorithm in real-time processing and thus, the reliance on
348 ancillary data.

349 In case the ancillary information is not available, the gas absorption can still be estimated.
350 Either a forecast field (e.g. GDAS forecast) or a “climatology” can be used. For example, if the
351 US76 atmosphere is assumed as the climatology for gas profiles, then τ^i for that gas is given in

352 Table 3.1 and 3.2. In this case, we use Equations (7) and (8) to calculate correction factors for
 353 water vapor and ozone respectively:

$$\tilde{T}_\lambda^{H_2O} = \exp(G^{H_2O} \overline{\tau^{H_2O}}) \quad \dots \dots \dots (7)$$

$$\tilde{T}_\lambda^{O_3} = \exp(G^{O_3} \overline{\tau^{O_3}}) \quad \dots \dots \dots (8)$$

354 where $\overline{\tau^{H_2O}}$ and $\overline{\tau^{O_3}}$ are the climatological mean values of gas optical depth for water vapor
 355 and ozone, respectively.

356 $\tilde{T}_\lambda^{Dry Gas}$ is the correction factor due to dry gas, which includes CO₂, CO, N₂O, NO₂, NO,
 357 CH₄, O₂, SO₂, and other trace gases in LBLRTM calculations. For the DT retrieval bands only
 358 CO₂, N₂O, CH₄, O₂ contribute to absorption (Tables 2.1, 2.2). Since the gases are generally well-
 359 mixed throughout the entire atmosphere and do not experience day-to-day changes, we only
 360 consider the climatological mean of the total optical depth of the combined dry gases, and
 361 compute its transmittance factor as follows:

$$\tilde{T}_\lambda^{Dry Gas} = \exp(G^i \overline{\tau^{Dry Gas}}) \quad \dots \dots \dots (9)$$

362
 363 Fig. 5 presents the gas optical depth for a US76 atmosphere, for the MODIS bands and
 364 corresponding VIIRS bands. In some cases, (e.g. B4 vs. M5) the differences are small. In other
 365 cases (e.g. B5 vs. M8), the total optical depth may be similar, but the relative contribution
 366 between different gases is different. Finally, in at least one set of bands (B7 vs. M11), both the
 367 total optical depth and the relative contributions between gases is very different. The US76 is a
 368 case of-with a small amount of water vapor ($w=1.4$ cm), but one can see how quadrupling the w
 369 (e.g. as in a tropical atmosphere) would greatly change the relative correction needed for B7 vs.
 370 M11, or even B1 vs. M5.

371

372 **3.3 Application within the DT algorithm.**

373 Whether using “climatology” for water vapor and ozone columns, or using the estimates
374 from a meteorological assimilation system (e.g. GDAS for the current DT algorithm), we need to
375 correct for the combined absorption of all gases. The total gas absorption correction term, \tilde{T}_λ^{gas} ,
376 is the product of individual gas corrections, that is

$$\tilde{T}_\lambda^{gas} = \tilde{T}_\lambda^{H_2O} \tilde{T}_\lambda^{O_3} \tilde{T}_\lambda^{Dry\ Gas} \dots \dots \dots (10)$$

377 The MODIS DT aerosol retrieval algorithm ingests calibrated and geolocated MODIS-
378 measured reflectance data, known as the Level 1B (L1B) product. The corresponding VIIRS DT
379 algorithm ingests a similar VIIRS-measured product. This measured reflectance, (ρ_λ^{L1B}), is
380 corrected for atmospheric water vapor, ozone and dry gas, using the correction factors derived
381 above for each wavelength band:

$$\rho_\lambda = \tilde{T}_\lambda^{gas} \rho_\lambda^{L1B} \dots \dots \dots (11)$$

382 where ρ_λ is the corrected or brightened reflectance that ~~will-can~~ now be used to match-compare
383 with the calculated TOA reflectances of the LUT, as described in Section 2.2. Note that this
384 spectral reflectance ρ_λ , represents the combination of Rayleigh (molecular scattering), plus
385 aerosol in the atmosphere. It also includes contributions from Earth’s surface (land or water).

386
387 The gas-absorption correction methodology is the same whether performed for MODIS
388 or VIIRS. In fact, the equations (Eqs 5-11) have remained the same throughout all versions of
389 the DT algorithm. As our ability to characterize absorption lines as well as the spectral response
390 of the sensor has improved, it is the coefficients of the equations that have evolved. When the
391 DT algorithm was updated from Collection 5 (C5) to Collection 6 (C6), the underlying gas

392 absorption corrections became more sophisticated (Levy et al., 2013). This is represented in
393 Table 4. The primary differences between C5 and C6 are that [the](#) HITRAN database in
394 LBLRTM is used in C6 instead of the MODTRAN parameterization available in 6S that was
395 used in C5, and that additional “dry” gases have been included in C6’s correction. These changes
396 made a difference. The latest version of aerosol data from DT is Collection 6.1 that uses the
397 same gas absorption corrections as C6. As the DT algorithm is ported from MODIS to VIIRS
398 data, the quality of gas correction will also make a difference.

399 **4 Impact of new gas coefficients**

400 The DT retrieval is based on a LUT approach wherein the measured and modeled spectral
401 reflectance is matched for inversion. Any change affecting the calculation of gas-corrected
402 spectral reflectance will subsequently affect the retrieved AOD. *Levy et al.*, [2013] showed the
403 impact of using the updated atmospheric [gas](#) corrections on MODIS C6 AOD retrievals. This led
404 to higher AODs globally. Over land (ocean), the 0.55 μm global mean AOD differed by ~ 0.02
405 (0.007). The large (>0.02 regionally) change over land was primarily due to a larger gas
406 correction in the 1.24 μm MODIS B5 band (see *Levy et al.*, 2013; Fig. A2), which in turn
407 increased the reflectance in B5, and [the](#) subsequent estimate of the NDVI in the SWIR channels
408 (B5 vs. B7) used to estimate surface reflectance in other bands (Levy et al., 2010). The stronger
409 gas correction in B5 came from including the O₂ absorption, which had not been accounted for in
410 C5 (see Table 2.1). Interestingly, Levy et al. [2013] noted that while the overall correction in B7
411 (2.11 μm) remained similar, the relative weightings of “dry gas” and H₂O was revised.

412 Even though [the](#) MODIS and VIIRS instruments have similar channels, the MODIS gas
413 correction coefficients cannot be applied to aerosol retrievals from VIIRS observations. The

414 slight differences in the bandwidth and channel's central wavelengths (See Fig. 5) will
415 compromise the accuracy of aerosol retrievals. For example, as compared with MODIS B7 (2.11
416 μm), the VIIRS M11 (2.25 μm) band has less absorption from H_2O . However, MODIS B7 lies in
417 a CO_2 absorption band, while VIIRS M11 lies in a region of CH_4 absorption. Although the CH_4
418 optical depth in VIIRS M11 is small (~ 0.03), it will affect the dark-target retrievals in the same
419 way as O_2 inclusion affected C6 retrievals (when compared to C5).

