



Preliminary analysis of FAIRMODE CT9 results

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with inputs from CT9 participants modelling teams

JRC (EC/EU)
ZAMG (AT)
Met Norway (NO)
CyI (CY)
NKUA (GR)
DHMZ (HR)
LMD/IPSL (FR)
UH-CACP (UK)
CIEMAT (ES)
ENEA (IT)
UNIBS (IT)
IRCELINE (BE)

JRC, Ispra, IT

FAIRMODE CT9

➤ Models provide different absolute results C_{scen}^M

➤ **BUT HOW DO THEY BEHAVE ON DELTAS?**

$$\Delta = C_{scen}^M - C_{bc}^M$$

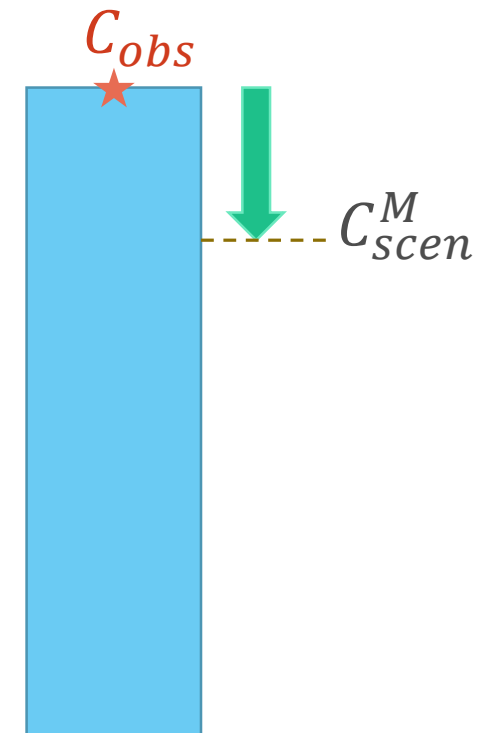
➤ What is the order of magnitude of differences? How to evaluate these differences? Which indicators?

➤ Can we explain the differences, what are the main drivers?

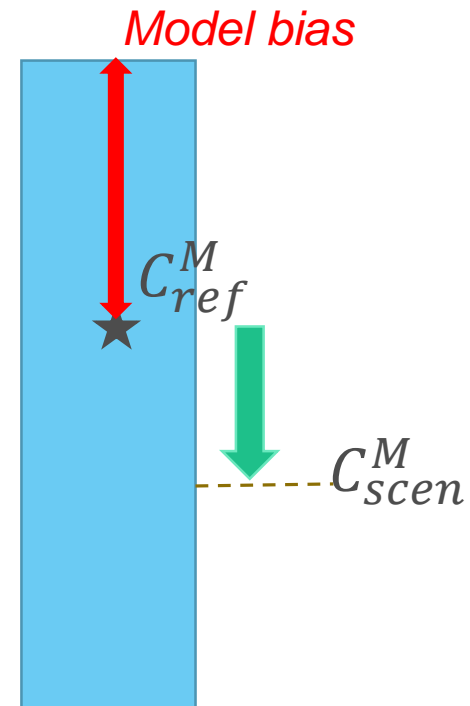
FAIRMODE CT9

- Many intercomparison exercises of air quality models
- Willing to have a long term intercomparison platform to continually assess the response of model
- No recent exercises to assess the capacity of models to simulate “delta” (Formerly CityDelta, EURODELTA)
- A Concentration Delta for model M can be applied to an observation C_{obs} to evaluate a scenarios based on ‘bc’ reference and ‘scen’ simulations:
 - ❑ Absolute (for O3?): $C_{scen} = C_{obs} + (C_{scen}^M - C_{bc}^M)$
 - ❑ Relative (for NO2 or PM?): $C_{scen} = C_{obs} \times (C_{scen}^M - C_{bc}^M) / C_{bc}^M$
 - ❑ **Techniques often used but rarely assessed**

Obs. based method



Mod. only based method

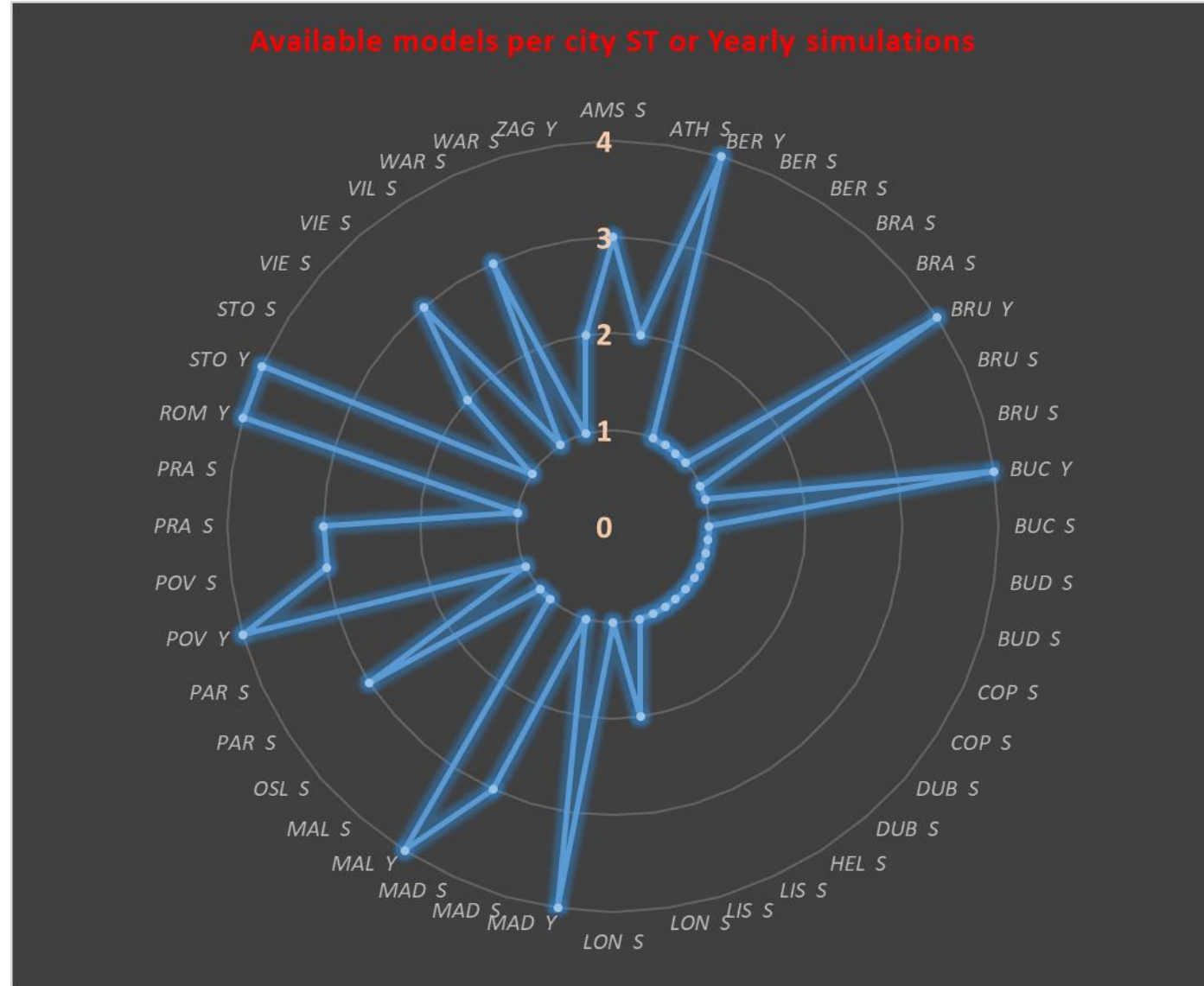


FAIRMODE CT9

- Models have different behaviours between each other because:
 - ❑ Different processes (chemistry and physics) ... magic parameters 😊
 - ❑ Different settings like resolutions (vertical, horizontal)
 - ❑ Different input data
 - Emissions
 - Meteorology
 - Boundary conditions
- Even if sharing the “same” inputs, models remain different...
 - ❑ Remaining meteorological diagnosed data (diffusion, PBL)
 - ❑ Emission pre-processing (spatial/temporal redistribution, VOC split, PM split)
 - ❑ Management of interpolations, grids
 - ❑ Others like : user dependency, compilers...

CT9 Database

- JRC box
 - ❑ 10 Participants
 - ❑ 15 Models
 - ❑ Many cities and regions are covered by one or more models
 - ❑ Number of DBMC (Concentration) files > 5000, Resol: HL, DL, YL
 - ❑ Number of DBME (Emissions) files = 170, Resol: YL, ML
- Box will be moved to an other place this month
- Access: Contact coordinators



IDL Vizualisation tool

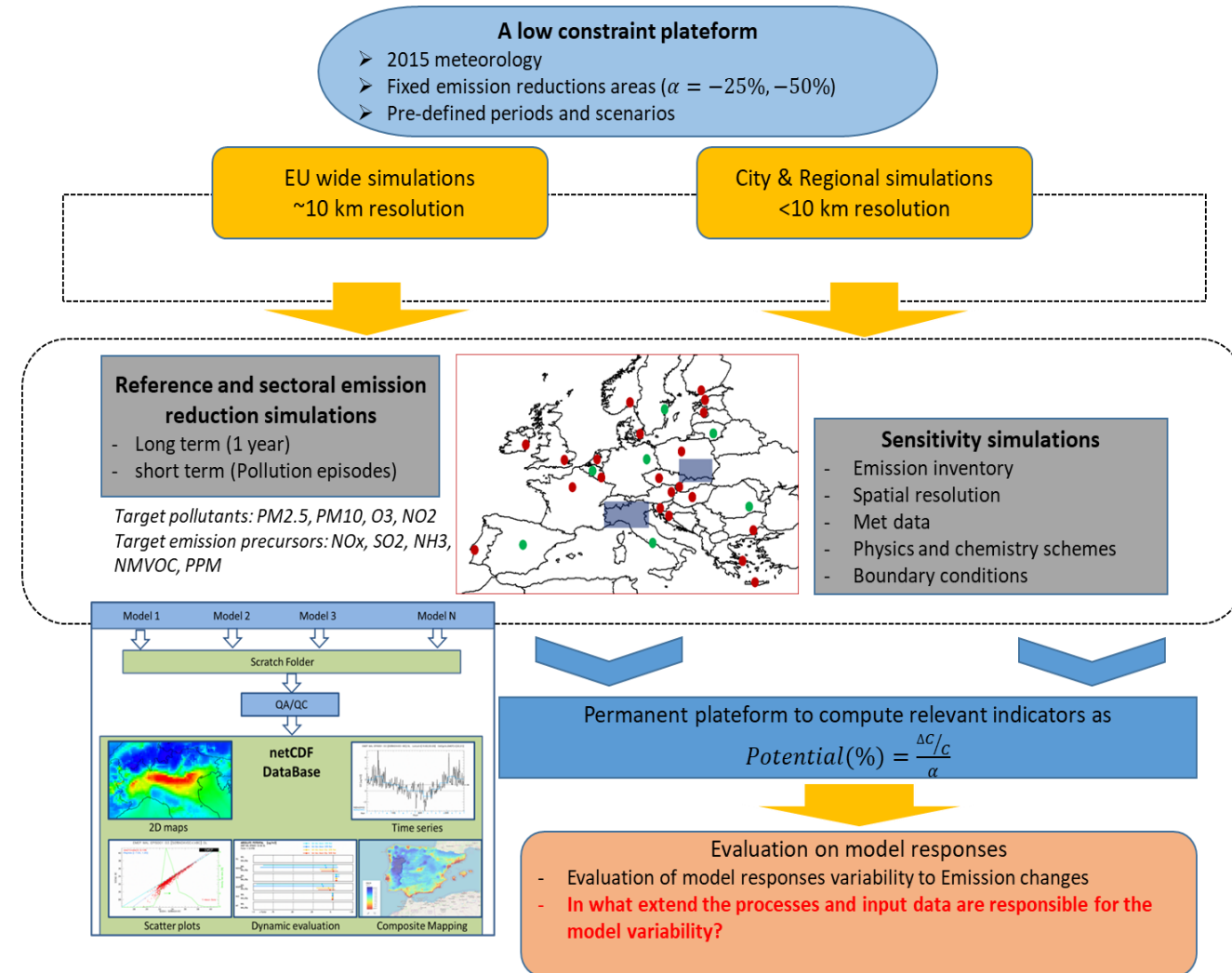
- 2D maps
- Scatter Plots
- Dynamic Evaluation diagrams
- Tests for compliance with the CT9 standards
- Link to Composite Mapping
- Freely downloadable
- Easy installation
- Easy to use
- No license

