# The SPARC Data Initiative -A multi-instrument comparison of stratospheric limb measurements

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and the SPARC Data Initiative Team

7<sup>th</sup> Atmospheric Limb Conference, Bremen, 2013

# Motivation

# Knowledge of quality of different satellite data sets needs to be improved for different applications: Tracer scenario validation (Montreal Protocol, Cl<sub>y</sub>) Model validation projects (CCMI, IPCC)

Trend analyses (e.g., stratospheric water vapour)

Empirical studies of stratospheric climate and variability

# **Objectives**

Inter-comparison of vertically resolved climatologies of 25 chemical tracers and aerosol from 18 multi-national satellite instruments

- Will be published as a peer-reviewed SPARC report, as well as in journal publications
- Will summarize useful information and highlight differences between data sets
- Will provide guidance to space agencies about required improvements in existing data sets and future observations

#### **SPARC** Data Initiative

#### SPARC Data Assimilation workshop, Paris, 2015

Team
Co-leads:
Michaela Hegglin
Susann Tegtmeier

- HALOE (UARS): John Anderson
   MLS (Aura/UARS): Lucien Froidevaux, Ryan Fuller
  - TES (Aura): Jessica Neu
  - ACE-FTS (SCISAT-1): Kaley Walker, Ashley Jones
  - MAESTRO: Kaley Walker
  - OSIRIS (Odin): Doug Degenstein, Adam Bourassa
  - SMR (Odin): Joachim Urban
  - MIPAS (ENVISAT): Thomas von Clarmann, Bernd Funke
  - SCIAMACHY (ENVISAT): Alexei Rozanov
  - GOMOS (ENVISAT): Erkki Kyröla
  - SAGE I / II / III: Ray Wang
  - HIRDLS (AURA): John Gille, Lesley Smith
  - SMILES (ISS): Yasuko Kasai
  - LIMS (NIMBUS-7): Ellis Remsberg, Gretchen Lingenfelser
  - POAM II / III: Jerry Lumpe
  - Matthew Toohey

SPARC Data Initiative	<b>O</b> <sub>3</sub>	H <sub>2</sub> O	CH₄		CCI3F	CCI <sub>2</sub> F <sub>2</sub>	CO	HF	${\sf SF}_6$	NO		NO <sub>x</sub>		HNO₄	$N_2O_5$	<b>CIONO</b> <sup>2</sup>	NOy	HCI	CIO	HOCI	BrO	НО			<b>CH</b> <sup>3</sup> <b>CN</b>	aerosol
ACE-FTS	X	X	X	X	X	X	Χ	X	Χ	X	X	X	Χ	X	Χ	X	Χ	X						X		
Aura-MLS	Χ	X		X			Χ						Χ					X	Χ	X		X	X			
GOMOS	Χ										Χ															X
HALOE	X	X	X					X		X	Χ	Χ						X								
HIRDLS	Χ				Χ	X					Χ		Χ													
LIMS	Χ	Χ									Χ		Χ													
MAESTRO	Χ																									
MIPAS	X	Χ	Χ	X	X	X	X		Χ	X	Χ	X	Χ	Χ	X	X	X		Χ	X				X		
OSIRIS	Χ										Χ	Xd					Xm				Х					Χ
POAM II	Χ										Χ															Χ
POAM III	Χ	Χ									Χ															Χ
SAGE I	Χ																									
SAGE II	Χ	Χ									Χ															Χ
SAGE III	Χ	Χ									Χ															Χ
SCIAMACHY	Χ	Χ									Χ	Xd									Χ					Χ
SMILES	Χ												Χ					Χ	Χ	Χ	Χ				Χ	
Odin/SMR	Χ	Χ		X			Χ			X			Χ				Xm		Χ				Xlc			
UARS-MLS	Χ	X											Χ						Χ							
TES	Xt																									

# 'Climatologies'

- Monthly mean zonal mean time series
  - VMR or aerosol extinction coefficients,
  - 1σ standard deviation,
  - number of measurements per grid box,
  - mean, min, and max local solar time,
  - average day of month and latitude.
- Range: upper troposphere to the lower mesosphere
- Time period covered: 1978 2010
- Grid: 5° latitude bins on the CCMVal-2 pressure grid (28 levels)
- Data sets are provided in a common format (netcdf) easily useable by the atmospheric science community

# **SDI report**

- SPARC report no. 7
- All chapters written, most typeset, some in proof read
- Publication in 2015/2016



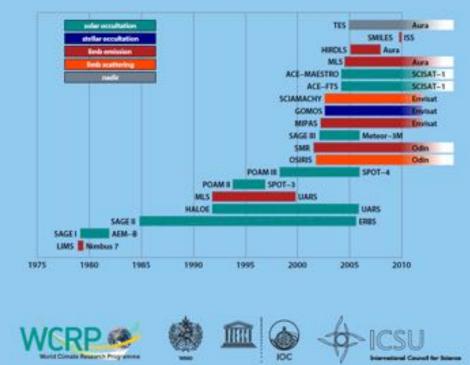


Core Project of the WMO/ICSU/IOC World Climate Research Programme

The SPARC Data Initiative: Assessment of stratospheric trace gas and aerosol climatologies from satellite limb sounders

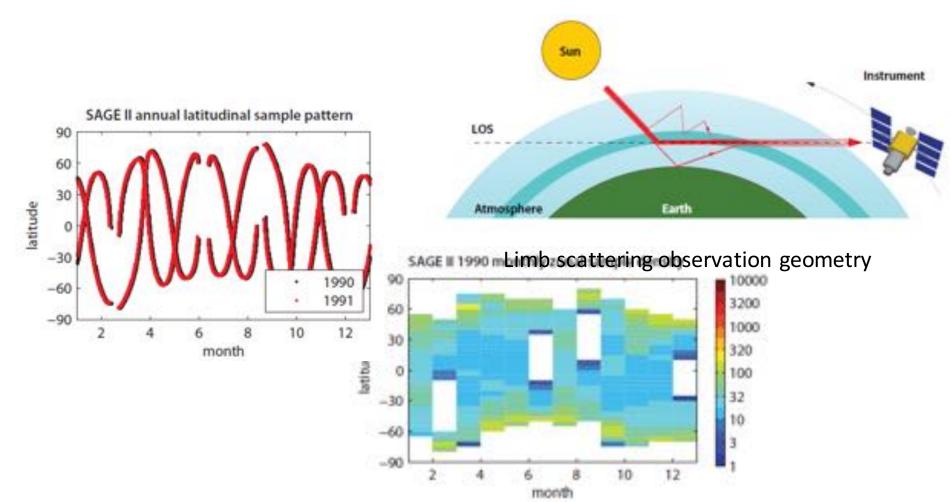
Edited by M. I. Hegglin and S. Tegtmeier

SPARC Report No. 7, WCRP-01/2015, February/2015



# **Chapter 2 - Satellite instruments and data sets**

Satellite measurement techniques (observation geometry, wavelengths, orbits)



# **Chapter 3 - Climatology framework**

Climatology construction (methodology, local time scaling, instrument-specific information)

Instrument	Latitudinal coverage	LT at equator <sup>1</sup>	LT of measure- ment <sup>2</sup>	Inc. <sup>3</sup>	Vert. Grid <sup>4</sup>	Alternate grid <sup>s</sup>	Meas."	Conver- sion to VMR <sup>7</sup>	Data density per day	
LIMS on Nimbus 7	64°S-84°N (daily)	a: 11:51 am d: 11:51 pm	a:1 pm d: 11 pm	99.3"	р	N/A	VMR	N/A	3000	
SAGE I on AEM-B	75°S–75°N (~one month)	N/A	N/A	56°	z	NCEP	ND	NCEP	30	
SAGE II on ERBS	75°S–75°N (~one month)	N/A	N/A	57°	z	NCEP	ND	NCEP	30	
OSIRIS on Odin	82°S–82°N (daily, no winter hemisphere)	a: 6:30 pm d: 6:30 am	a: 6:30 pm d: 6:30 am	97.8°	z	ECMWF operation- al analysis	ND	ECMWF operation- al analysis	300 - 975	
SMR on Odin	83°S–83°N (daily)	a: 6:30 pm d: 6:30 am	a: 6:30 pm d: 6:30 am	97.8°	р	N/A <sup>8</sup>	VMR	N/A	600-975	
GOMOS on Envisat	90°S–90°N (daily, no summer poles for night )	a: 10:00 pm d: 10:00 am	a: 10-12 pm d: 8-10.30 am	98.55°	z	ECMWF operation- al analysis	ND	ECMWF operation- al analysis	100-300 (night mea- surments)	
MIPAS	90°S-90°N	a: 10:00 pm	a- 10-00 pm	98.55°	7	MIPAS	VMR	N/A	1000 (1300	

Climatology uncertainties and diagnostics

# 'Climatological' validation approach based on binned/interpolated datasets

# Advantages

- Consistent between all instruments
- Avoids sensitivity to arbitrary coincidence criteria
- Larger sample sizes (minimize the random sampling error)

# Disadvantages

- Biased mean values due to non-uniformity of sampling
- Different resolutions in altitude (averaging kernels)

# **Evaluations**

- Need for a reference that does not favor a certain instrument -> Use of the multi-instrument mean (MIM)
- But: MIM is not the best climatology available, can suffer from changes in time periods and set of available instruments

What can we learn from the SPARC Data Initiative?