420 As a perturbation experiment we intentionally apply the MODIS gas corrections to the
421 VIIRS observations, even though we know this to be incorrect. Figure 6a plots the spatial
422 distribution of spectral TOA reflectance after applying VIIRS-appropriate gas corrections. It
423 shows the mean monthly TOA reflectance for VIIRS. Figure 6b are the reflectance differences
424 between applying VIIRS-appropriate gas corrections and MODIS gas corrections to VIIRS
425 observations. From top to bottom, we find a mean difference of 0%, -0.5%, -6.6%, -2.7%, -
426 1.5%, 3.2% and 5.3% respectively in VIIRS channels M3, M4, M5, M7, M8, M10, M11.
427 Looking back at Fig. 5, one can see that ~~that~~ for example, by using proper M5 assumptions
428 instead of the B1 MODIS assumptions, we now apply only about half the correction as before,
429 resulting in a 6.6% reduction of reflectance. Channel M7, with about 50% less water vapor
430 correction (see Fig. 5), results in 2.7% lower reflectance. Larger gas corrections owing to CO_2
431 absorption in M10 and CH_4 absorption in M11 (Fig. 5), result in positive bias in M10 and M11
432 reflectance values globally.

433 Now, we continue the perturbation experiment and test the impact of slight differences in
434 the band positioning between MODIS and VIIRS on AOD retrieval by performing two sets of
435 retrievals. The first set (a) is if we applied appropriate VIIRS band corrections, while the second
436 (b) is as if we had simply (naively) applied MODIS (C6) coefficients to VIIRS data. Figure 7

437 shows the AOD retrieved from these two cases (panels a and b) for an entire month (July 2013)
438 of VIIRS data. While general AOD spatial patterns are in agreement, panel (c) shows differences
439 in AOD of up to 0.07 between the two retrievals. Clearly, naively applied MODIS gas
440 corrections to VIIRS data, would lead to a global mean AOD underestimate of ~ 0.01 for July
441 2013. While these differences are within the global uncertainties for AOD (e.g. GCOS), the
442 regional differences can be much larger.

443 Although once considered to be trivial in magnitude, accurate atmospheric gas
444 corrections have become more important as we strive towards better accuracies in AOD products
445 and towards a seamless climate data record. It is noteworthy that the gas absorption spectra of
446 Figure 1 have been updated several times in recent years [Alvarado et al. 2012] as the scientific
447 community continues to engage in study of gas absorption lines with improved instrumentation
448 and gas spectroscopic measurements. Changing gas absorption spectra will affect the channels
449 designed for new remote sensing instruments and in understanding how these lines might affect
450 the retrieval of proposed geo-physical products. Every instrument design involves
451 characterization of channel bandwidths and the spectral response functions of the instrument's
452 channels. This aptly calls for updates in modeling the absorption by gases in the channels used
453 for aerosol retrievals. For the MODIS Collection 6 AOD product, the team switched from using
454 a MODTRAN gas spectroscopic database to the HITRAN spectroscopic database and found
455 differences.

456 **5. Summary and Conclusions**

457 Performing aerosol optical depth retrieval, from satellite measurements, requires
458 extracting the aerosol signal from the total radiance measured by the sensor at the top-of-

459 atmosphere. The total radiance includes signal from the underlying surface and from atmospheric
460 constituents such as gases, clouds and aerosols. In this paper, we have described the physics and
461 methodology employed by the Dark-Target aerosol retrieval algorithm for atmospheric gas
462 correction of the cloud-free radiance measurements from the MODIS and VIIRS sensors. We
463 have shown that the empirical correction applied to one sensor (MODIS) cannot be applied to
464 another sensor (VIIRS) even when the channels of the two sensors may be similar. For a specific
465 month of VIIRS observations (July, 2013), not accounting for the sensor's bandwidth and
466 positioning of its central wavelength in the electro-magnetic spectrum, can result in an AOD
467 retrieval bias of about 0.01 (global average) and up to 0.07 at regional scales.

468 Water vapor, ozone and carbon dioxide are the major absorbers of solar radiation.
469 Historically, they have been accounted for in atmospheric gas corrections by aerosol retrieval
470 algorithms. However, until recently, standard routine algorithms (e.g. the DT algorithm used on
471 MODIS) did not consider other gases. For example, oxygen with a gas optical depth of about
472 0.016 is important in the MODIS Band 5 (1.24 μm) [Levy et al., 2013]. Methane is an important
473 absorber in band M11 (2.25 μm) of VIIRS with an optical depth of ~ 0.05 . Starting with MODIS
474 Collection 6, and the DT algorithm ported to VIIRS, ~~seven-three~~ additional atmospheric gases
475 [~~CO~~, N₂O, ~~NO₂~~, ~~NO~~, CH₄, O₂, SO₂] are addressed by the gas correction in these DT algorithms.

476 For the 'dry gas' component, the DT gas correction assumes a homogeneous global
477 distribution spatially and a US76 type of vertical distribution for the eight gases. Carbon dioxide,
478 oxygen, nitrous oxide and methane are major absorbers in our 'dry gas' category. Except for
479 NO₂, which is highly variable in both horizontal and vertical, the other gases tend to be well-
480 mixed throughout the atmosphere. Spatial variability of well-mixed gases is typically around
481 10%, mostly latitudinal and is smaller than seasonal variability (e.g. see methane maps here:

482 <http://www.temis.nl/climate/methane.html>). For the nadir view, 10% error due to spatial
483 variability will only introduce an error of 0.005 in the methane correction (optical depth ~ 0.05 in
484 VIIRS channel M11). For now, this is a small uncertainty in the overall retrieval. However, as
485 requirements for aerosol retrieval accuracies tighten, even these well-mixed dry gases will
486 require removal of any seasonal and regional biases by using ancillary measurements of these
487 gases or at least seasonal global climatology of gas optical depths, instead of a single
488 climatological value for the entire globe.

489 Since the DT algorithm corrects for H₂O and O₃ using ancillary data at every 1° X 1° grid
490 box, spatial and seasonal variability of these gases is being accounted for. However, the ancillary
491 data has its own uncertainties that propagate into the gas correction and aerosol retrieval. The
492 Dark-Target team is working towards estimating the error in per-pixel AOD retrievals introduced
493 from several error sources including the errors in H₂O and O₃ ancillary data (GDAS) used for
494 atmospheric gas corrections. Preliminary analysis suggests (not shown here) that gas corrections
495 errors, stemming from considering 20% errors in ancillary data, are much smaller (more than an
496 order of magnitude) than errors from surface albedo uncertainty, aerosol model selection, spatial
497 heterogeneity in a scene, calibration and cloud contamination errors. This is work in progress and
498 subject to future publication.