Models and teams involved

Team name - Country	Model Name	Emission Inventory, resolution, date
JRC (EU)	EMEP	EDGAR V5.0, 2015
JRC (EU)	EMEP	CAMS V2.2.1, 2015
JRC (EU)	EMEP	EMEP - GNFR, 2015
JRC (EU)	EMEP	CAMS REG V4.2 + Condensables 2015
JRC (EU)	WRF-Chem	EDGAR V5.0, 2015
ZAMG (AT)	WRF-Chem	CAMS-REG 2015
Met Norway (NO)	EMEP	EMEP, 0.1x0.1, 2015
Met Norway (NO)	EMEP + uEMEP	EMEP, 0.1x0.1, 2015
Cyl (CY)	WRF-Chem	EDGAR V5.0, 0.1° x 0.1°, 2015
NKUA (GR)	WRF-Chem	EDGAR HTAP, 2010
DHMZ (HR)	ADMS-Urban	Croatian National Emission Inventory for Zagreb
DHMZ (HR)	LOTOS-EUROS	CAMS-AP-v2.2.1 2015
LMD/IPSL (FR)	WRF-CHIMEREv2020r1	CAMS REG V4.2 2015
UH-CACP (UK)	WRF-CMAQ	EDGAR V5.0, 2015
CIEMAT (ES)	IFS-CHIMEREv2017r4	EMEP + NEI, 2015
ENEA (IT)	WRF-MINNI	ISPRA Italian national inventory 2015
UNIBS (IT)	WRF-CAMx	INEMAR 2015+EMEP
IRCELINE (BE)	CHIMERE + RIO + ATMOSTREET	Local inventories

The overall framework

- Short term (ST) on episodes (PM10, NO2, O3)
 - ❑ Emissions reduced only during the episodes from 00:00 to 23:00
- Long term (LT) simulations (PM10, NO2, O3)
 - ❑ Emissions reduced the whole year
- Two reductions so far:
 - ❑ 25% and 50% from a base case (BC)
- Reduced species depends on target pollutants
 - ❑ **PM10**: PPM, NOx, VOC, NH3, SO2, ALL
 - ❑ **Ozone**: NOx, VOC, ALL
 - ❑ All together or separately
- Notation:
 - ❑ **POL#PRE**: Pollutant concentrations POL for an emission reduction of precursor PRE
 - ❑ Ex. **PAR014** : Paris episode 014



Indicators I/II

- Indicators defined for a single pollutant reductions
- The **Absolute Potential** is defined as the reduction in $\mu\text{g}/\text{m}^3$ scaled by the reduction α of the scenario (25 or 50%) of a precursor from base case BC
 - ❑ $API = (C - C^{BC}) / (\alpha \times C^{BC})$ ($API \times \alpha$ is the delta of concentrations)
- The **Relative Potential** is defined as the reduction in % scaled by the reduction α of the scenario (25 or 50%) of precursor n from base case BC and by the BC concentrations.
 - ❑ $RPI = (C - C^{BC}) / (\alpha \times C^{BC})$
- The **Absolute Potency** in $\mu\text{g}/\text{m}^3/(\text{ton}/\text{day})$ is defined as the derivative of the concentration with respect to the emissions density E of a precursor or in other words the rate with which the concentrations (C) will change as a result of an emission density E
 - ❑ $APy = (C - C^{BC}) / (\alpha \times E^{BC})$

Indicators II/II

- Absolute Potential **ApI** or **AbsPotential** directly linked to concentration delta
- Relative Potential **RPI** or **RelPotential** allows to indirectly scale the concentration delta to model peculiarities, input, settings
- Absolute Potency **APy** or **AbsPotency** is a direct scaling with absolute emissions reductions, then excluding model settings and other inputs

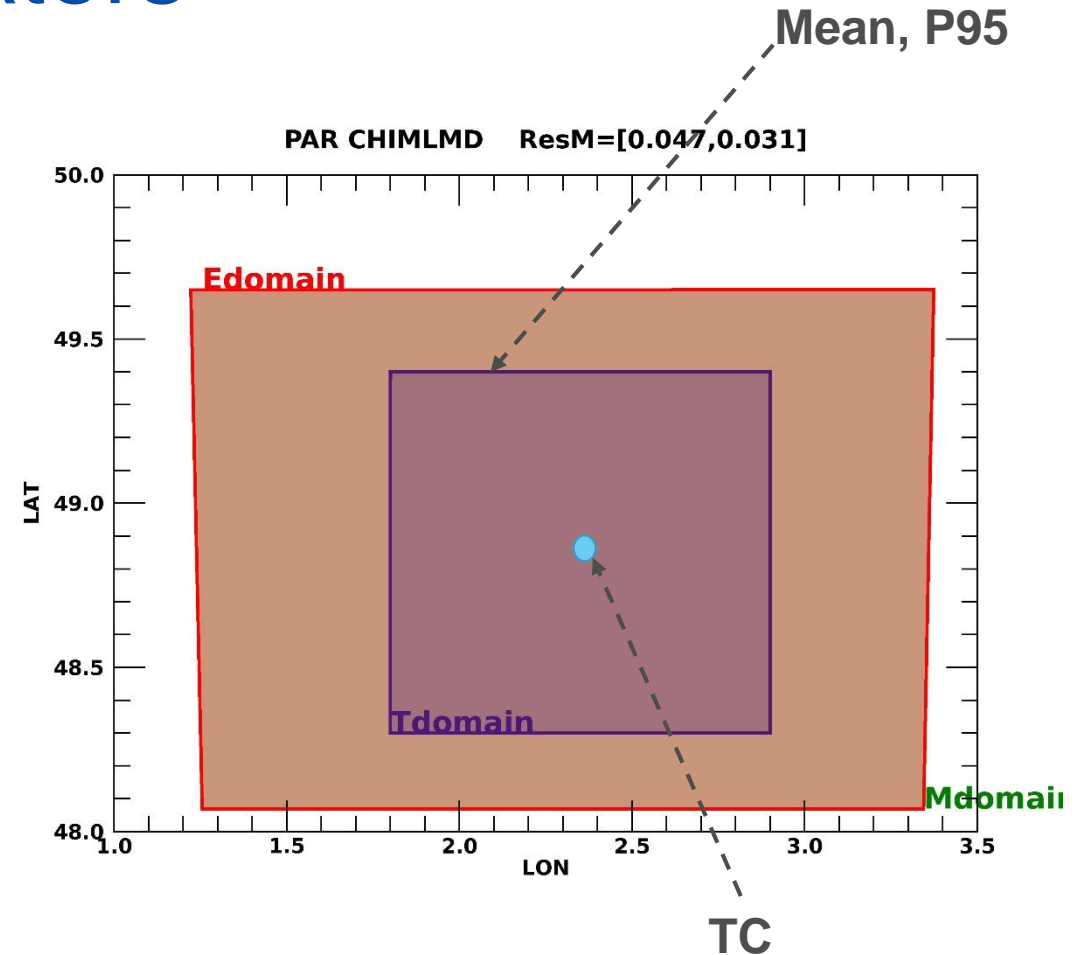
Where to calculate indicators

➤ Reminder

- ❑ Edomain: model emission reporting
- ❑ Mdomain: model concentration reporting
- ❑ Tdomain: target domain where emissions are reduced

➤ Three target locations in **Tdomain** for the concentrations common to all teams so far:

- ❑ Averaged over the Target domain Tdomain : **Mean**
- ❑ Take the values above P95th: **P95**
- ❑ Take central point of Tdomain: **TC**



Assessing the variability of indicators

➤ The Normalized Standard Deviation (NSD) is adapted

□ IND can be calculated for:

- Concentration value (Mean, Tcenter, P95) over the Target area:
 - **Mean** : average over time and Target domain of all concentrations
 - Tcenter (**TC**): center of the domain
 - **P95**: averaged values over percentile 95th

□ For a given indicator : $NSD_{IND} = \sqrt{\frac{\sum_M (IND_M - \overline{IND})^2}{(\overline{IND})^2}}$

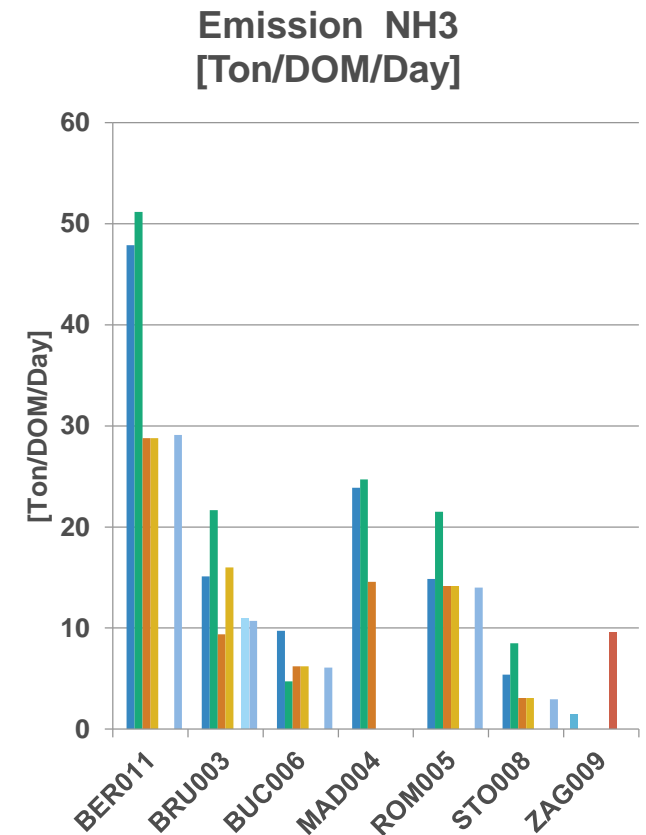
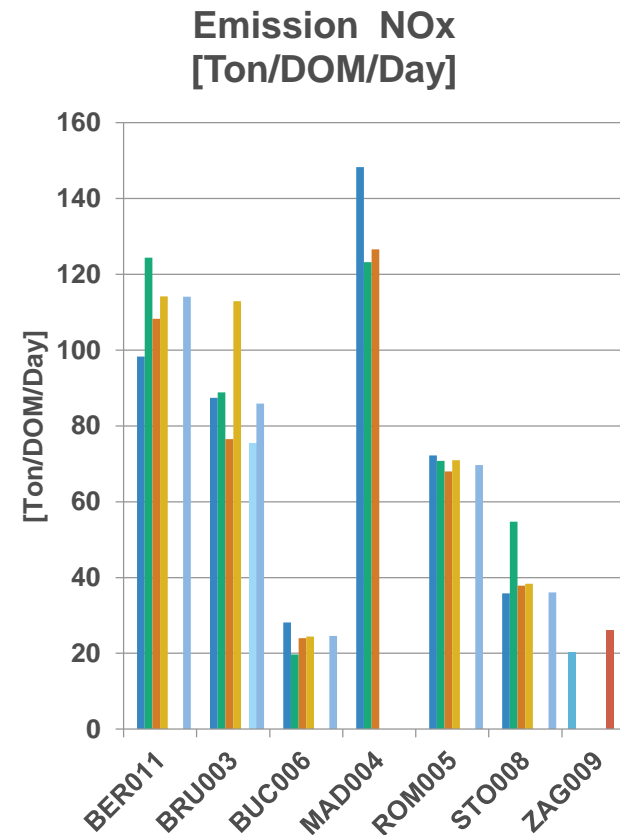
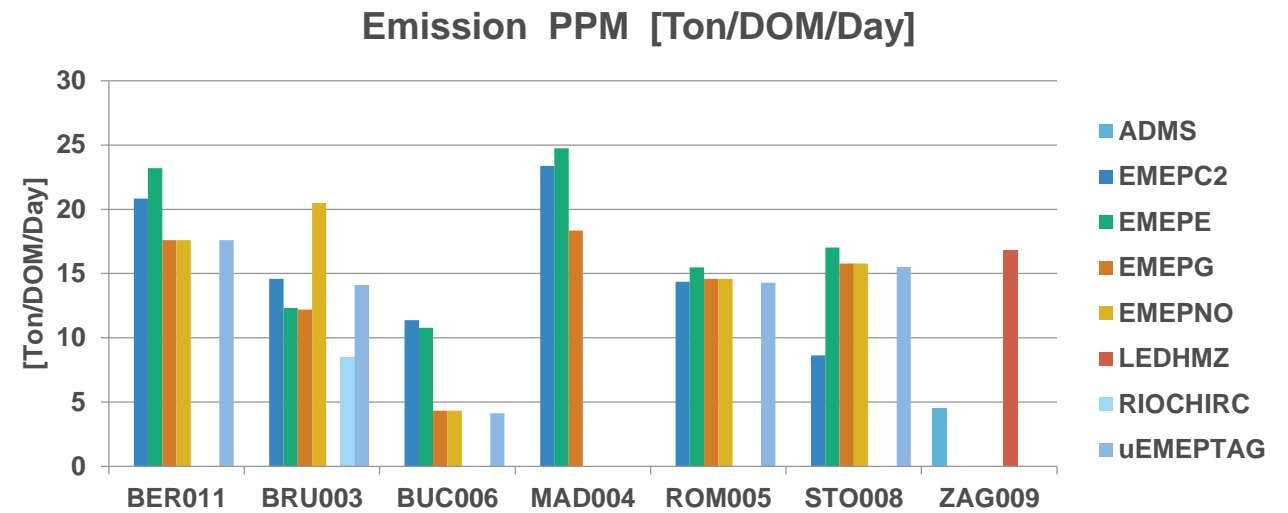
- For the list of models M in a given city/region
- \overline{IND} the average value for all models $\overline{IND} = \frac{1}{N} \sum_M IND_M$