- 1) Sampling Bias Study
- 2) Study on the impact of the vertical resolution
- 3) Uncertainty in our knowledge of the atmospheric mean state
  - Based on diagnostics of monthly /annual zonal mean cross-sections
  - Mean spread between data sets over maximum number of years

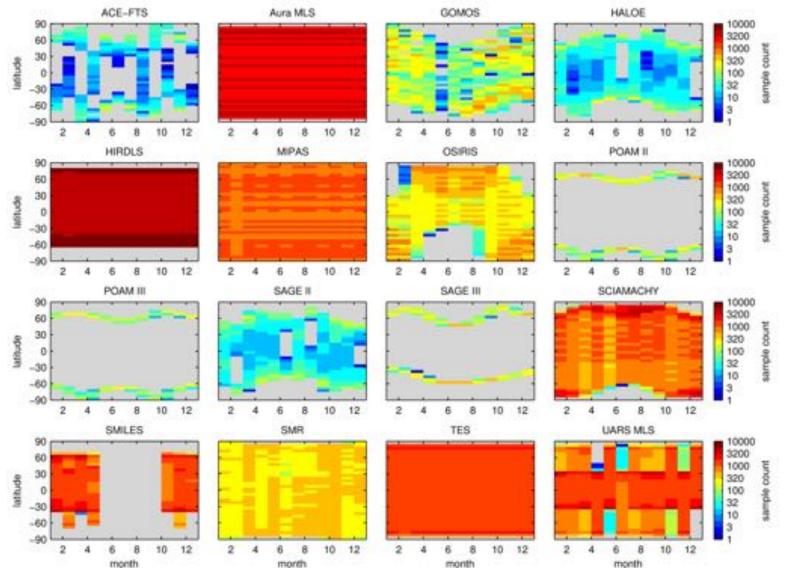
## 4) Outliers and unphysical features

 Evaluation of variability and other physical features (e.g., seasonal cycle, QBO, tape recorder, Antarctic ozone, polar vortex dehydration, EPP NO<sub>x</sub>)

# 5) Implications for model-measurement intercomparison

#### SPARC Data Assimilation workshop, Paris, 2015

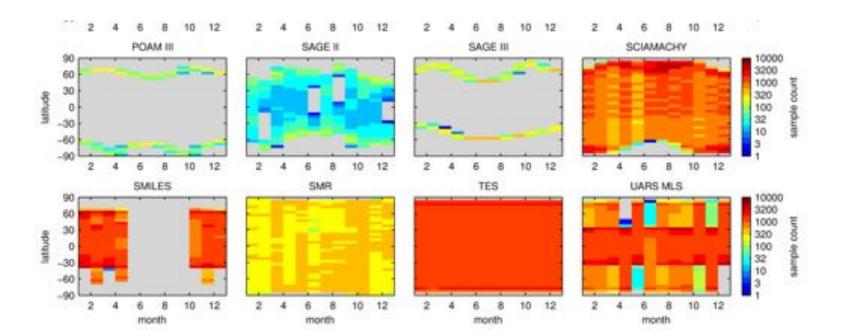
#### Sampling patterns: sample counts



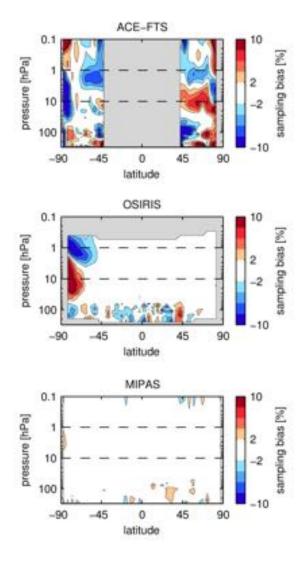
month

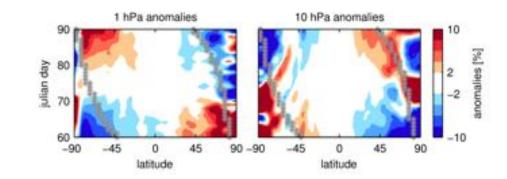
# **Sampling Bias Study**

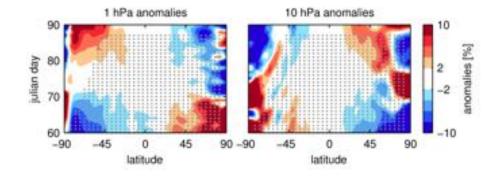
- Use chemical fields from a coupled chemistry model, (e.g., WACCM)
- Sample model fields based on space-time sampling patterns of specific instruments ("satellite simulator")
- Difference of sample mean and population mean (with full resolution model fields) gives estimate of potential sample bias in climatologies.