499 The VIIRS instrument onboard Suomi-NPP is a follow-on of the MODIS instrument on
500 Terra and Aqua satellites. While the Dark-Target team strives to create a seamless climate data
501 record (CDR) of AOD from MODIS and VIIRS, it requires a consistency in AOD retrieval of
502 about 0.02. Any compromise with the accuracy of AOD retrieved from either sensor will impact
503 the CDR consistency requirement. To strive toward these requirements, we cannot ignore quality

504 atmospheric gas corrections in AOD retrievals and we will update the gas correction factors for
505 each instrument as the community updates the gas absorption database.

506 As we move into an era of new aerosol missions, revisiting and updating atmospheric-gas
507 corrections in state-of-art algorithm becomes as important as improving upon other factors (e.g.
508 better surface characterization, cloud clearing, aerosol properties etc.) that affect the AOD
509 retrieval. The dark-target algorithm software has now been generalized to retrieve AOD from
510 sensors other than MODIS and VIIRS. It will be necessary to accurately characterize gases from
511 such current and future instruments as Himawari, GOES-R, etc.

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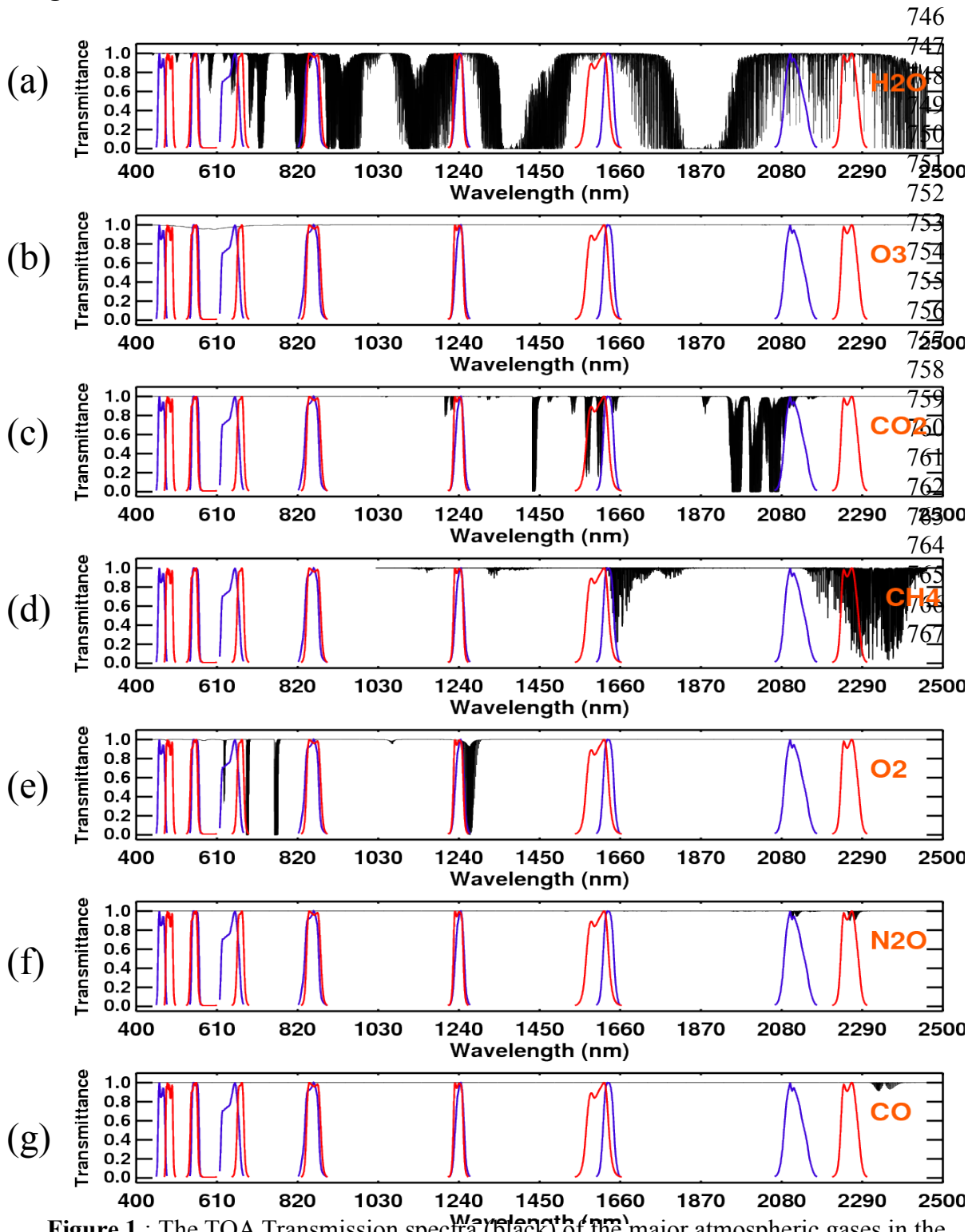


Figure 1 : The TOA Transmission spectra (black) of the major atmospheric gases in the Visible and Near Infrared part of electromagnetic spectrum (400 – 2500 nm). The Line-by-line radiative transfer model (LBLRTM) was used to calculate these gas spectra for a nadir viewing geometry and the 1976 US Standard atmosphere. The spectral response functions of MODIS channels B1-B7 (blue curves) and seven VIIRS channels (red curves) are overlaid to visualize their position in the atmospheric ‘window’ regions where gas absorption effect is minimal