Outline

- **Results mainly based on EMEP simulations with different settings, we wait for more runs on episodes and long term simulations**
- Focus on PM10 and O3 for the indicators mainly the RPI
- Parameter of variability based on NSD
- Linearity assessment
- Different type of concentrations:
 - ❑ **Mean** : average over time and Target domain of all concentrations
 - ❑ **Tcenter (TC)**: center of the domain
 - ❑ **P95**: averaged values over percentile 95th
- Some results on Short Term simulations
- Concluding remarks and next steps

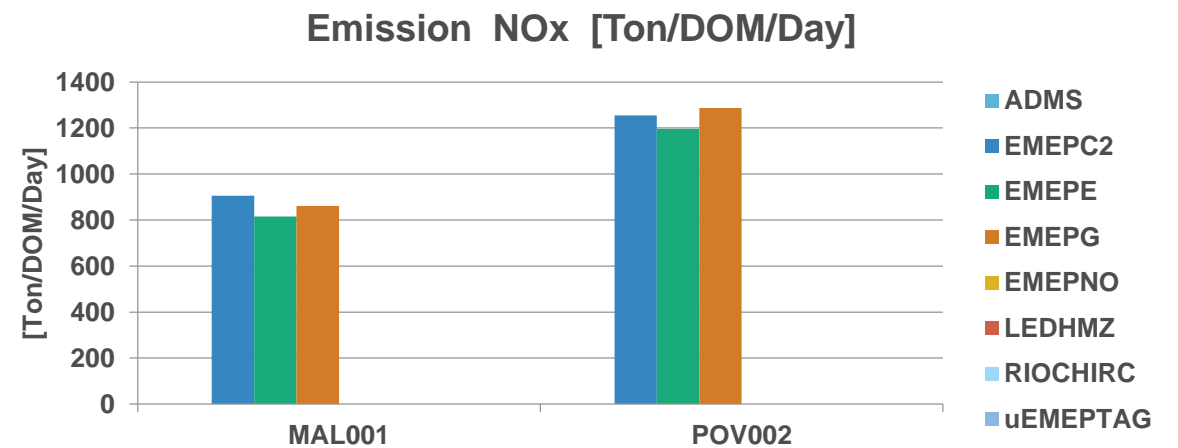
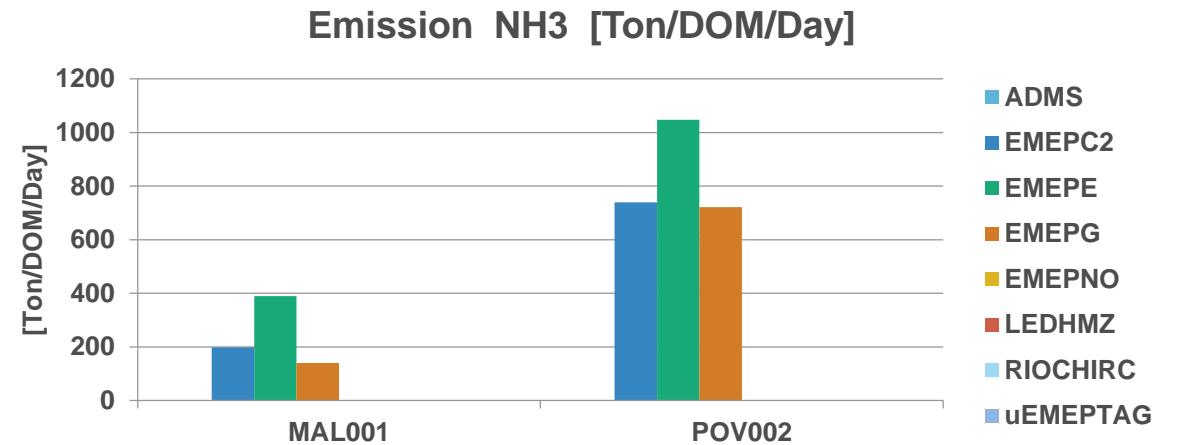
Yearly Emissions

- Large differences observed for the BC
- Less differences for NO_x emissions
- Keep in mind that NH₃ has a low molar mass compared to NO_x related species
 - Crucial for ammonium nitrate formation



Yearly Emissions

- For MAL and POV regions, some significant differences for NH3 even for similar inventories



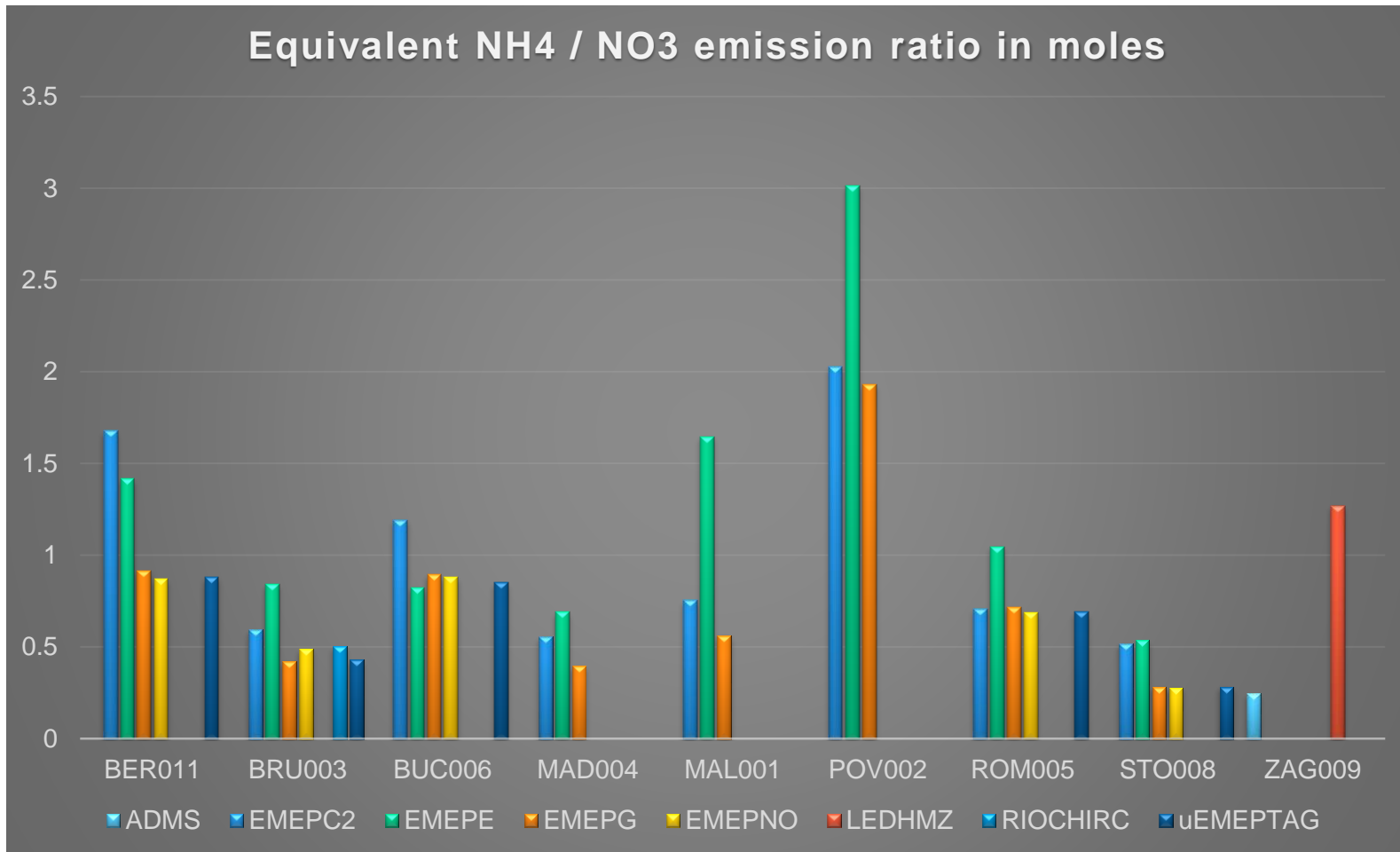
Emission ratio NH3 vs NOx

➤ NH3 + NO3 >> NH4NO3

■ Emission ratio = $\frac{63}{18} \times \frac{E_{NH3}}{E_{NOx}}$

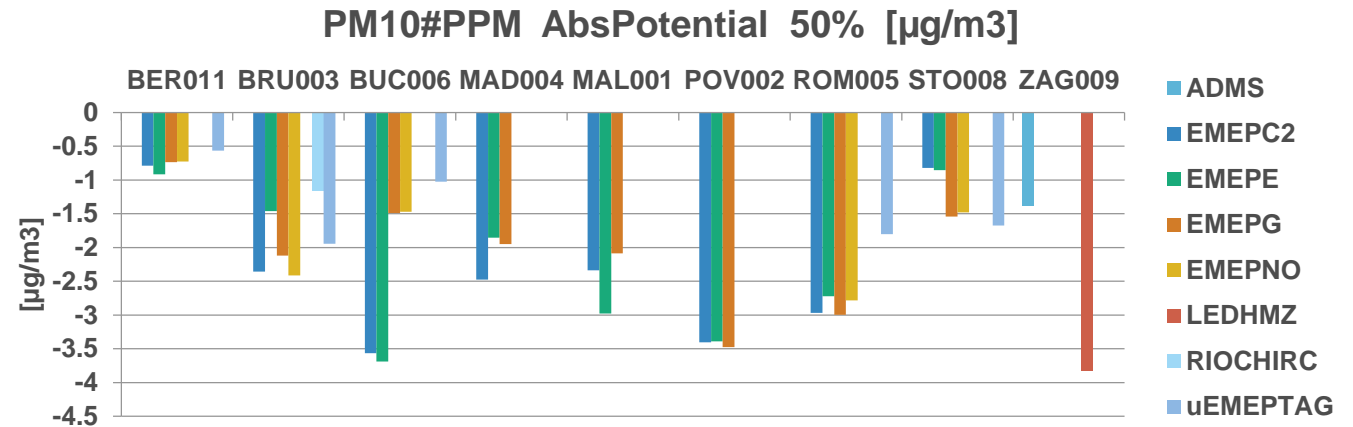
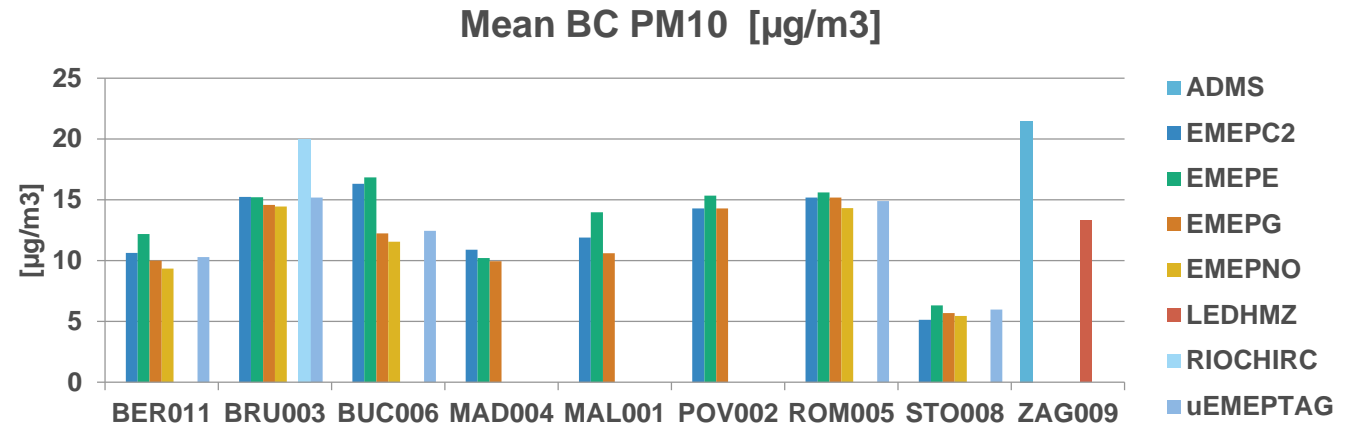
➤ An indicator of chemical regime for the BC