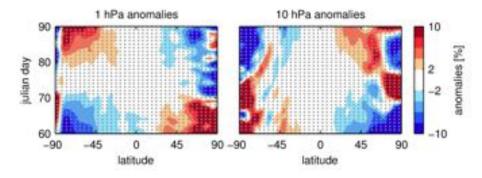


# March O<sub>3</sub> case study: impact of temporal non-uniformity



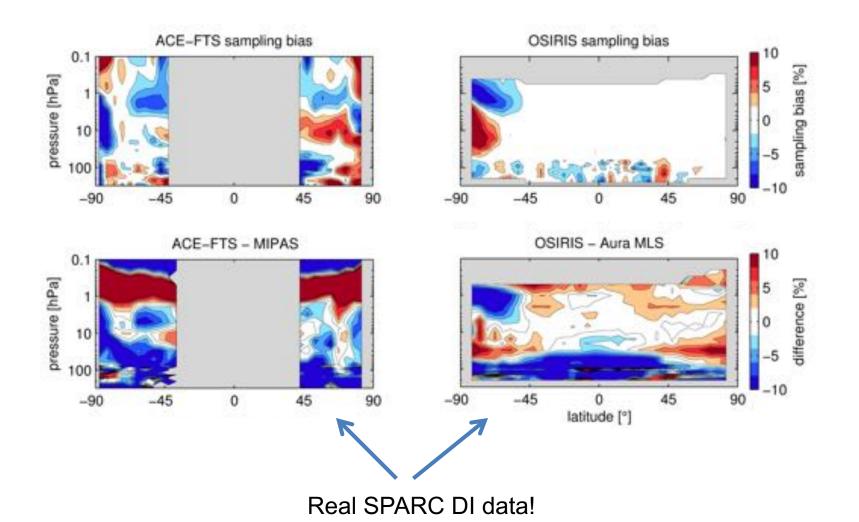






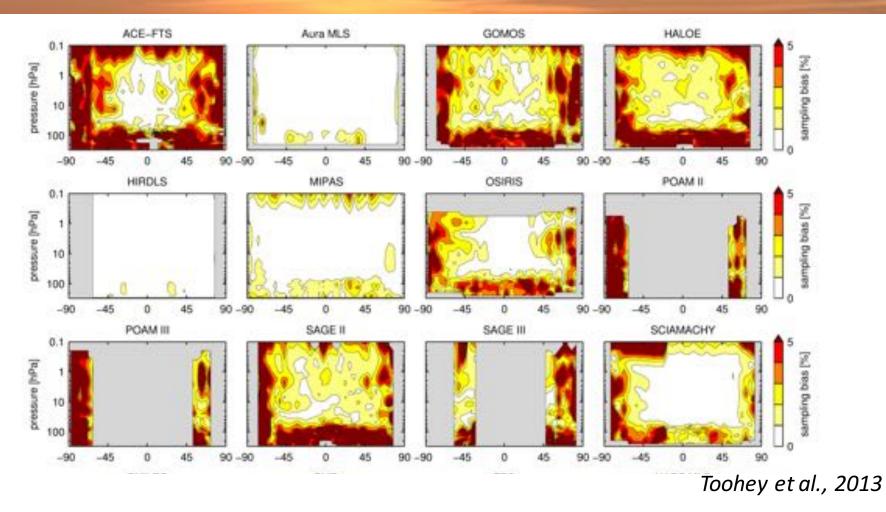
Toohey et al., 2013

# March O<sub>3</sub> Case study: reality check



# **Sampling bias**

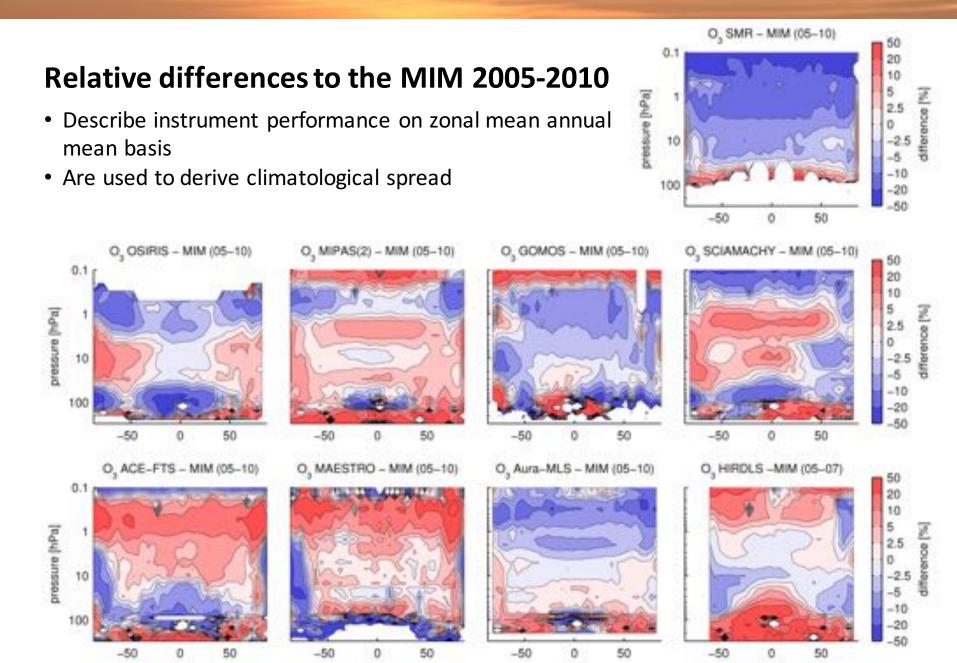
#### SPARC Data Assimilation workshop, Paris, 2015



- Instruments with regular and uniform sampling patterns -> small sampling bias
- Instruments with varying latitudinal coverage -> strong sampling biases for certain months and locations
- Sampling biases for  $O_3$  can be  $\geq 10\%$  (non-uniformity in day-of-month sampling)

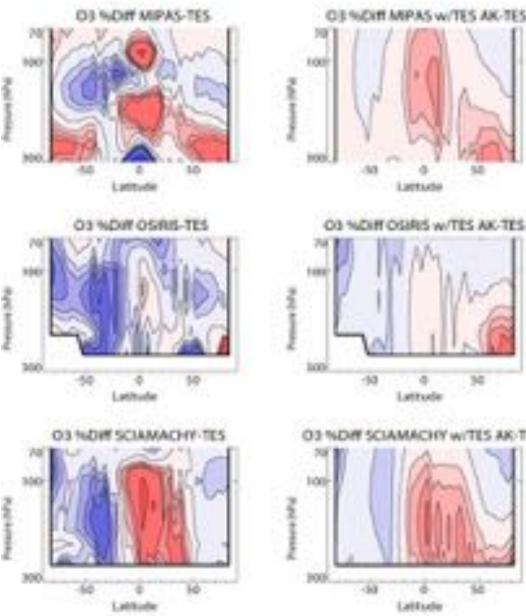
#### **Ozone comparisons**

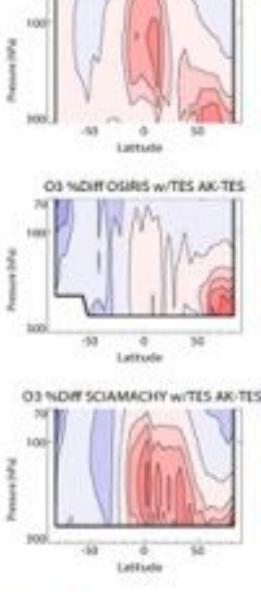
#### SPARC Data Assimilation workshop, Paris, 2015



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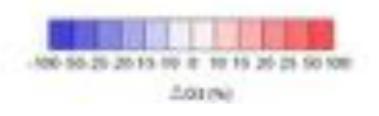
#### SPARC Data Assimilation workshop, Paris, 2015





# **Application of TES** observational operator to limb-viewing instruments

- Minimizes impact of vertical resolution
- Allows for identification of • systematic difference e.g., positive bias of most instruments in tropical LS



Neu et al., 2014

# Impact of vertical resolution

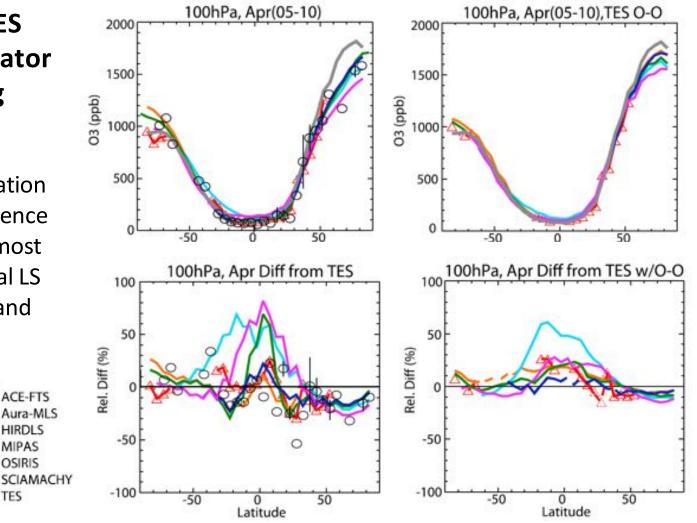
# **Application of TES** observational operator to limb-viewing instruments

Allows for identification of systematic difference e.g., positive bias of most instruments in tropical LS with respect to TES and ozone sondes