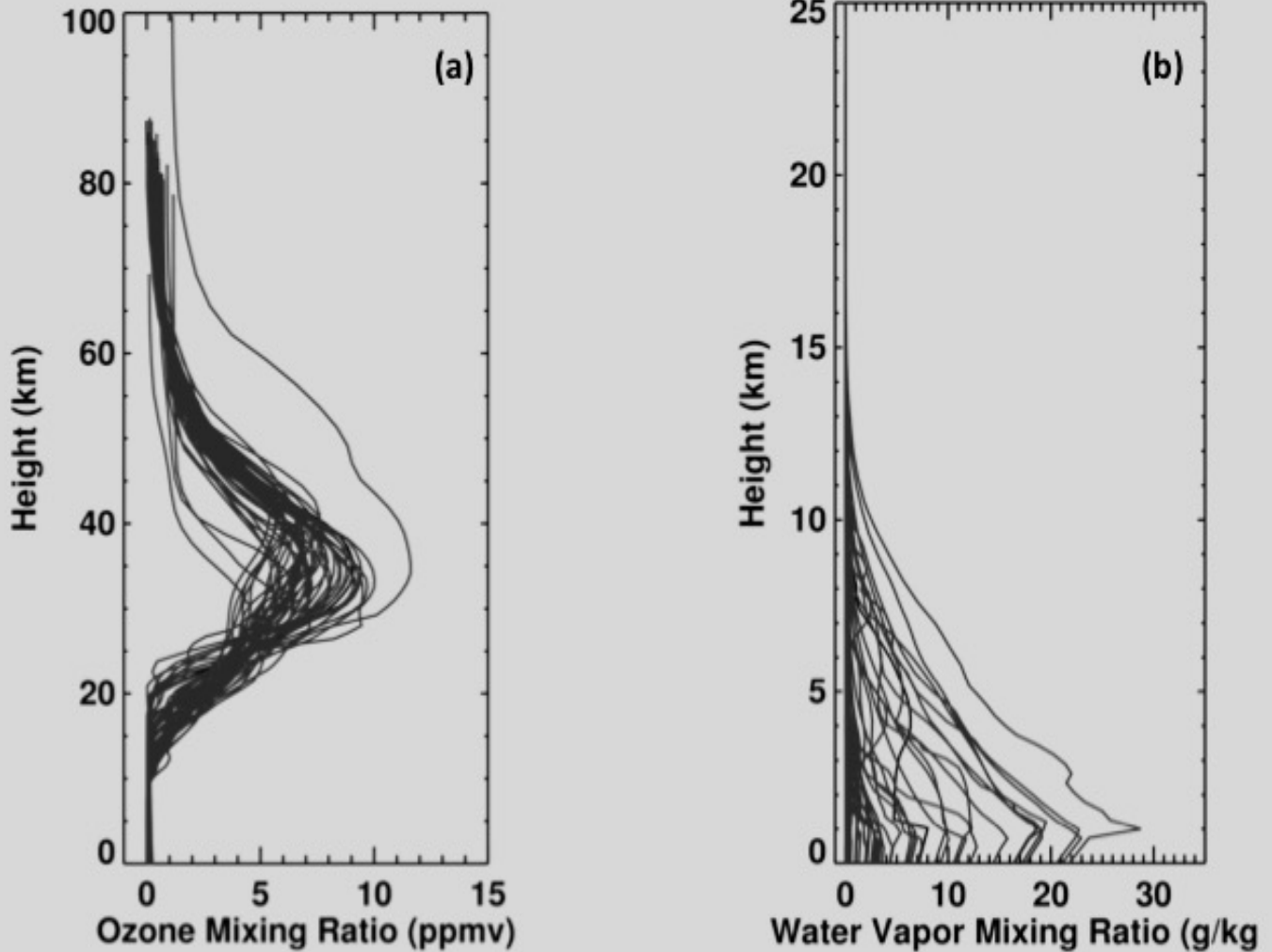


Figure 2: 52 different ECMWF profiles for (a) water vapor and (b) ozone used in the Line-by-line radiative transfer model to calculate the respective gas transmittance.

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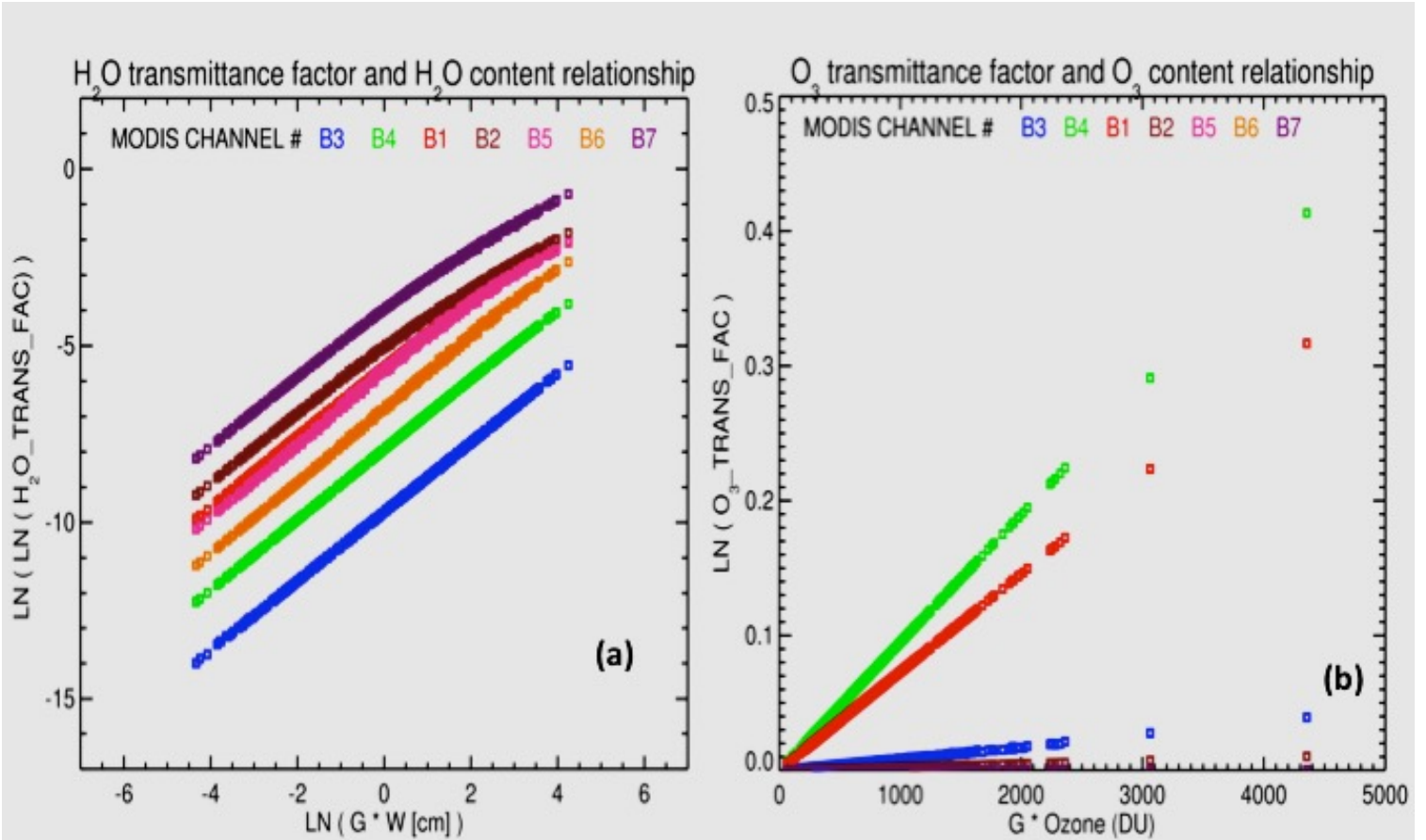


Figure 3: Relationship between Gas Transmittance factor and Gas Content in the MODIS channels B1 – B7 : (a) For H₂O and (b) for O₃. Gas content is scaled by the airmass factor [G]