➤ But large spatial variability



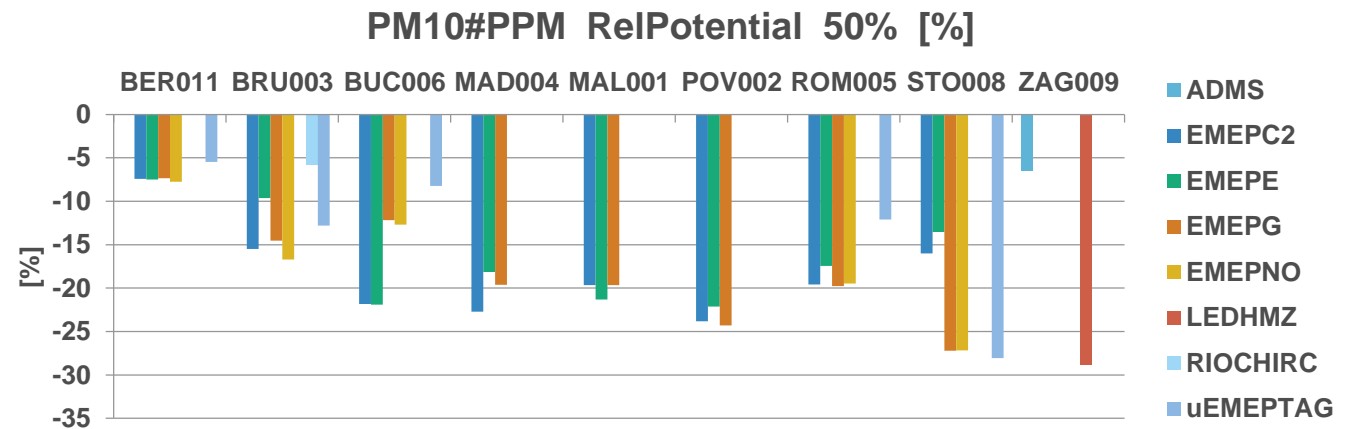
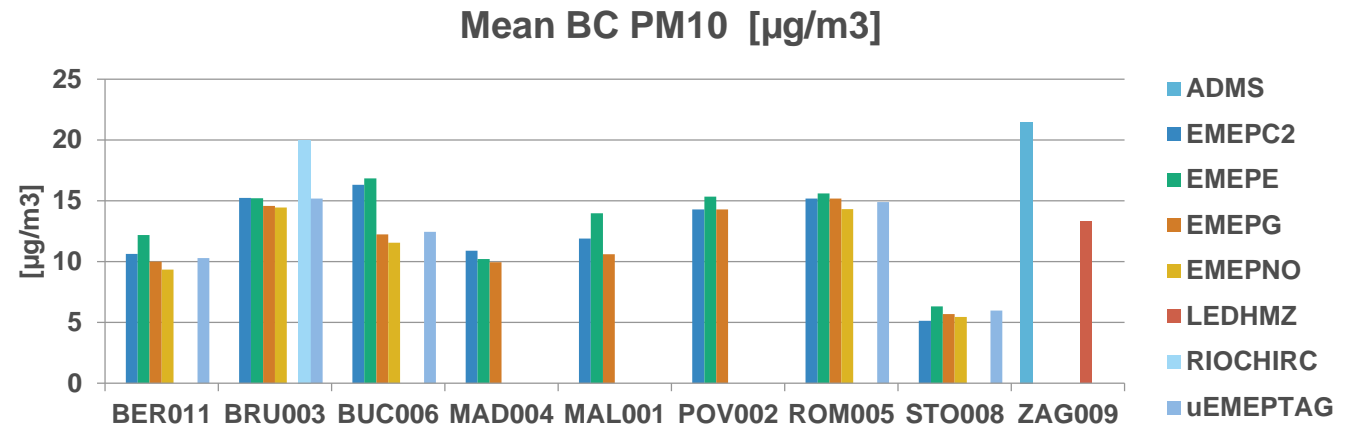
Yearly mean API – PM10

- API is the delta divided by the reduction (%)
- Negative every where (Fortunately...)



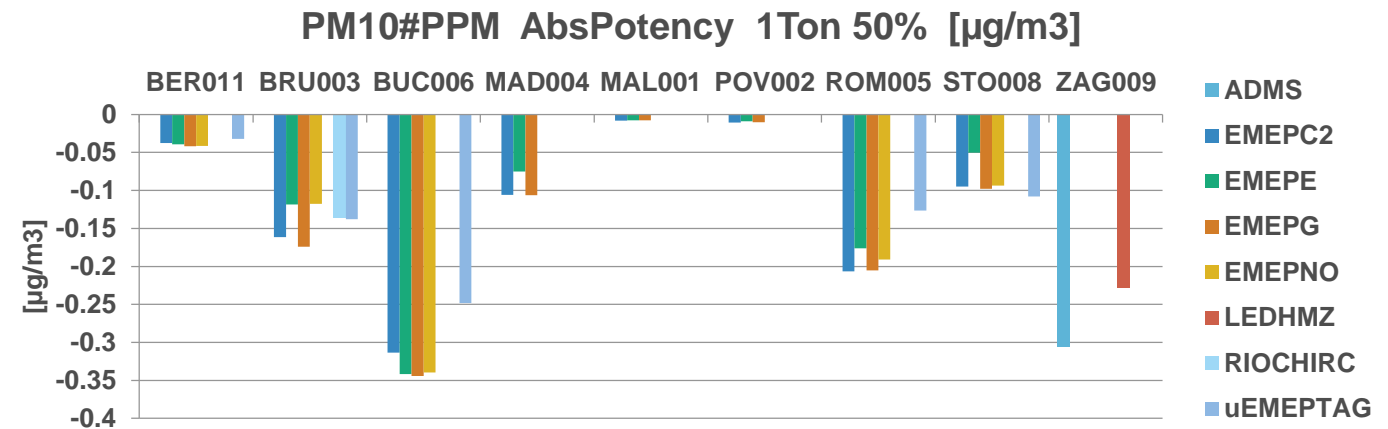
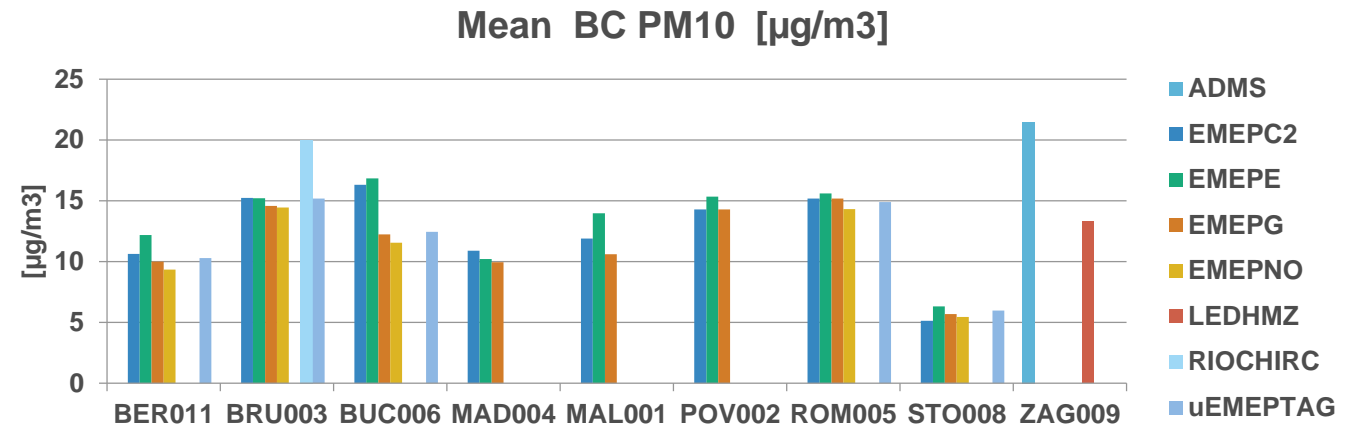
Yearly mean RPI – PM10

- Reduction of variability
- But still important in STO, a factor of 2



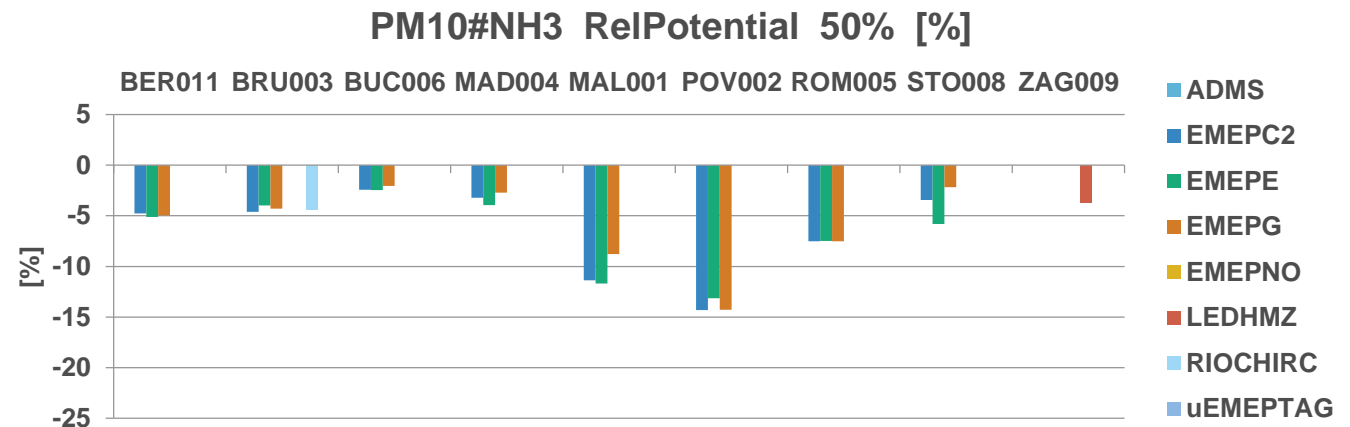
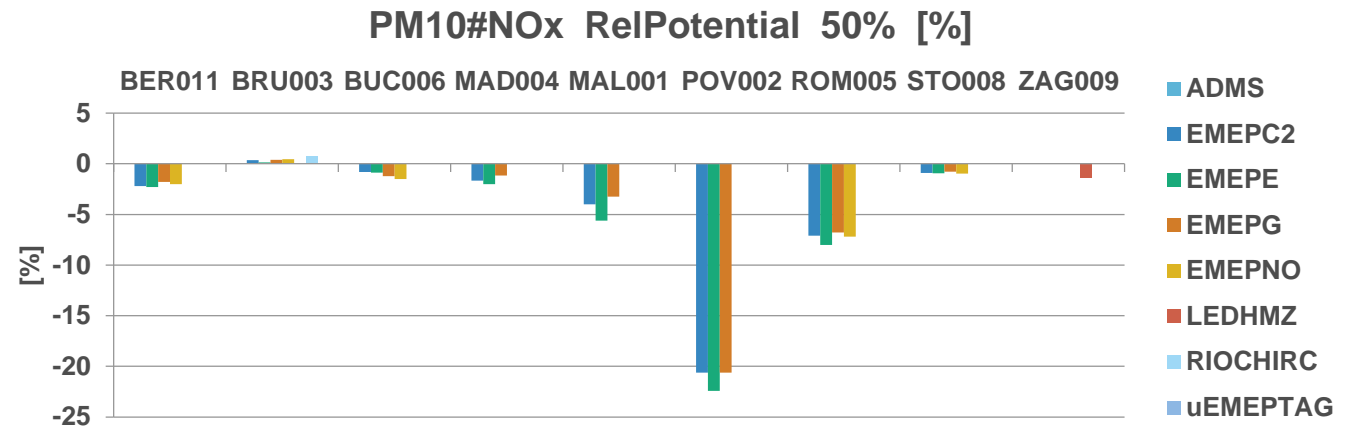
Yearly mean APy – PM10

