



ILMATIETEEN LAITOS
METEOROLOGISKA INSTITUTET
FINNISH METEOROLOGICAL INSTITUTE

**Information content
analysis: Combination of
satellite and ground-
based observations
enables more accurate
aerosol SSA retrievals at
low aerosol loadings**

Antti Lipponen and Antti Arola





ILMATIETEEN LAITOS
METEOROLOGISKA INSTITUTET
FINNISH METEOROLOGICAL INSTITUTE

x: meeting was in Barcelona



x: meeting was in Barcelona

Helsinki 50%

Barcelona 50%



x: meeting was in Barcelona
y: it was afternoon

Helsinki 50%

Barcelona 50%



x: meeting was in Barcelona
y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem $p(\mathbf{x}|\mathbf{y}) = \frac{p(\mathbf{y}|\mathbf{x}) p(\mathbf{x})}{p(\mathbf{y})}$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL \& afternoon}) + p(\text{BCN \& afternoon})}$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem

$$p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{afternoon}) + p(\text{BCN} \ \& \ \text{afternoon})}$$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL \& afternoon}) + p(\text{BCN \& afternoon})}$

$$= \frac{p(y|x) p(x)}{p(\text{HEL \& afternoon}) + p(\text{BCN \& afternoon})}$$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL \& afternoon}) + p(\text{BCN \& afternoon})}$

$$= \frac{0.6 * 0.5}{0.5 * 0.6 + 0.5 * 0.6} =$$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning	40%
Afternoon	60%

Morning	40%
Afternoon	60%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL \& afternoon}) + p(\text{BCN \& afternoon})}$

$$= \frac{0.6 * 0.5}{0.5 * 0.6 + 0.5 * 0.6} = 0.5$$

x: meeting was in Barcelona

y: it was afternoon

Helsinki 50%

Barcelona 50%

Morning 40%
Afternoon 60%

Morning 40%
Afternoon 60%

Bayes' theorem

$$p(\mathbf{x}|\mathbf{y}) = \frac{p(\mathbf{y}|\mathbf{x}) p(\mathbf{x})}{p(\mathbf{y})}$$

\mathbf{x} : meeting was in Barcelona



Bayes' theorem

$$p(\mathbf{x}|\mathbf{y}) = \frac{p(\mathbf{y}|\mathbf{x}) p(\mathbf{x})}{p(\mathbf{y})}$$

x: meeting was in Barcelona

y: it was sunny



Bayes' theorem
$$p(x|y) = \frac{p(y|x) p(x)}{p(y)}$$

x: meeting was in Barcelona

y: it was sunny

Helsinki 50%

Barcelona 50%

Sunny	10%
Rainy	90%

Sunny	95%
Rainy	5%

Bayes' theorem

$$p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{sunny}) + p(\text{BCN} \ \& \ \text{sunny})}$$

x: meeting was in Barcelona
y: it was sunny

Helsinki 50%
Barcelona 50%

Sunny	10%
Rainy	90%

Sunny	95%
Rainy	5%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{sunny}) + p(\text{BCN} \ \& \ \text{sunny})}$

$$= \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{sunny}) + p(\text{BCN} \ \& \ \text{sunny})}$$

x: meeting was in Barcelona

y: it was sunny

Helsinki 50%

Barcelona 50%

Sunny	10%
Rainy	90%

Sunny	95%
Rainy	5%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{sunny}) + p(\text{BCN} \ \& \ \text{sunny})}$

$$= \frac{0.95 * 0.5}{0.5 * 0.1 + 0.5 * 0.95}$$

x: meeting was in Barcelona

y: it was sunny

Helsinki 50%

Barcelona 50%

Sunny	10%
Rainy	90%

Sunny	95%
Rainy	5%

Bayes' theorem $p(x|y) = \frac{p(y|x) p(x)}{p(\text{HEL} \ \& \ \text{sunny}) + p(\text{BCN} \ \& \ \text{sunny})}$

$$= \frac{0.95 * 0.5}{0.5 * 0.1 + 0.5 * 0.95} \approx 0.90$$

x: meeting was in Barcelona
 y: it was sunny

Helsinki 50%
 Barcelona 50%

Sunny	10%
Rainy	90%

Sunny	95%
Rainy	5%

INFORMATION CONTENT ANALYSIS

- Bayes' theorem
- Multivariate Gaussian distributions + additive obs. noise
- Combine AERONET inversion & MODIS observations

→ Can we improve SSA retrievals at low AOD?