HIRDLS

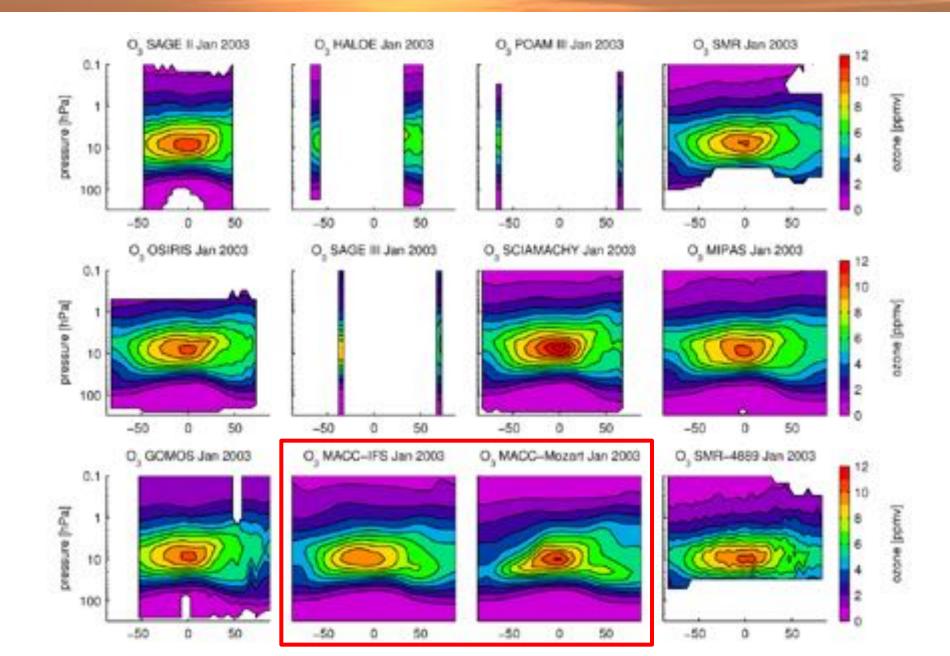
MIPAS OSIRIS

TES



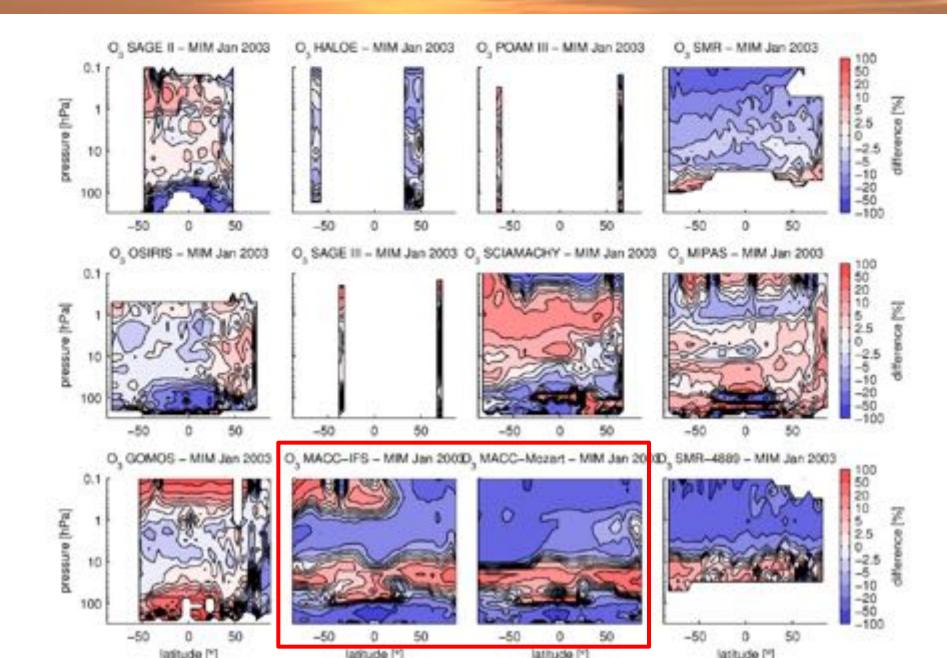
#### **SPARC** Data Initiative

#### SPARC Data Assimilation workshop, Paris, 2015



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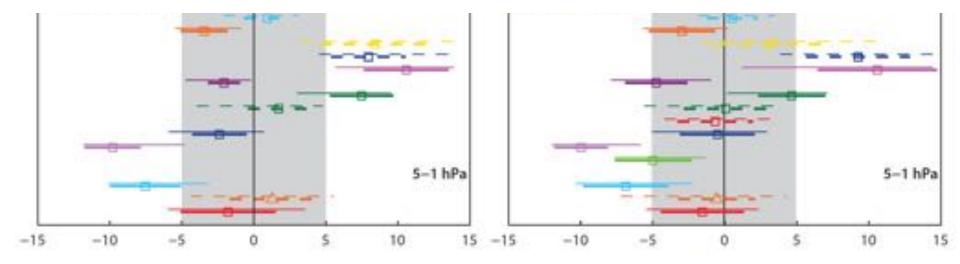
#### SPARC Data Assimilation workshop, Paris, 2015



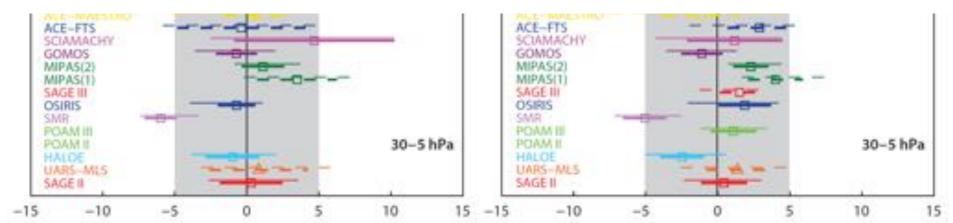
#### SPARC Data Assimilation workshop, Paris, 2015

#### Evaluation of 18 ozone profile data sets [relative differences %]

**Upper stratosphere** (5-1 hPa) - Good agreement: ±5% to ±10%.

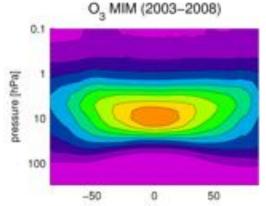


Middle stratosphere (30-5 hPa) - Lowest spread between the instrument data sets: ±5%



# O<sub>3</sub> atmospheric mean state

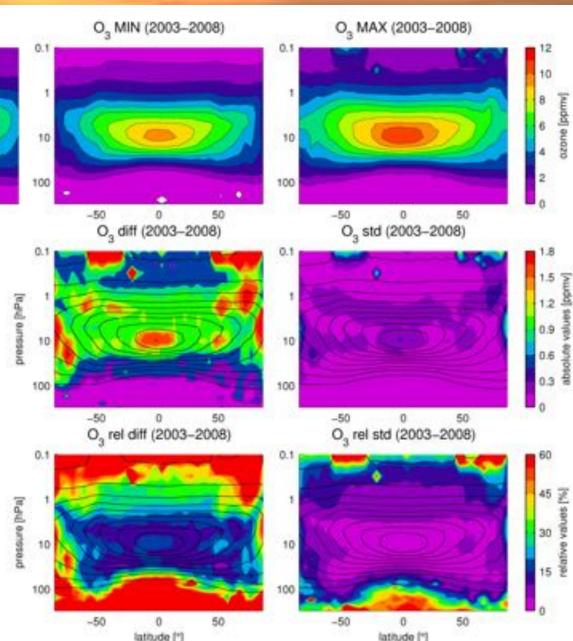
#### SPARC Data Assimilation workshop, Paris, 2015



# Uncertainty in our knowledge of the atmospheric mean state

- Smallest in the tropical MS and midlatitude LS/MS (1σ multiinstrument spread ±5%)
- Maximum ozone VMR (large spread 10 and 12 ppmv)
- Polar latitudes: larger spread of the ozone mean state (1σ of ±15%) and maximum variations (1σ of ±30%) in the Antarctic LS

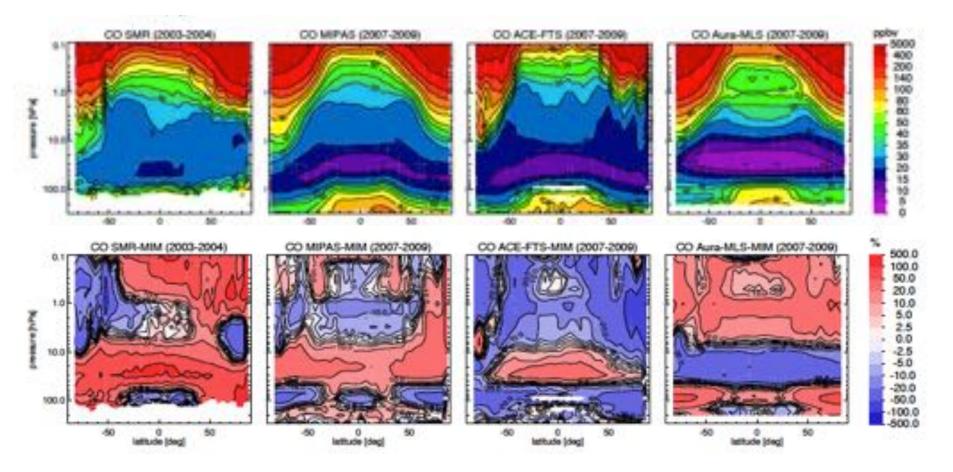
Tegtmeier et al., 2013



#### **CO** comparisons

#### SPARC Data Assimilation workshop, Paris, 2015

#### CO annual zonal mean cross sections

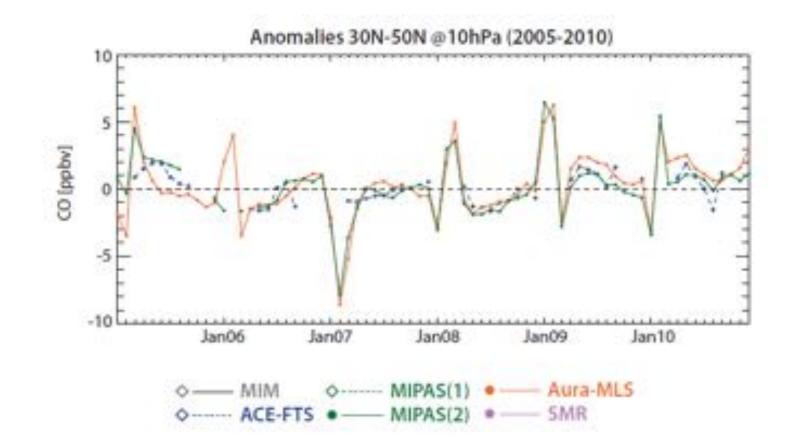


Large differences exist in some of the species in the annual zonal means.
Further retrieval studies are suggested to get at the cause of this discrepancy.