- Delta per ton of emission reduced over the target domain



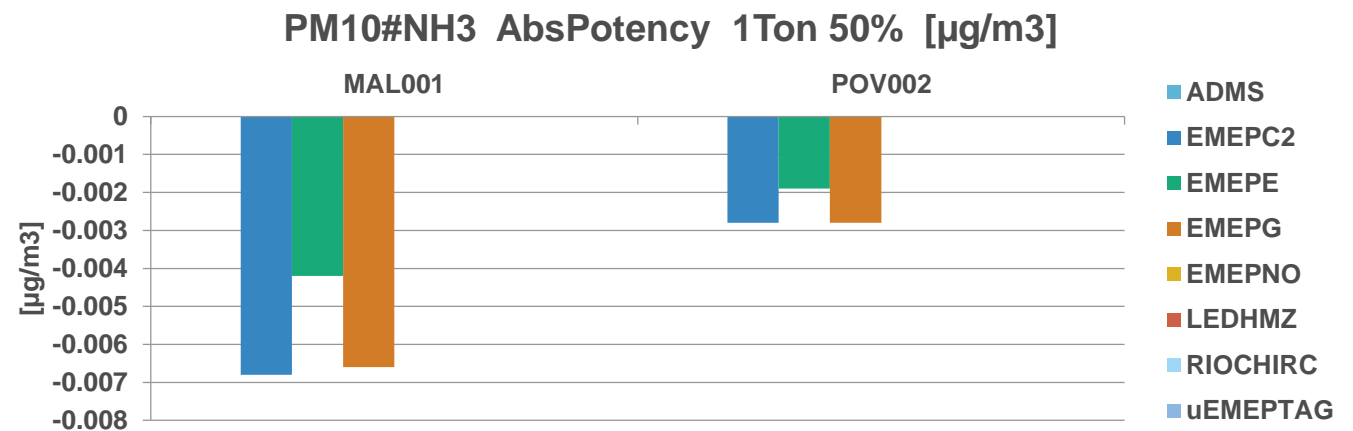
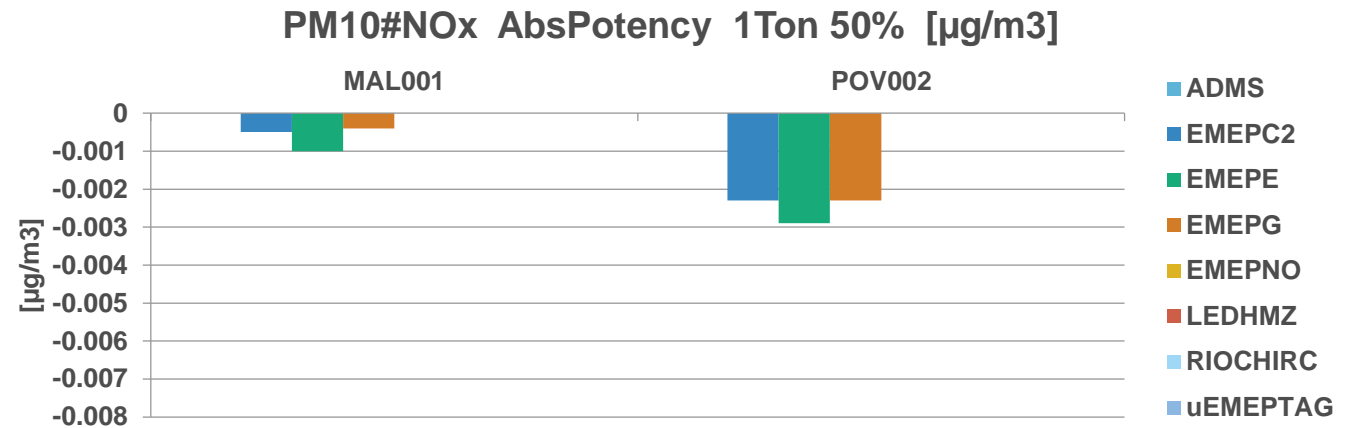
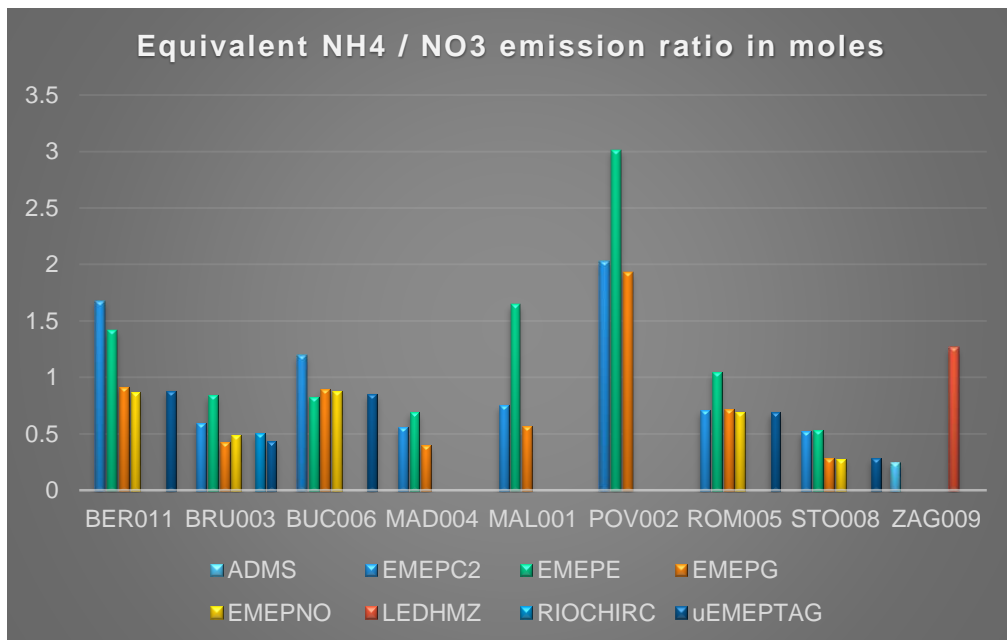
Yearly mean RPI – PM10 (NOx versus NH3 emissions)

- Slight increase of PM10 over Brussel for NOx emissions reductions
- Large reduction over POV
- In general low variability
- Large variability over STO for NH3 emission reductions



Yearly mean APy – PM10 (NOx versus NH3 emissions)

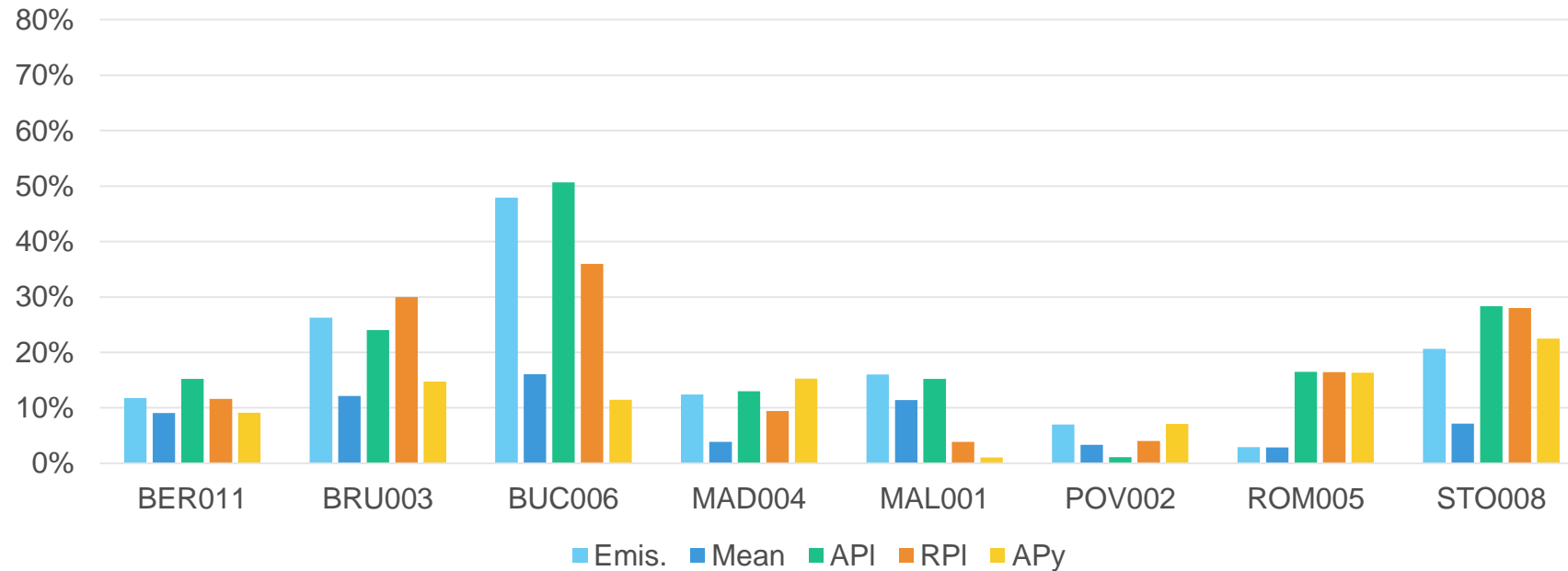
- Much more effective to reduce PM in MAL region by reducing Ammonia per ton (excess of NOy species)



Variability on PM10 LT scenarios – PM10

➤ Reminder: $NSD_{IND} = \sqrt{\frac{\sum_{m=1}^M (IND - \overline{IND})^2}{(\overline{IND})^2}}$

Variability PM10#PPM - NSD

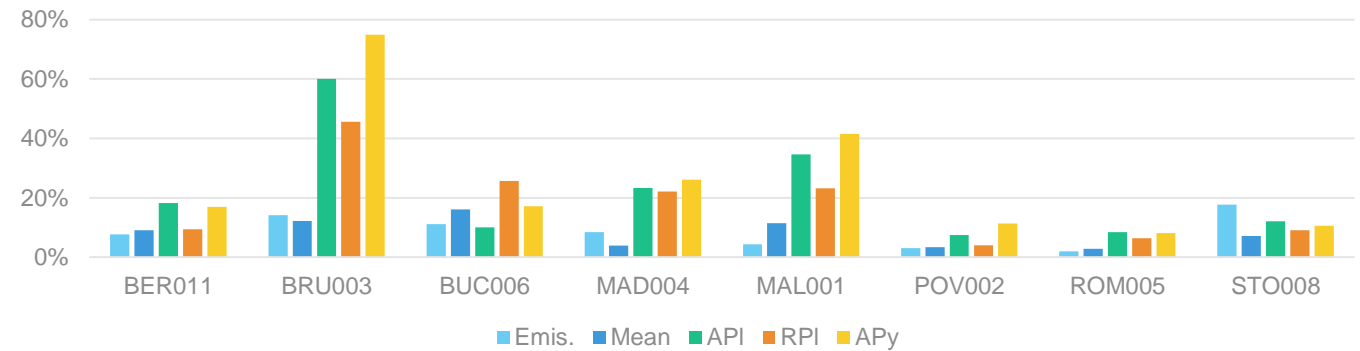


Variability on PM10 LT scenarios – PM10

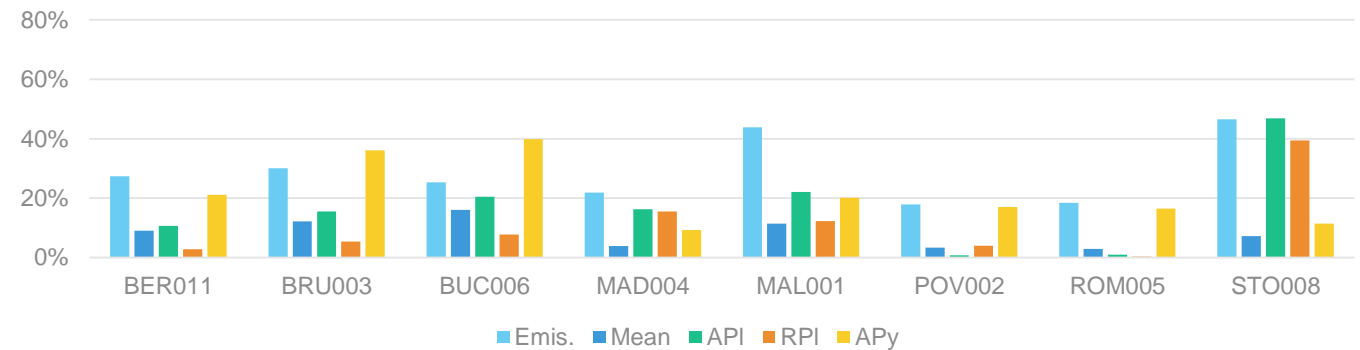
- Variability of RPI looks more important when the influence of the emission is high

$$NSD_{IND} = \sqrt{\frac{\sum_{m=1}^M (IND - \overline{IND})^2}{(\overline{IND})^2}}$$