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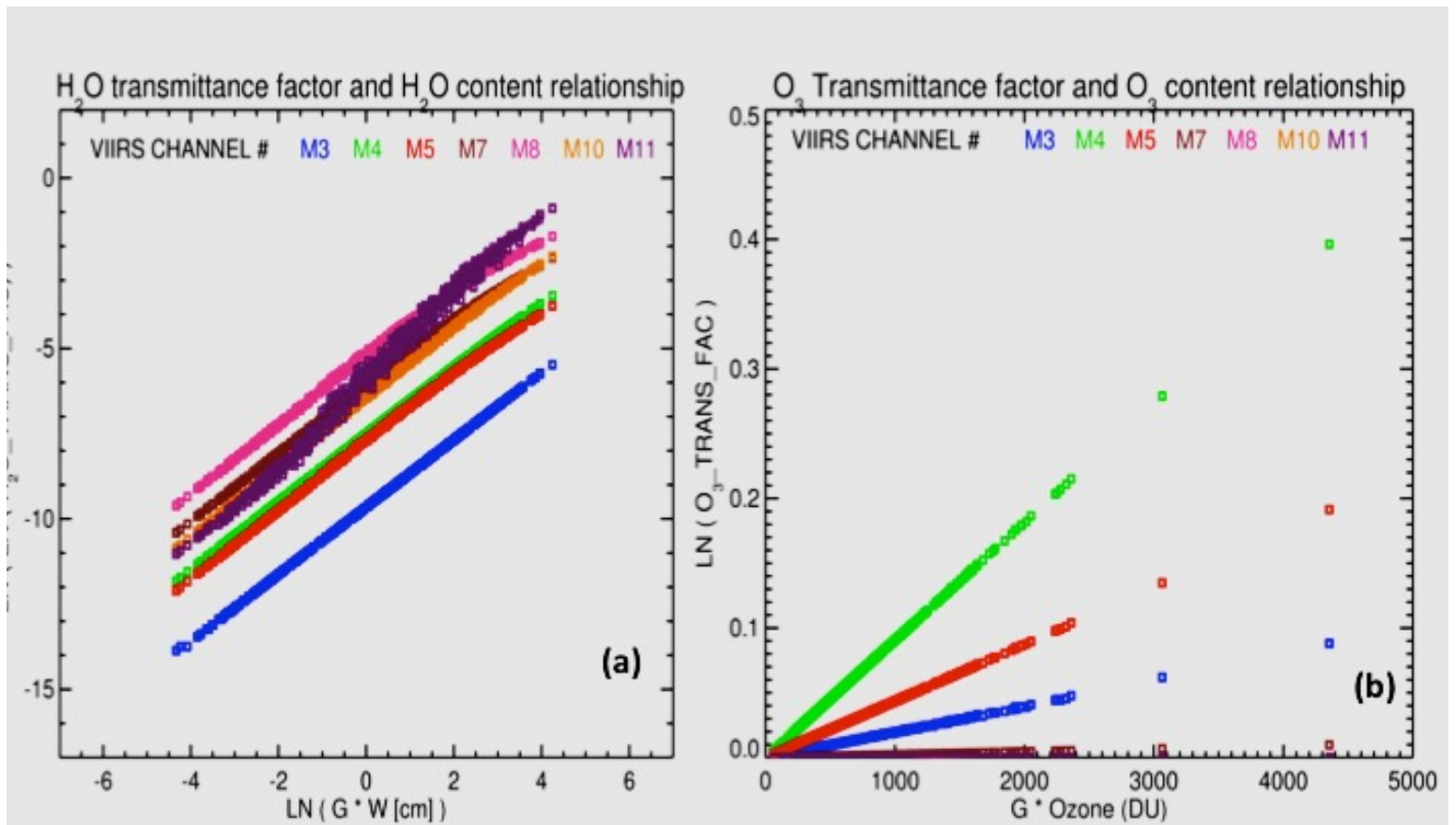
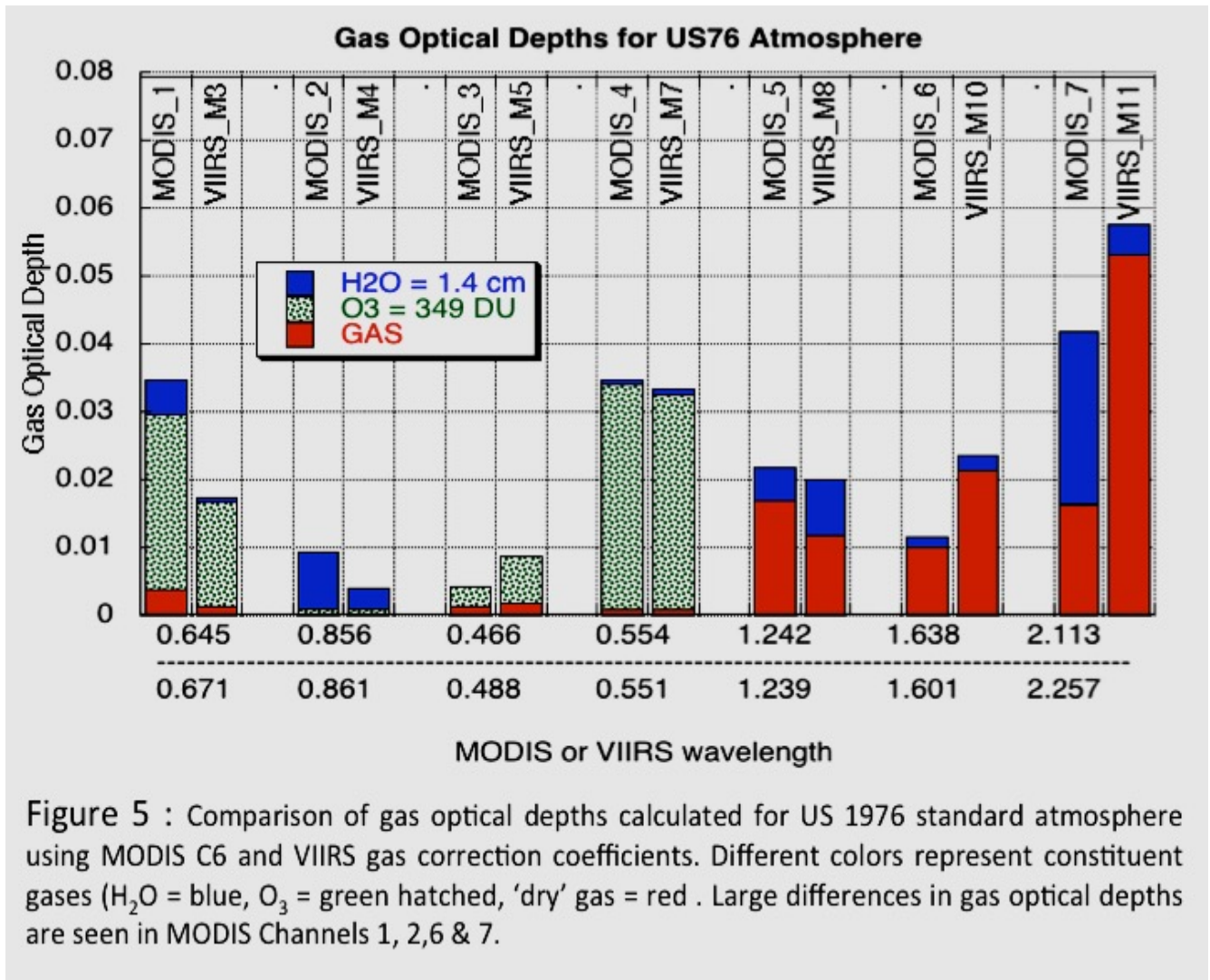