Analysis of fine-mode aerosol retrieval capabilities by different passive remote sensing instrument designs

Kirk Knobelspiesse,^{1,2,*} Brian Cairns,¹ Michael Mishchenko,¹ Jacek Chowdhary,^{3,1} Kostas Tsigaridis,^{3,1} Bastiaan van Diedenhoven,^{3,1} William Martin,^{3,1} Matteo Ottaviani,^{1,4} and Mikhail Alexandrov^{3,1}

¹NASA Goddard Institute for Space Studies, 2880 Broadway, New York, New York 10025, USA

²NASA Postdoctoral Program fellow, USA

³Columbia University, 2880 Broadway, New York, New York 10025, USA

⁴Stevens Institute of Technology, Hoboken, New Jersey, USA

[*kirk.knobelspiesse@nasa.gov](mailto:kirk.knobelspiesse@nasa.gov)

INFORMATION CONTENT ANALYSIS

$$\hat{\mathbf{S}}^{-1} = \mathbf{K}^T \mathbf{S}_{\varepsilon}^{-1} \mathbf{K} + \mathbf{S}_a^{-1}$$

Analysis of fine-mode aerosol retrieval capabilities by different passive remote sensing instrument designs

Kirk Knobelspiess^{1,2,*} Brian Cairns,¹ Michael Mishchenko,¹ Jacek Chowdhary,^{3,1} Kostas Tsigaridis,^{3,1} Bastiaan van Diedenhoven,^{3,1} William Martin,^{3,1} Matteo Ottaviani,^{1,4} and Mikhail Alexandrov^{3,1}

¹NASA Goddard Institute for Space Studies, 2880 Broadway, New York, New York 10025, USA

²NASA Postdoctoral Program fellow, USA

³Columbia University, 2880 Broadway, New York, New York 10025, USA

⁴Stevens Institute of Technology, Hoboken, New Jersey, USA

[*kirk.knobelspiess@nasa.gov](mailto:kirk.knobelspiess@nasa.gov)