Variability PM10#NOx - NSD

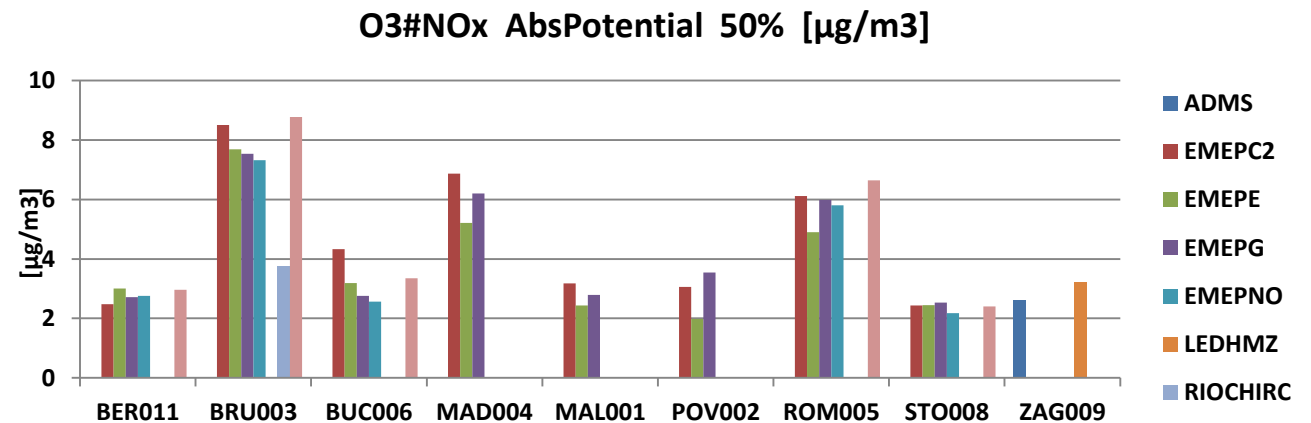
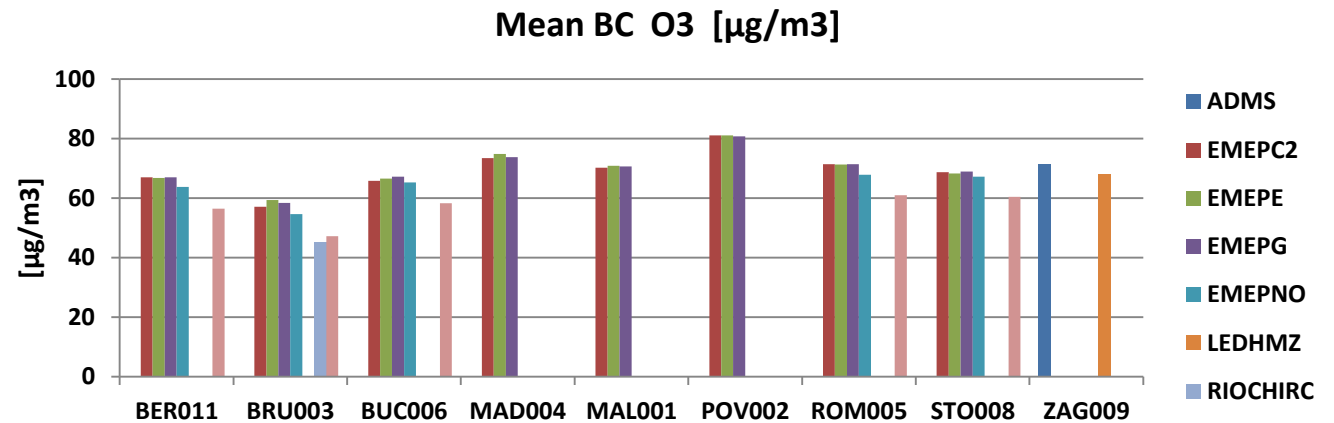


Variability PM10#NH3 - NSD



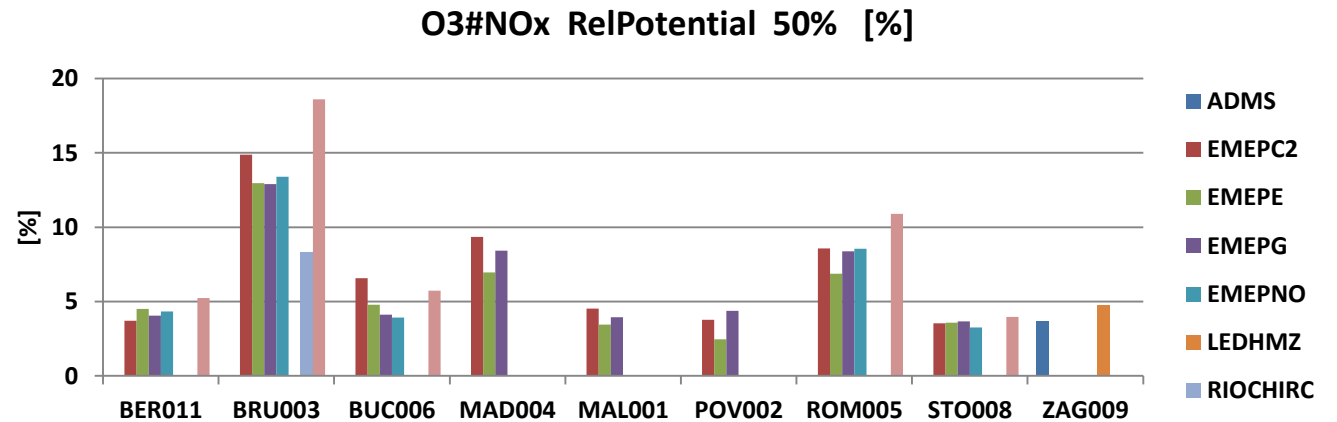
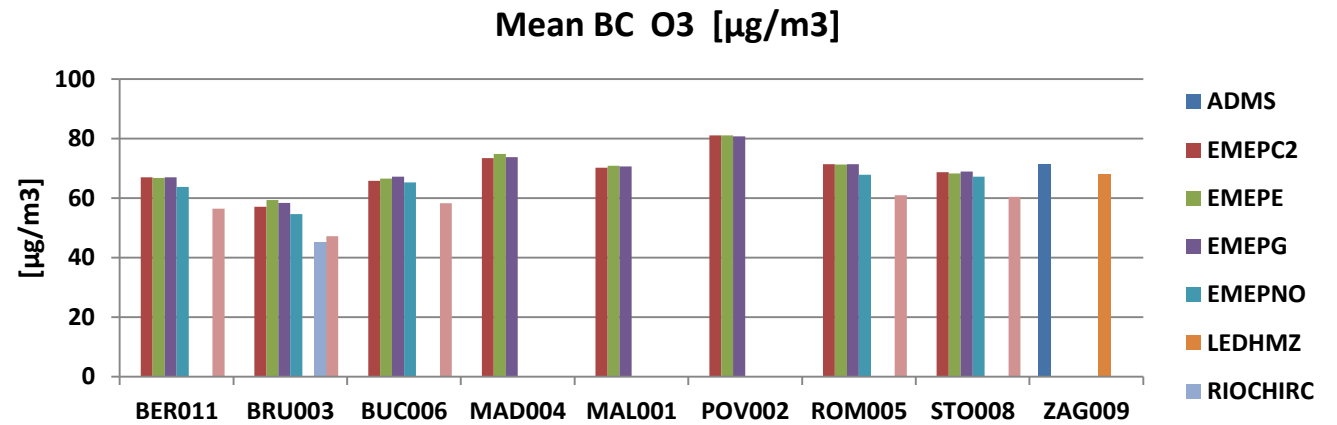
Yearly mean API – O3

- Low variability on mean concentrations
- API is the delta divided by the reduction (%)
- Positive every where due to the titration effect
- Lower over large domains like MAL and POV
- Slightly lower values for LOTOS and RIO over Brussels



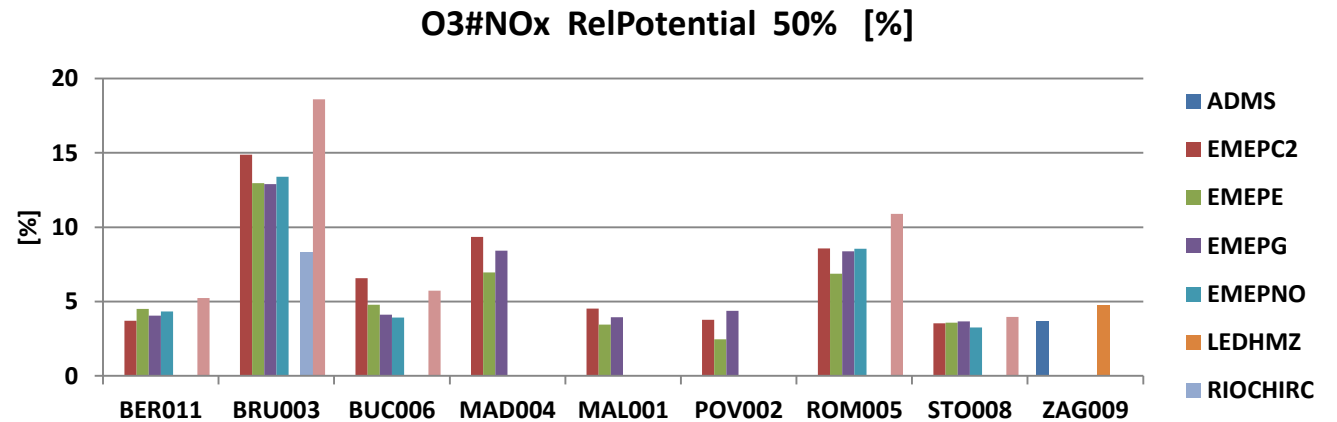
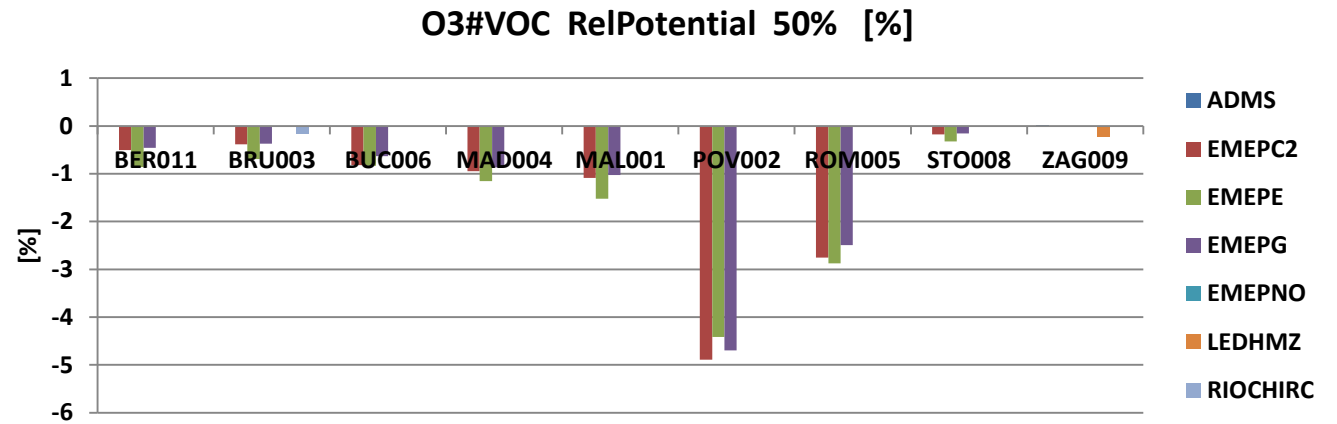
Yearly mean RPI – O3