#### **CO** comparisons

#### SPARC Data Assimilation workshop, Paris, 2015

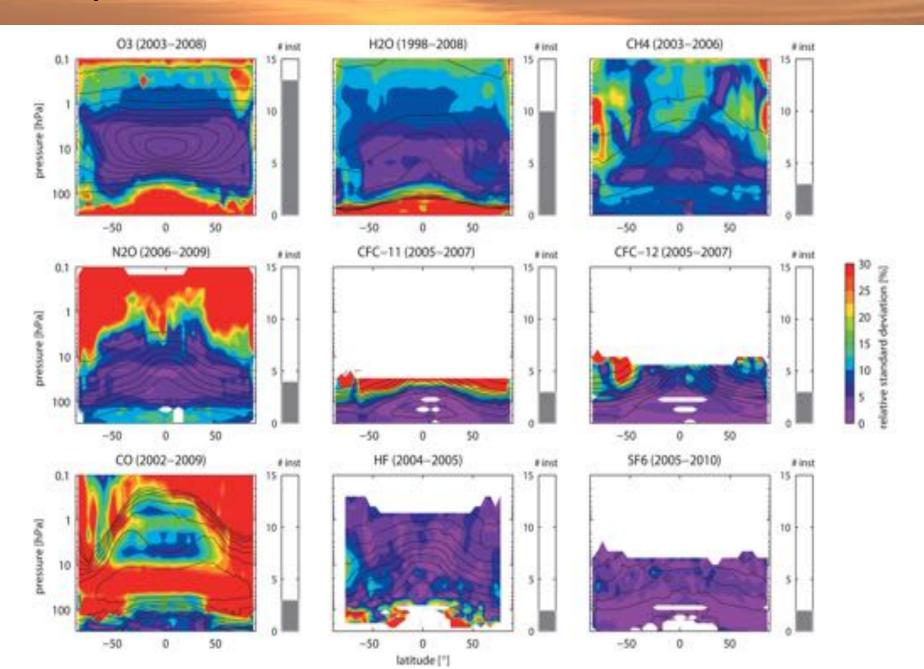
#### CO annual zonal mean cross sections

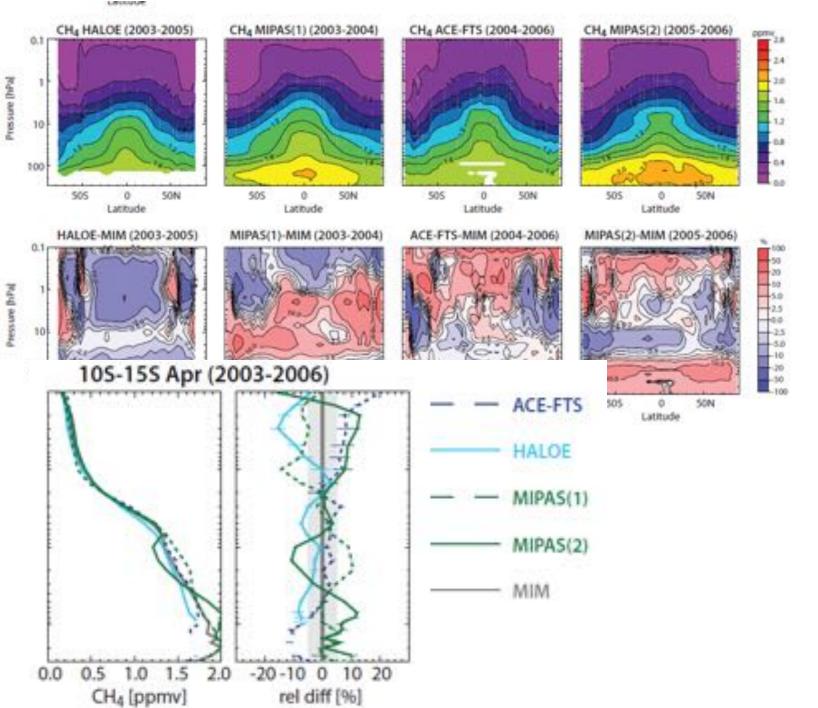


Large differences in the annual mean comparisons but very good agreement of the interannual variability.

#### **Atmospheric mean state**

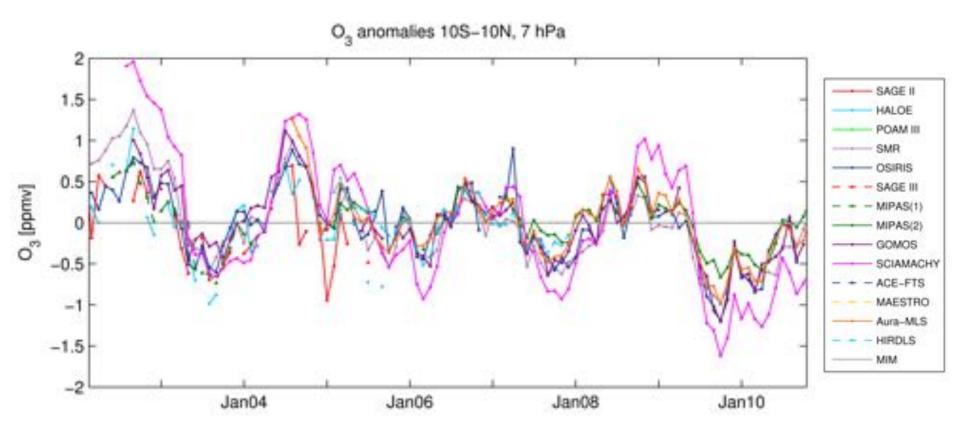
#### SPARC Data Assimilation workshop, Paris, 2015





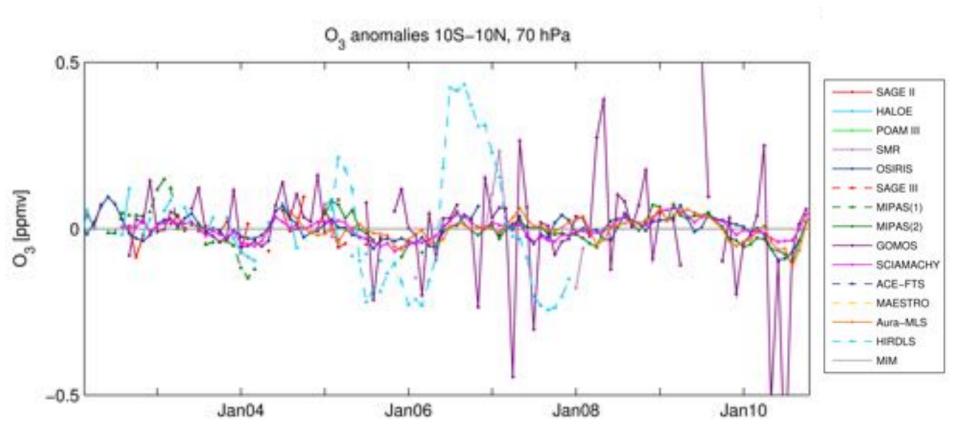
CH<sub>4</sub>

# **Ozone - Evaluation of interannual variability**



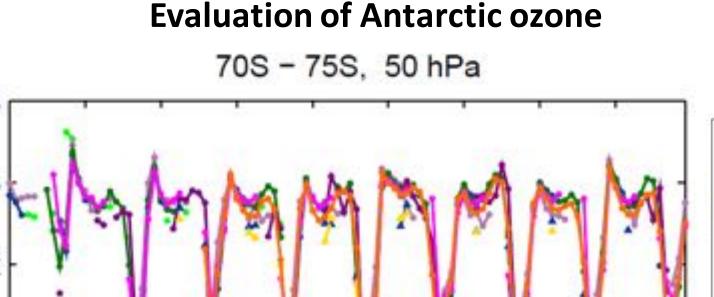
- Tropical QBO signal in the middle stratosphere is captured well by all instruments
- Slight deviations in displayed amplitude