AERONET SSA

- Available in the inversion product
 - Level-2
 - AOD (at 440 nm) > 0.4
-

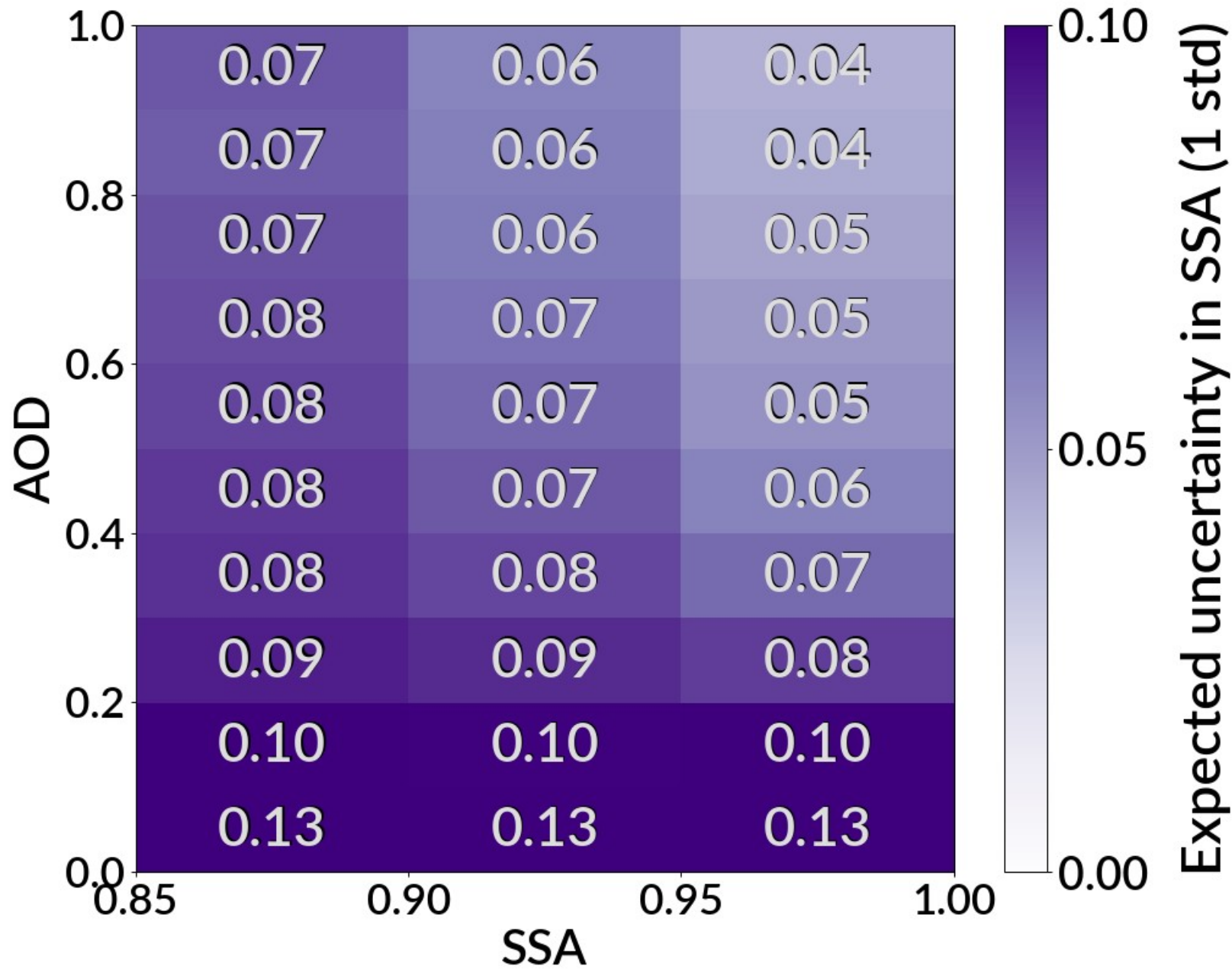
Technical details:

- Analysis based on radiative transfer (libRadtran) simulations
- MODIS bands 1-8 used, SZA=65°
- AERONET almucantar radiances at 440, 675, 870, 1020 nm, SZA=65°
- 2.5% uncertainty in MODIS reflectance, 5% AERONET almucantar radiances
- Parameters to be estimated
 - AOD, SSA, ozone, aerosol layer height, Angstrom exponent (AE), absorbing AE, asymmetry parameter, surface pressure