- Low variability on mean concentrations



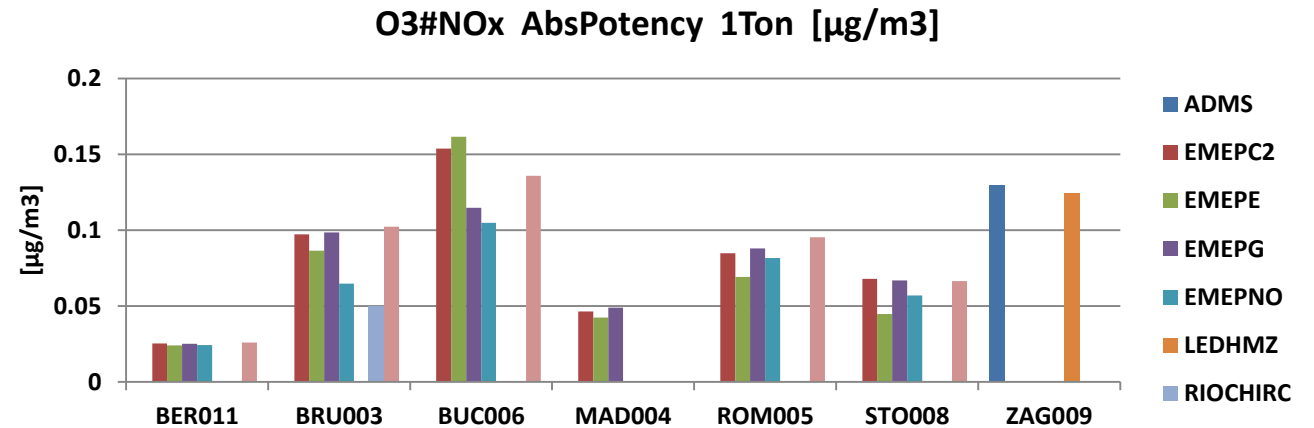
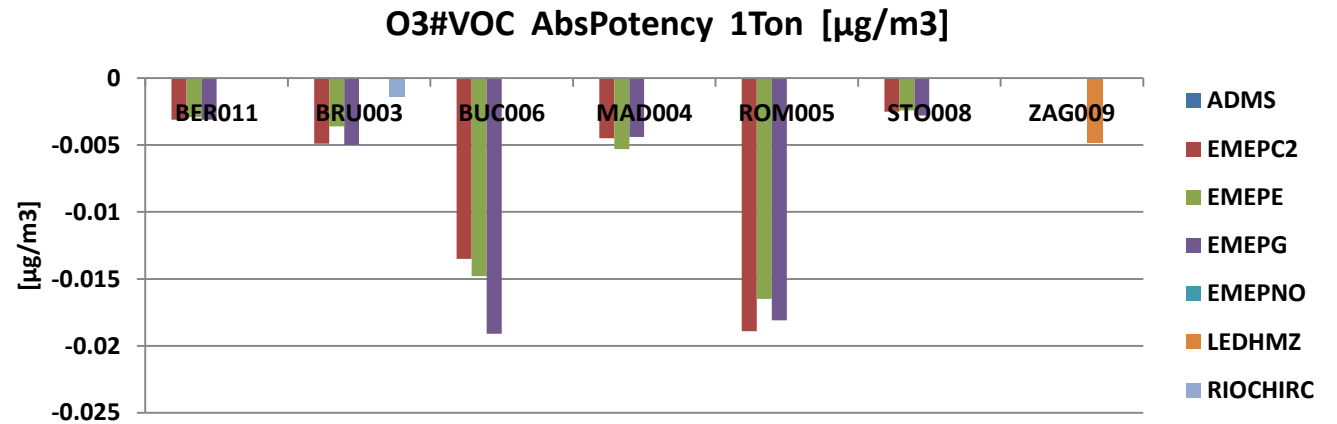
Yearly mean RPI – O3 (NOx versus VOC)

- Negative impact of VOC reduction (normal)
- Very slight increase in Zagreb for ADMS
- On absolute values lower impact of VOC emissions



Yearly mean APy – O3 (NOx versus VOC)

- Focus on cities (POV and MAL removed)
- Still larger values for NOx impact
- Very low variability in Berlin and low values
 - Other processes and input than emissions



Summary of variability

➤ Less variability on O3 BC Mean than PM10 BC Mean

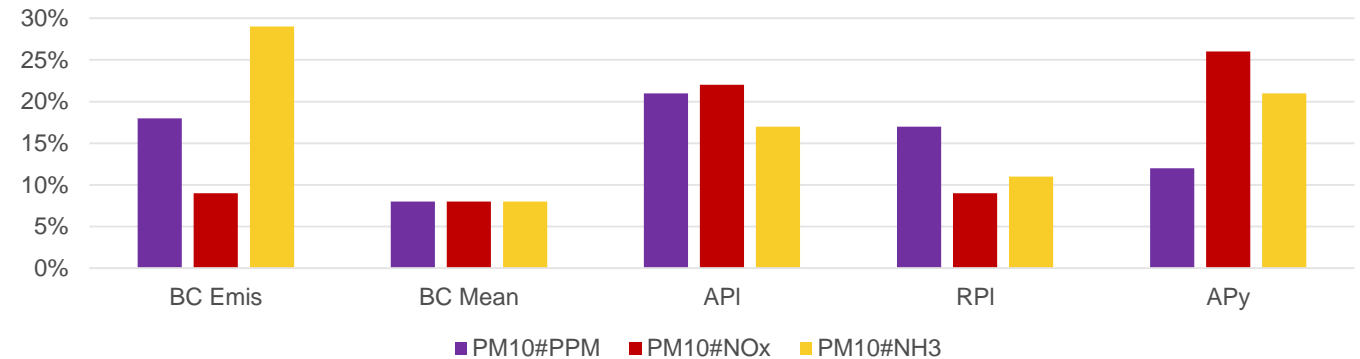
- 4 versus 8 %

➤ Variability of RPI << API (less clear for O3)

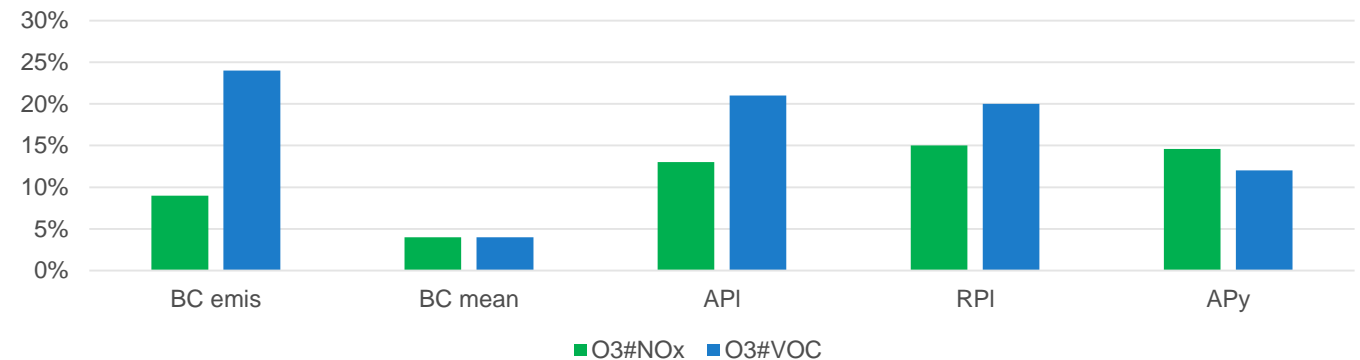
- 10 to 25% depending on the indicator

$$NSD_{IND} = \sqrt{\frac{\sum_{m=1}^M (IND - \overline{IND})^2}{(\overline{IND})^2}}$$