# **Ozone - Evaluation of interannual variability**

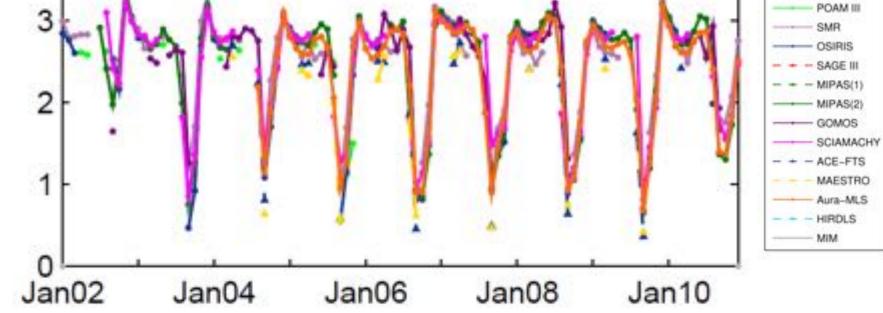


 Larger difficulties in the lower stratosphere where ozone abundances and inter-annual variations are small

SAGE II HALOE

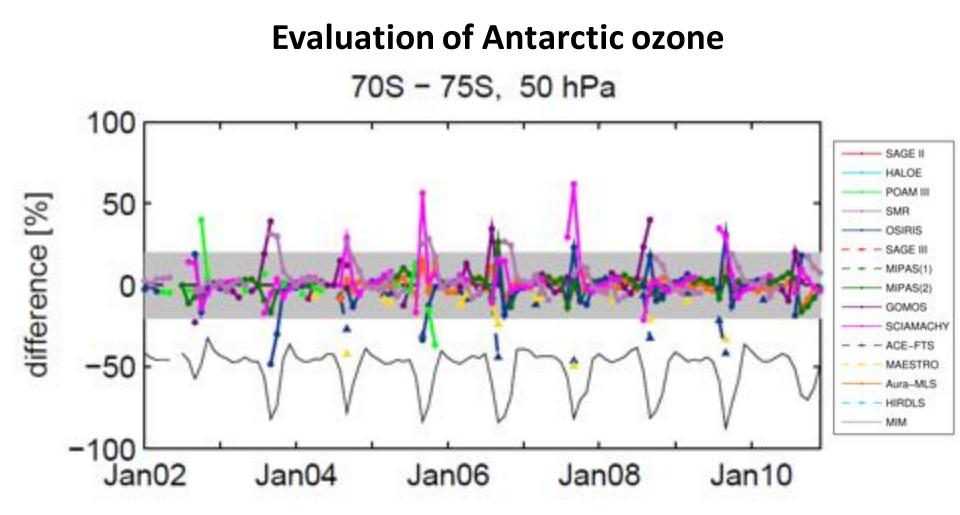


VMR [ppm]



- Large relative differences (to the MIM) in the Antarctic polar cap region during the time of the ozone hole
- Spread between the monthly zonal mean fields of ±50%





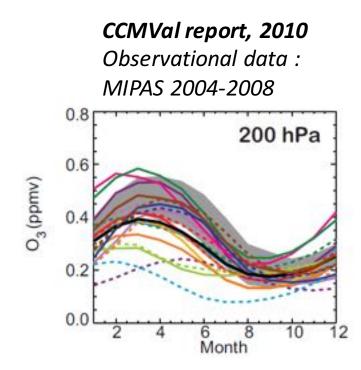
- Large relative differences (to the MIM) in the Antarctic polar cap region during the time of the ozone hole
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# Stratospheric satellite data for model evaluation

- Data Initiative: Provides the data and basic knowledge on data quality
- Mixed team of scientist
  - ✓ Generate list of diagnostics appropriate for model evaluations
  - ✓ Provide a 'best' estimate and its uncertainty range for ready use in model-measurement comparisons (CCMVal diagnostic tool)
- "Recipe" for deriving an observational uncertainty range

✓  $O_3$  seasonal cycle at 200 hPa, mid-latitudes

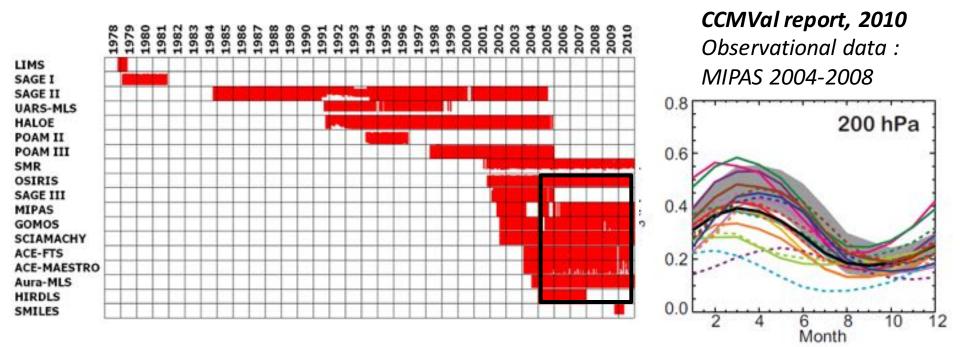
Evaluate the representation of large-scale transport and mixing processes



Evaluate the representation of large-scale transport and mixing processes

#### Step 1: Define time period

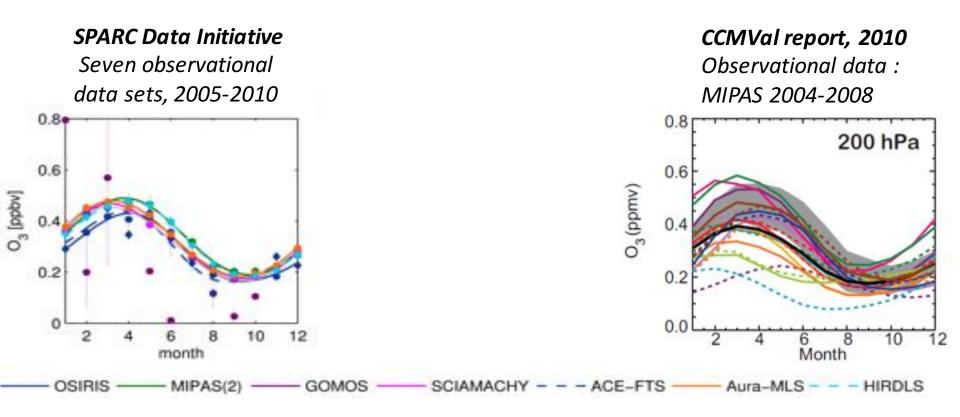
Find overlap period of maximum length including maximum number of instruments



Evaluate the representation of large-scale transport and mixing processes

#### Step 1: Define time period

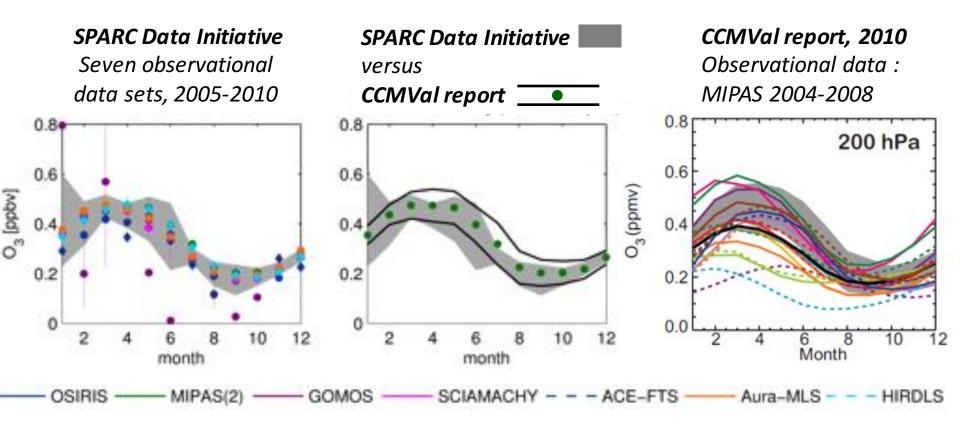
Find overlap period of maximum length including maximum number of instruments