Figure 4: Relationship between Gas Transmittance factor and Gas Content in the seven VIIRS channels: (a) For H₂O and (b) for O₃ Gas content is scaled by the airmass factor [G]

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VIIRS Reflectance

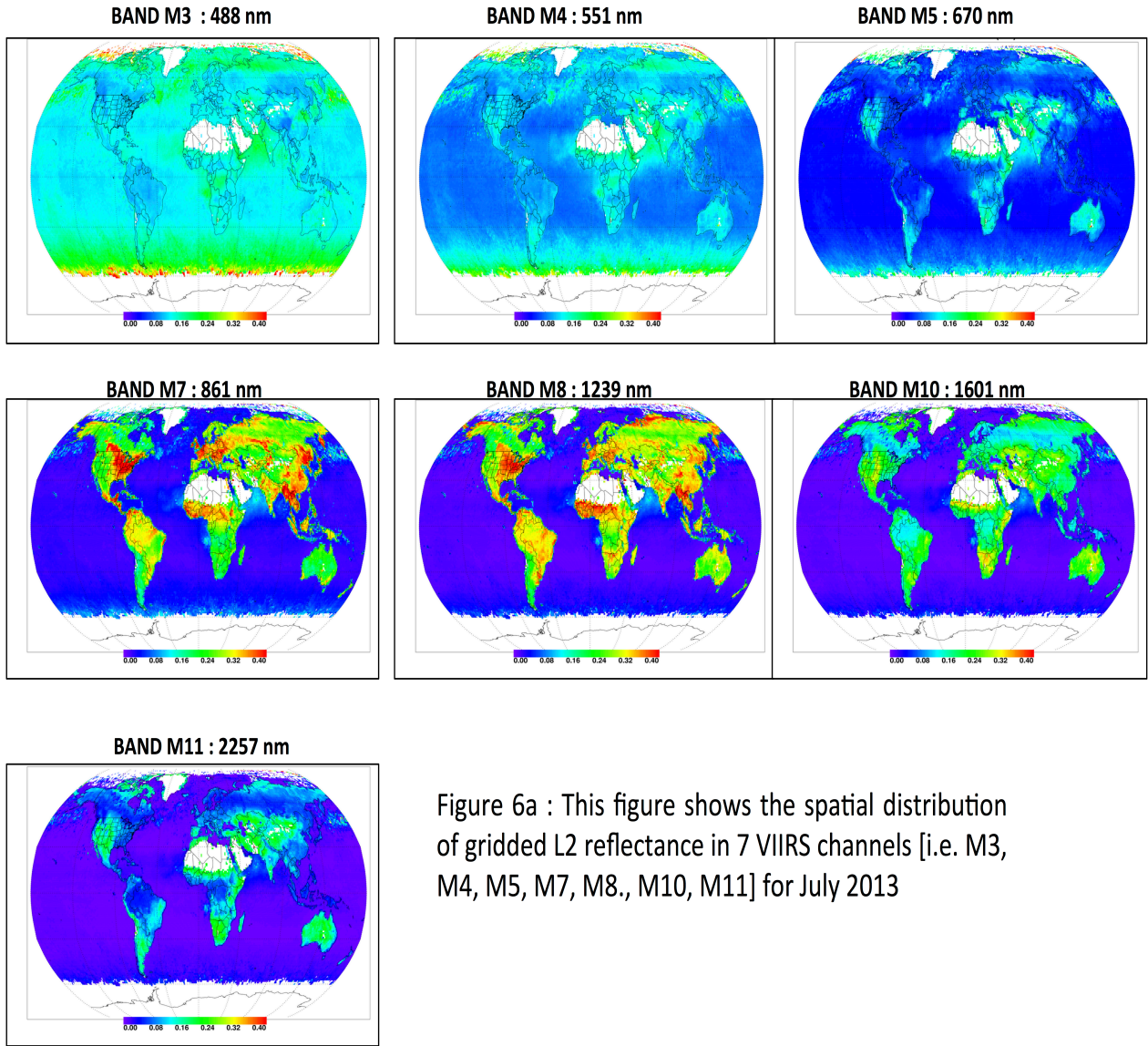


Figure 6a : This figure shows the spatial distribution of gridded L2 reflectance in 7 VIIRS channels [i.e. M3, M4, M5, M7, M8., M10, M11] for July 2013

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Difference between VIIRS – C6 Gas Reflectance