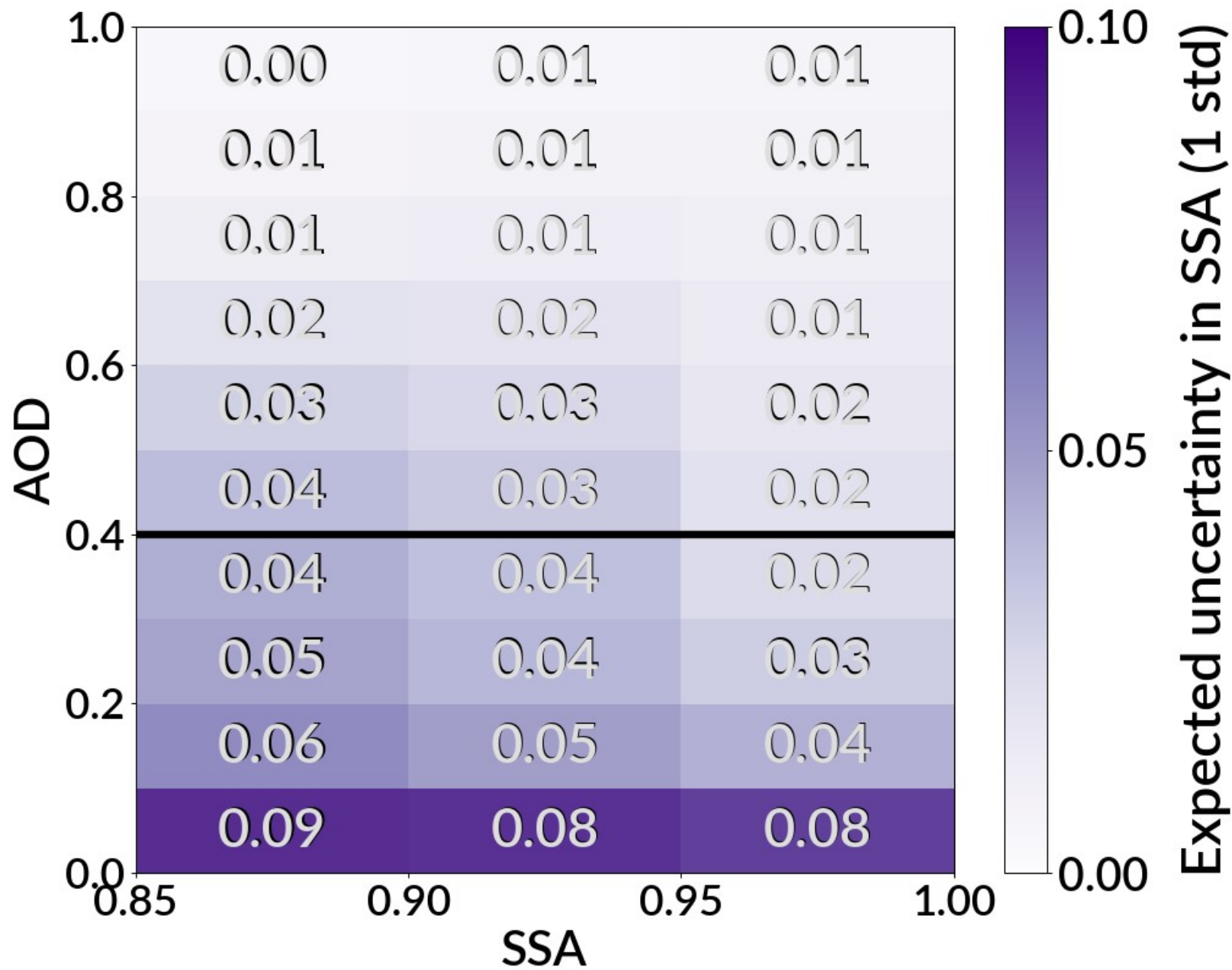
SSA

MODIS



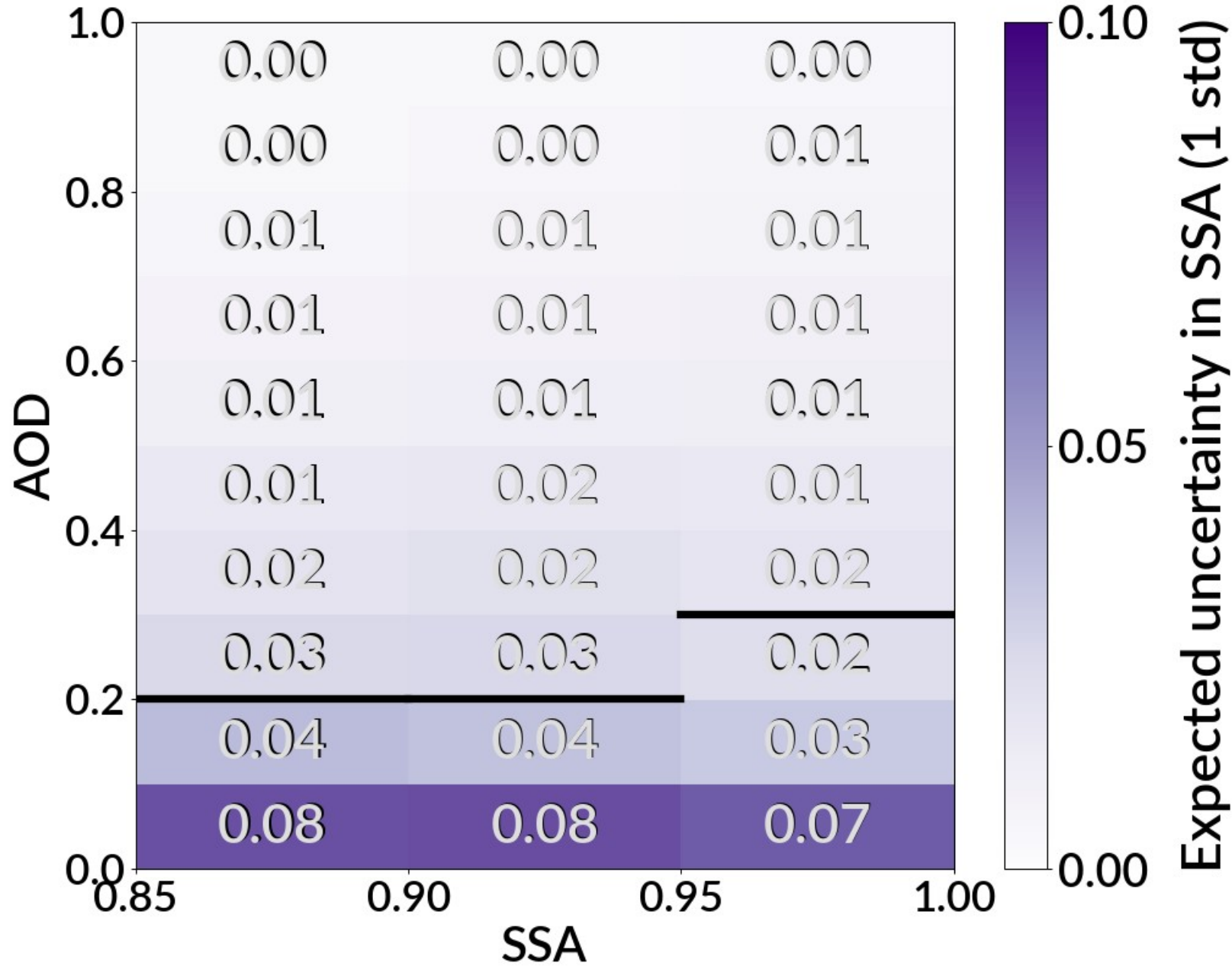
SSA
Prior
Std
0.15

AERONET



SSA
Prior
Std
0.15

AERONET + MODIS



SSA
Prior
Std
0.15

- Bayes theorem' is a useful tool for info content analysis
- Combining bottom-of-the-atmosphere (BOA) and top-of-the-atmosphere (TOA) observations may improve the SSA retrievals
- Improved retrievals of SSA (and possibly other aerosol parameters) may help aerosol typing



ILMATIETEEN LAITOS
METEOROLOGISKA INSTITUTET
FINNISH METEOROLOGICAL INSTITUTE

Bayes' theorem is useful in info content analysis

Sometimes $1 + 1 > 2$

**Improved SSA retrievals may
help aerosol typing**



@anttilip



antti.lipponen@fmi.fi