Evaluate the representation of large-scale transport and mixing processes

#### Step 2: Instrument spread

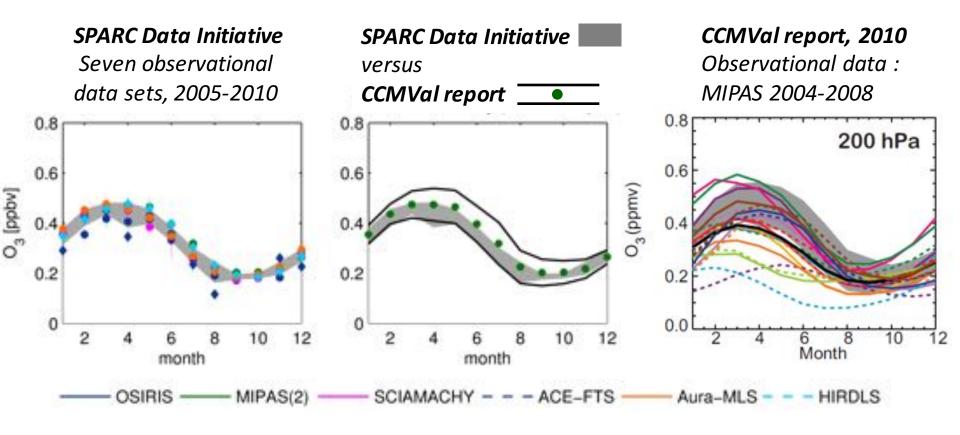
Calculate uncertainty range as the standard deviation over the instrument spread



Evaluate the representation of large-scale transport and mixing processes

#### **Step 3: Eliminate outliers**

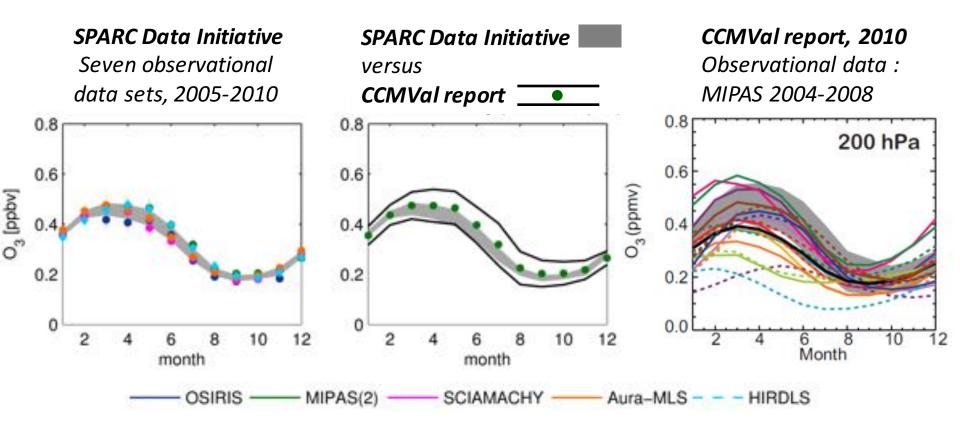
Remove all data points outside of the 3 $\sigma$  uncertainty range of all other instruments



Evaluate the representation of large-scale transport and mixing processes

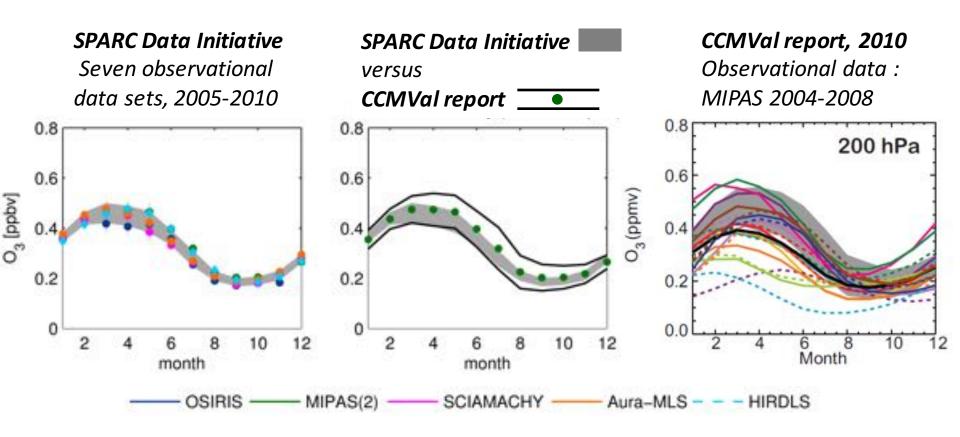
#### Step 4: Eliminate sampling bias

Remove data impacted by sparse sampling (Toohey et al., 2013)

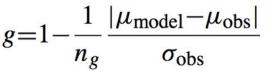


Evaluate the representation of large-scale transport and mixing processes

**Step 5: Include interannual variability** 

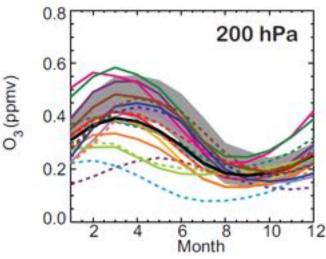


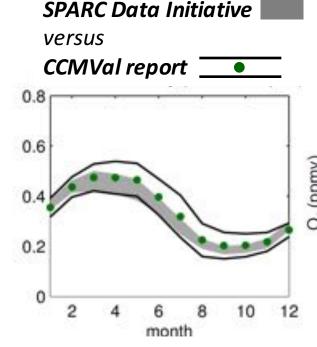
- Reduced uncertainty range
- Observational range shifted to slightly lower values
- Shifted phase (earlier maximum)



Douglass et al., 1998 Waugh and Eyring, 2008

CCMVal report, 2010 Observational data : MIPAS 2004-2008





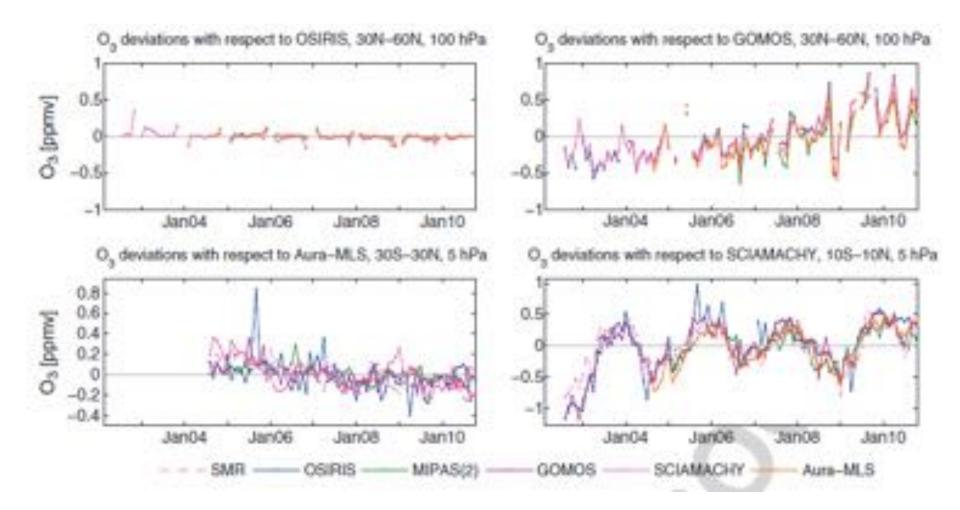
#### **Merging techniques**

- Merging of two single datasets by accounting for an inter-instrument bias calculated over some overlap time period (SAGE/GOMOS)
- Merging of multiple datasets that uses detailed error characterization of instruments (Froidevaux GOZCARDS, Davis SWOOSH)
- Statistical methods to fill in observational gaps (BDPD)
- Nudged chemistry-climate model as transfer function between the instruments (Hegglin et al., 2014)