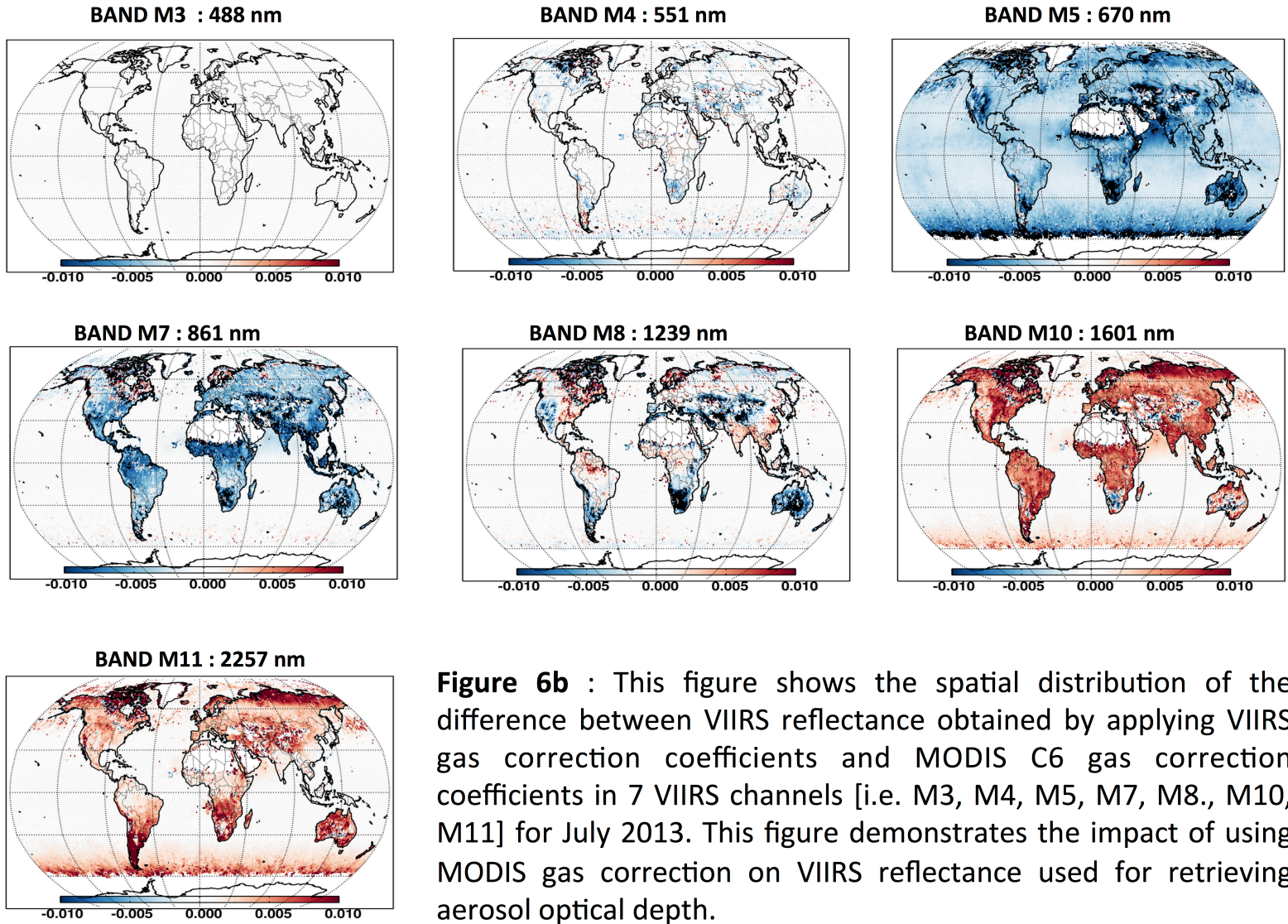


Figure 6b : This figure shows the spatial distribution of the difference between VIIRS reflectance obtained by applying VIIRS gas correction coefficients and MODIS C6 gas correction coefficients in 7 VIIRS channels [i.e. M3, M4, M5, M7, M8., M10, M11] for July 2013. This figure demonstrates the impact of using MODIS gas correction on VIIRS reflectance used for retrieving aerosol optical depth.

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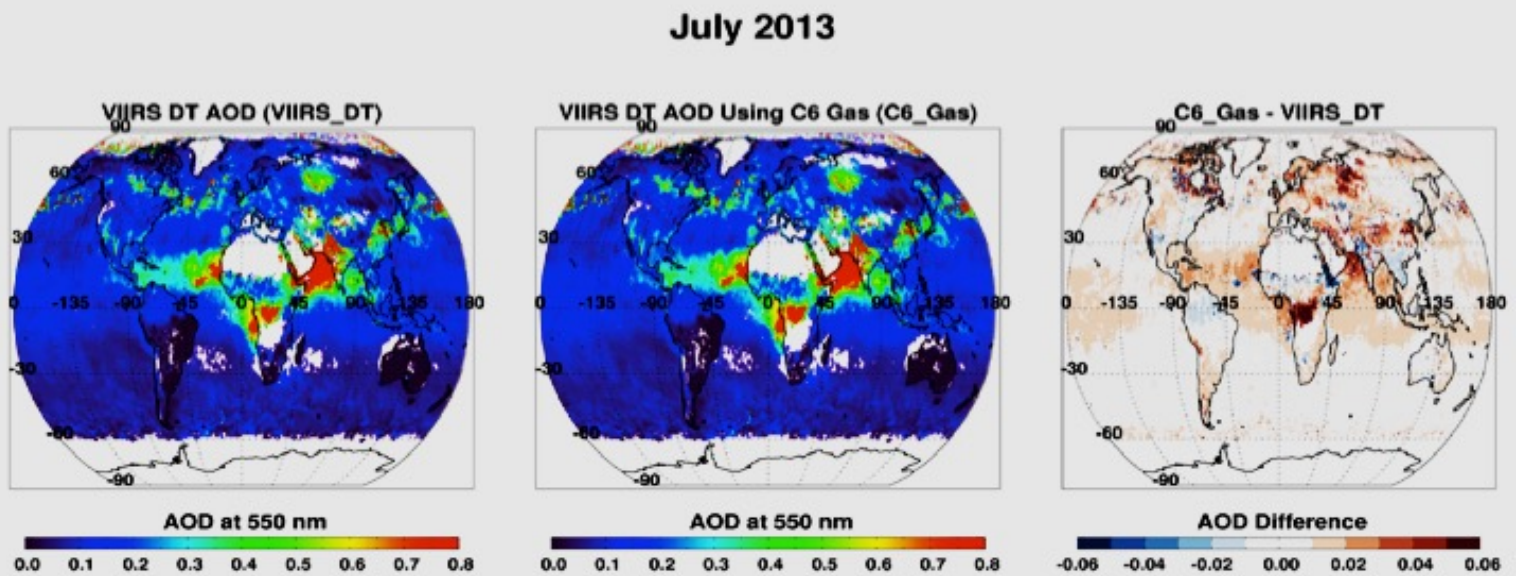


Figure 7: Impact of updated atmospheric corrections on VIIRS AOD (550 nm) retrieval. All things being equal, using C6 aerosol DT retrieval algorithm (a) is AOD using atmospheric coefficients calculated for VIIRS bands and (b) is AOD using C6 atmospheric corrections (c) is the difference between (b) and (a). The global mean AOD differs by ~ 0.012 over land and by ~ 0.004 over ocean. Differences are larger than these mean values regionally but < 0.08 . Differences are mostly positive (reds) except in some desert / bright regions where some negative differences appear.

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Table 1 : Absorption bands of atmospheric gases in visible and near-IR region

Major Atmospheric Gas	Center Wavelengths (μm)
H₂O	visible, 0.72, 0.82, 0.94, 1.1 1.38, 1.87, 2.7
CO₂	1.4, 1.6, 2.0, 2.7, 4.3
O₃	visible (0.45 - 0.75)
O₂	0.63, 0.69, 0.76, 1.06, 1.27, 1.58
N₂O	2.87, 4.06, 4.5
CH₄	1.66, 2.2, 3.3
CO	2.34, 4.67
NO₂	visible

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Table 2.1 : Optical depth of major atmospheric gases in 7 MODIS channels.

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Channel	B3	B4	B1	B2	B5	B6	B7
Wavelength (μm)	0.466	0.553	0.645	0.856	1.242	1.638	2.113
Gas							
H2O	0.0001	0.0005	0.0055	0.0086	0.005	0.0017	0.0254
O3	0.0029	0.0326	0.0250	0.0008	-	-	0.0000
CO2	-	-	-	-	0.0003	0.0050	0.0142
N2O	-	-	-	-	-	-	0.0020
CO	-	-	-	-	-	-	-
O2	0.0012	0.0010	0.0038	0.0000	0.0164	-	-
NO	-	-	-	-	-	-	-
SO2	-	-	-	-	-	-	-
NO2	-	-	-	-	-	-	-
CH4	-	-	-	-	0.0000	0.0051	0.0003
Total	0.0042	0.0341	0.0344	0.0094	0.0216	0.0118	0.0420

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Highlighted boxes show channels where total gas optical depth ≥ 0.02 to put in context the requirement of aerosol optical depth accuracy of better than 0.02

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Table 2.2 : Optical depth of major atmospheric gases in 7 VIIRS channels

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Channel Gas	M3	M4	M5	M7	M8	M10	M11
Wavelength (µm)	0.488	0.551	0.67	0.861	1.239	1.601	2.257
Gas							
H2O	0.00009	0.00078	0.00066	0.00324	0.00844	0.00234	0.00542
O3	0.00673	0.0312	0.01499	0.00075	0	0	0
CO2	0	0	0	0	0.00041	0.02048	0.00001
N2O	0	0	0	0	0	0.00001	0.00403
CO	0	0	0	0	0	0	0
O2	0.00184	0.00084	0.00144	0.00002	0.01147	0	0
NO	0	0	0	0	0	0	0
SO2	0	0	0	0	0	0	0
NO2	0	0	0	0	0	0	0
CH4	0	0	0	0	0.00001	0.00085	0.04914
Total	0.00866	0.03282	0.01709	0.00401	0.02033	0.02368	0.0586

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Highlighted boxes show channels where total gas optical depth ≥ 0.02 to put in context the requirement of aerosol optical depth accuracy of better than 0.02

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Table 3.1: Gas Absorption Coefficients and Climatology for MODIS

MODIS Band	Wavelength (μm)	Rayleigh Optical Depth	O ₃ Optical Depth [#]	H ₂ O Optical Depth [#]	Dry Gas* Optical Depth [#]	O ₃ _K0	O ₃ _K1	H ₂ O_K0	H ₂ O_K1	H ₂ O_K2
B3	0.4659	1.92E-01	2.90E-03	8.00E-05	1.25E-03	-1.14E-04	8.69E-06	-9.58E+00	1.23E+00	-1.16E-01
B4	0.5537	9.44E-02	3.26E-02	5.00E-04	9.50E-04	5.18E-06	9.50E-05	-7.91E+00	1.00E+00	-1.29E-02
B1	0.6456	5.08E-02	2.52E-02	5.11E-03	3.91E-03	1.16E-04	7.32E-05	-5.60E+00	9.40E-01	-1.78E-02
B2	0.8564	1.62E-02	8.10E-04	8.61E-03	2.00E-05	2.80E-07	2.36E-06	-5.07E+00	8.77E-01	-2.40E-02
B5	1.2417	3.61E-03	0.00E+00	5.23E-03	1.69E-02	1.19E-07	1.55E-25	-5.65E+00	9.81E-01	-2.38E-02
B6	1.6285	1.22E-03	0.00E+00	1.62E-03	9.98E-03	1.19E-07	5.17E-26	-6.80E+00	1.03E+00	-4.29E-03
B7	2.1134	4.30E-04	2.00E-05	2.53E-02	1.63E-02	6.29E-07	7.03E-08	-3.98E+00	8.86E-01	-2.56E-02

* Dry Gas includes CO₂, CO, N₂O, NO₂, NO, CH₄, O₂, SO₂

For each MODIS band, this nadir looking (viewing zenith angle = 0) optical depth for the gas is computed from the US 1976 Standard Atmosphere in LBLRTM.

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Table 3.2: Gas Absorption Coefficients and Climatology for VIIRS

VIIRS Band	Wavelength (μm)	Rayleigh Optical Depth	O ₃ Optical Depth [#]	H ₂ O Optical Depth [#]	Dry Gas [*] Optical Depth [#]	O ₃ _K0	O ₃ _K1	H ₂ O_K0	H ₂ O_K1	H ₂ O_K2
M3	0.488	1.60E-01	6.73E-03	8.94E-05	1.84E-03	-1.25E-04	1.98E-05	-9.65E+00	9.87E-01	1.80E-04
M4	0.5511	9.76E-02	3.11E-02	7.69E-04	8.34E-04	-4.75E-05	9.08E-05	-7.50E+00	9.84E-01	-3.87E-03
M5	0.6704	4.40E-02	1.50E-02	6.64E-04	1.44E-03	-4.79E-05	4.37E-05	-7.69E+00	9.95E-01	-1.10E-02
M7	0.8612	1.60E-02	7.70E-04	3.37E-03	2.45E-05	4.18E-07	2.24E-06	-6.05E+00	9.65E-01	-1.53E-02
M8	1.2389	3.67E-03	0.00E+00	8.44E-03	1.19E-02	1.19E-07	5.17E-26	-5.16E+00	9.59E-01	-2.67E-02
M10	1.6012	1.32E-03	0.00E+00	2.34E-03	2.13E-02	1.19E-07	1.03E-25	-6.43E+00	1.02E+00	-3.60E-03
M11	2.257	3.50E-04	1.07E-06	5.42E-03	5.32E-02	-2.61E-08	3.28E-09	-5.85E+00	1.28E+00	-5.04E-03

* Dry Gas includes CO₂, CO, N₂O, NO₂, NO, CH₄, O₂, SO₂

For each VIIRS band, this nadir looking (viewing zenith angle = 0) optical depth for the gas is computed from the US 1976 Standard Atmosphere in LBLRTM.

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Table 4: Atmosphere Gas Correction Table Differences : C5 vs C6

	C5	C6	Comment
RT Code	6s	LBLRTM (Line-by-Line Radiative Transfer Model)	6S is MODTRAN (Ref) database LBLRTM is HITRAN (Ref) database
# Gases Considered	3 [H ₂ O, O ₃ , CO ₂]	10 [H ₂ O, O ₃ , O ₂ , CO, CO ₂ , CH ₄ , NO, N ₂ O, NO ₂ , SO ₂]	Inclusion of 'other' dry gases in C6 created big differences in MODIS bands 5 & 7 (See Fig. 2)
Climatological GODs	Mid-latitude- Summer	US76 Standard Atmosphere	Ref

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