#### Problems

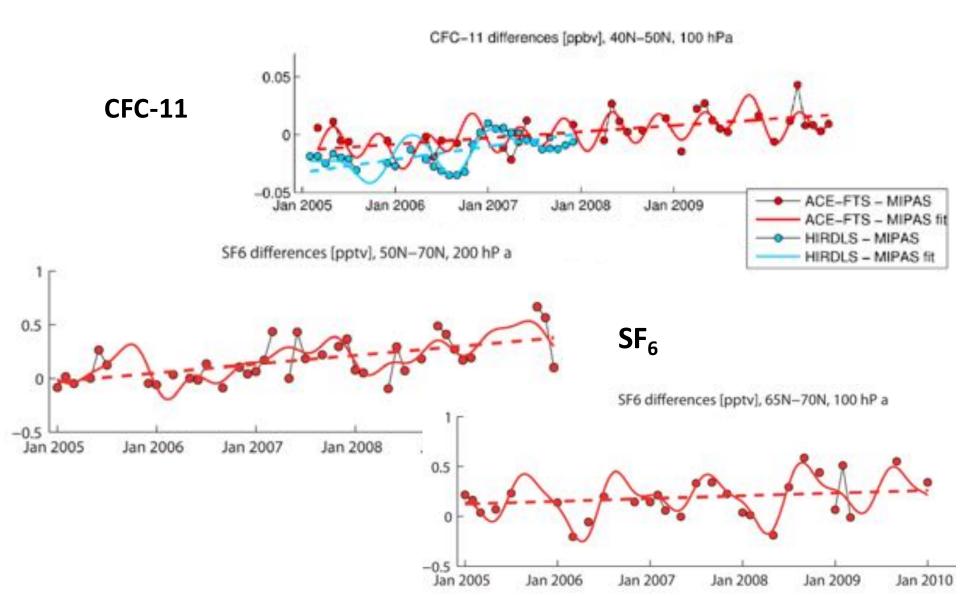
- Inverse estimated error (can differ between data sets, ideally instruments error covariances)
- Key problems: different altitude resolutions and different content of a priori information (application of the averaging kernel matrix)
- Drifts hard to identify (comparisons between two satellites non-conclusive, in situ measurements as reference often lead to lack of statistical significance)

### Drifts identified by comparison to all other available data sets



#### SPARC Data Assimilation workshop, Paris, 2015

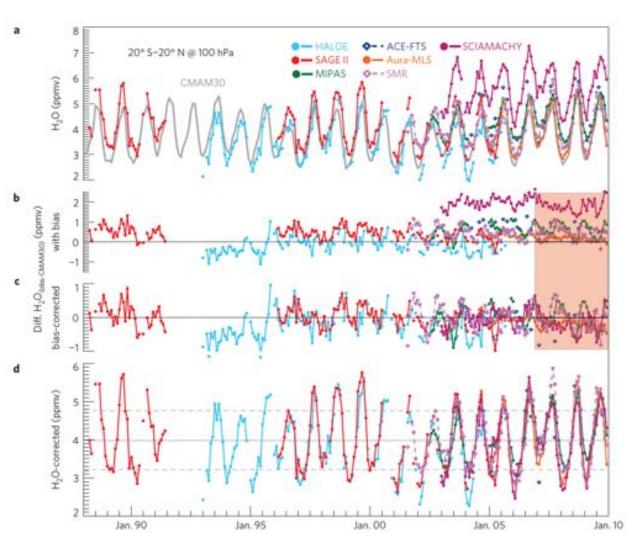
Multi-linear regression of time series of differences between pairs of instruments



# **Implications for merging**

# CCM nudged to observed meteorology is used as transfer function between observational data sets

- Long-term data record
- Instrumental drifts
- Representativeness of spatially limited data sets.
- Lower/mid stratospheric water vapour trends are negative
- Upper stratospheric water vapour trends are positive, (accelerated BDC in the lower stratosphere)



Hegglin et al, 2014

#### **Summary**

#### **Comprehensive comparison of satellite instrument observations**

- Better knowledge of the quality of available data products including information on where they are consistent and where they exhibit unphysical features or strong deviations
- Assessment of the range of measurements as an estimate of the systematic uncertainty in the measured field
- Need for further evaluation activities (e.g., in the UTLS and at high latitudes) identified
- Motivation for improvement of data products

#### Provide monthly zonal mean time series in a common format

- Will be published on the SPARC data archive website
- Will be updated in the future as soon as new time series are available)

#### Improve future model-measurement comparison activities

• Depending on the evaluation and trace gas, individual instruments may need to be excluded from the comparison (e.g., seasonal cycle in LS)

#### 7<sup>th</sup> Atmospheric Limb Conference, Bremen, 2013

### Aerosol

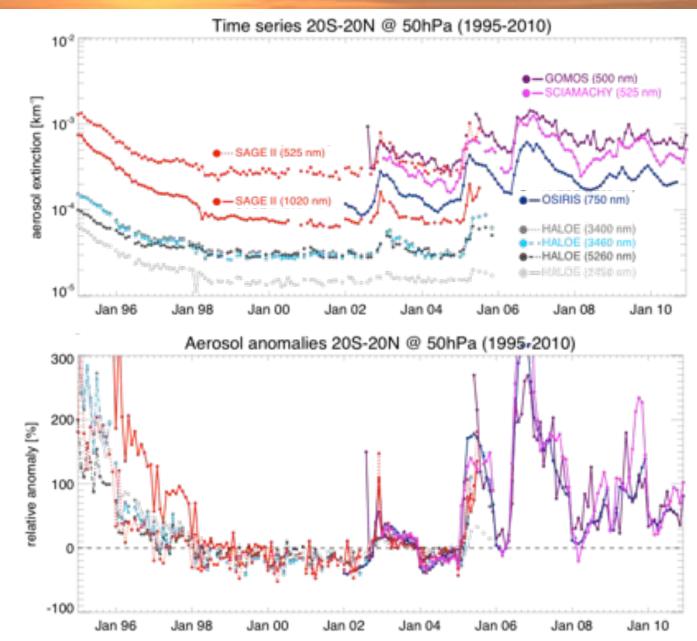
# Aerosol anomaly evaluations

Aerosol time series between 1995 and 2010.

Aerosol anomalies are shown relative to 2003-2004 monthly means.

Relative anomalies compare well, with exception:

- SAGE II 1020 nm
- HALOE 5260 nm



#### 7<sup>th</sup> Atmospheric Limb Conference, Bremen, 2013

# Aerosol

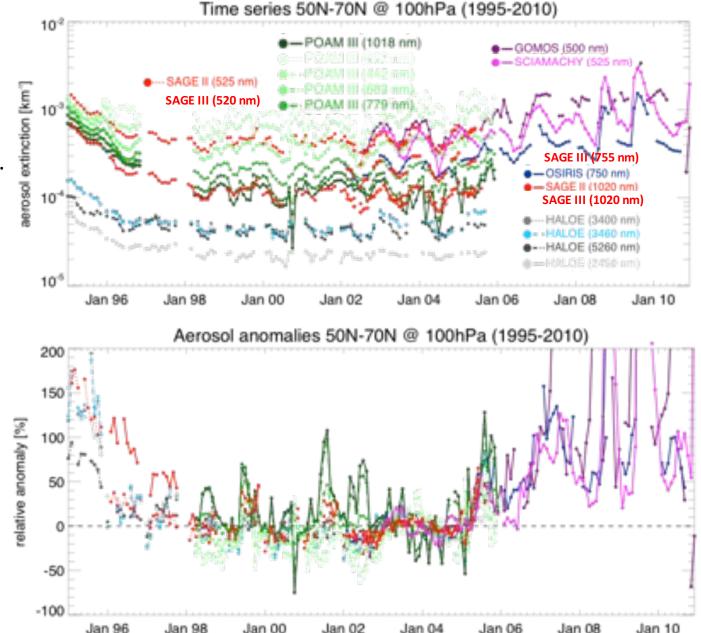
# Aerosol anomaly evaluations

Aerosol time series between 1995 and 2010.

Aerosol anomalies are calculate using the 2003-2004 monthly means.

Extra-tropical lower stratosphere 'outliers':

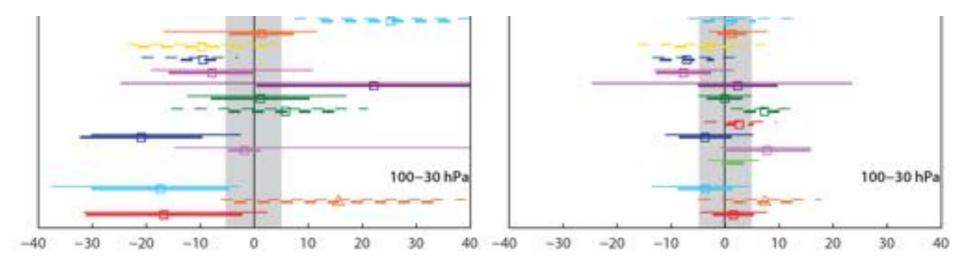
- SAGE II 1020 nm
- POAM III 779 nm



#### SPARC Data Assimilation workshop, Paris, 2015

#### **Evaluation of 18 ozone profile data sets**

Lower stratosphere (100-30 hPa) - Tropics: ±20% and Mid-latitudes: ±10%



Upper troposphere (300-100 hPa) - Tropics: ±20% to ±50% and Mid-latitudes: mostly ±10%