Average variability PM10 - NSD



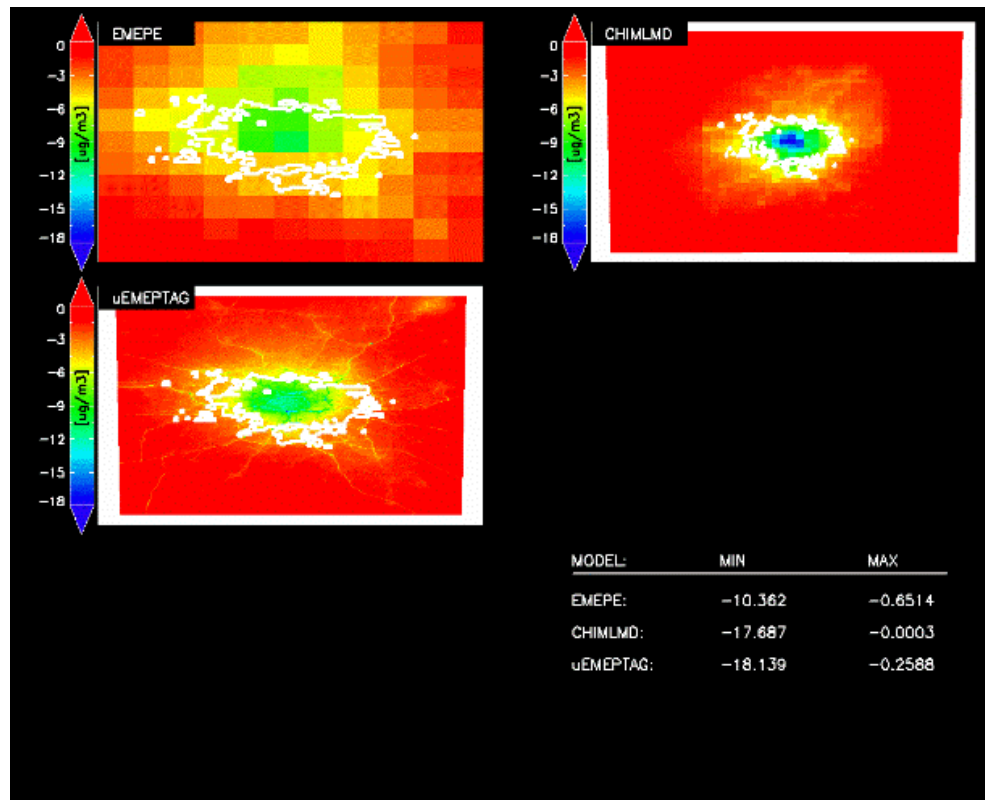
Average variability O3 - NSD



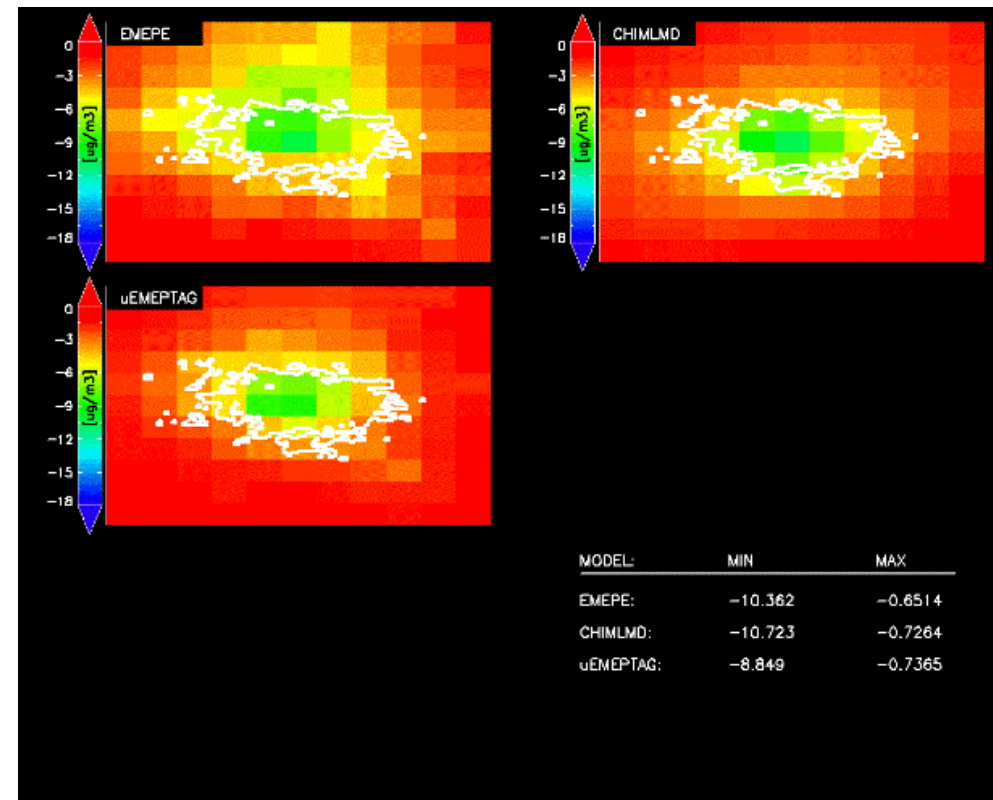
Spatial variability with 2D plots

Absolute Potential (API) 25% PM10#PPM over Berlin

Model grid

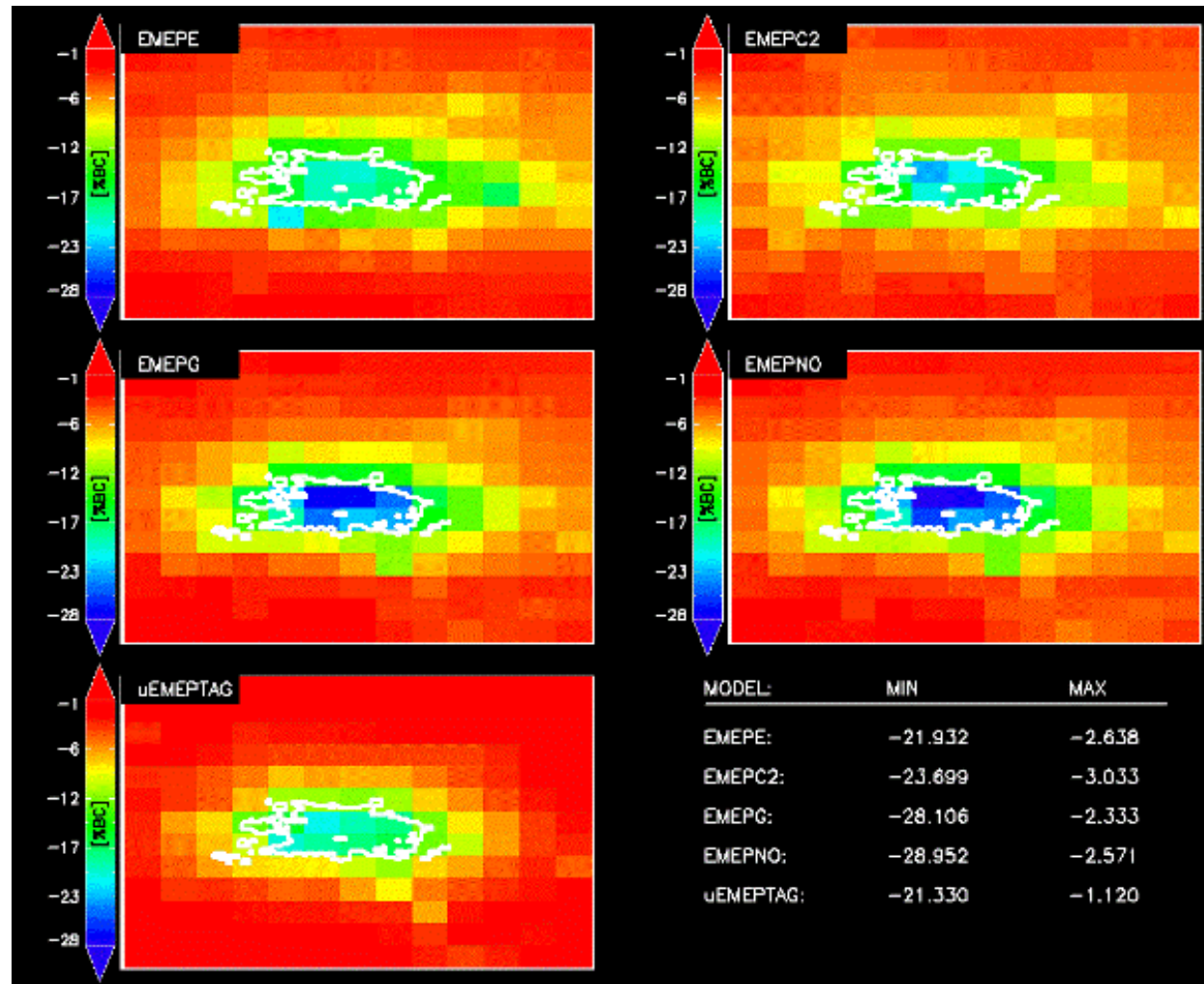


Common grid



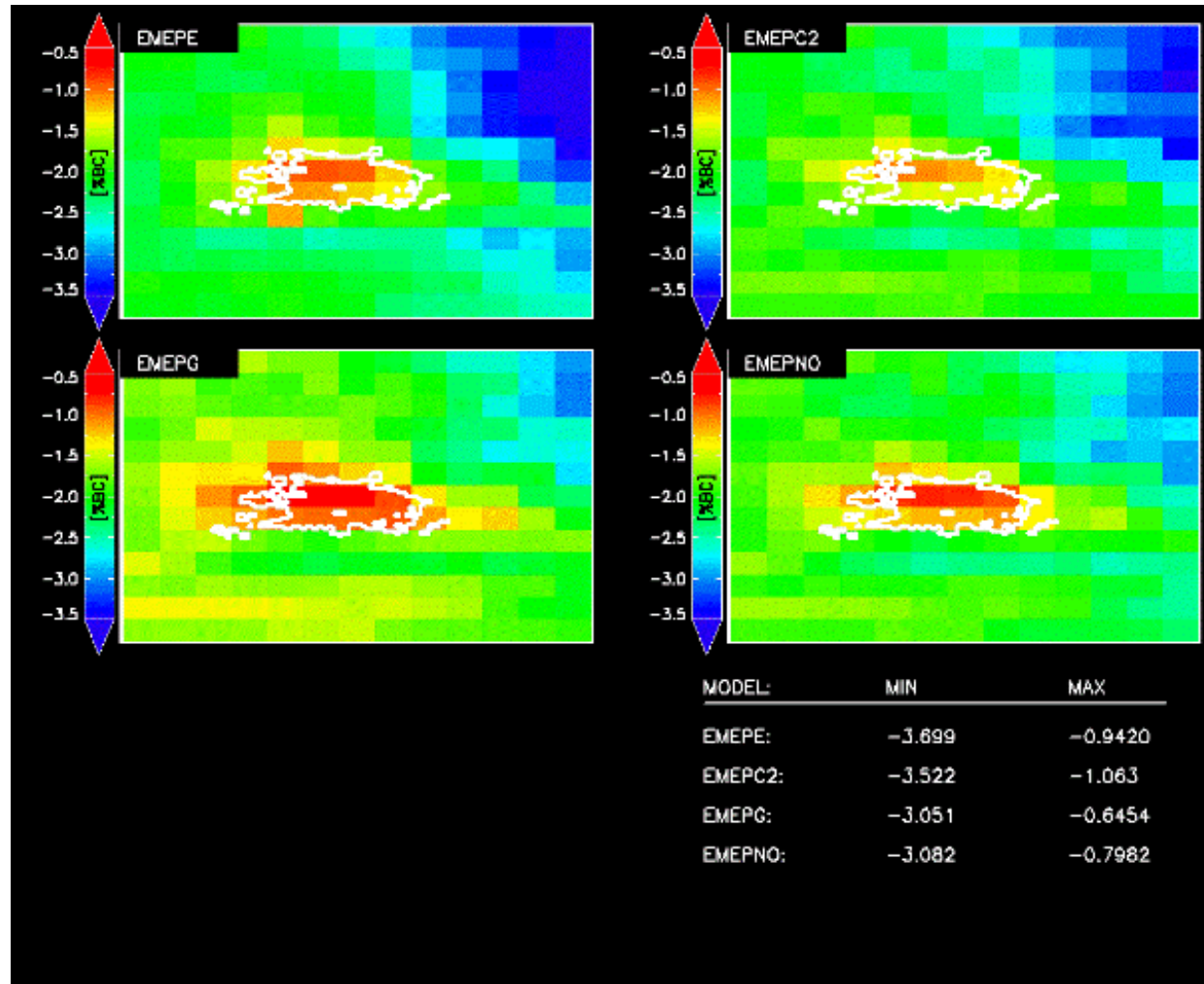
Spatial variability with 2D plots

PICT_2D_PM10_BER011_RPI_50%PPM



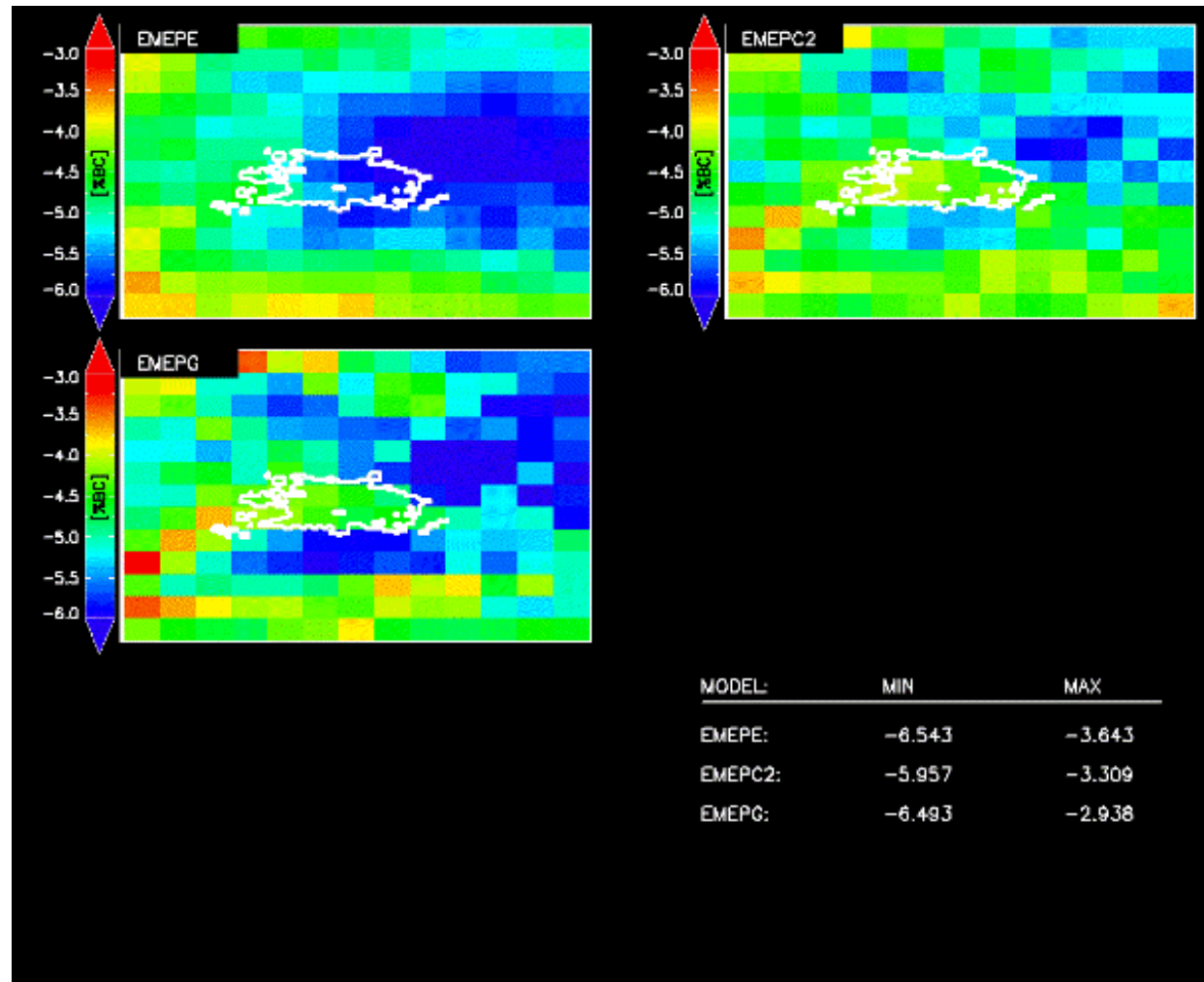
Spatial variability with 2D plots

PICT_2D_PM10_BER011_RPI_50%NOx



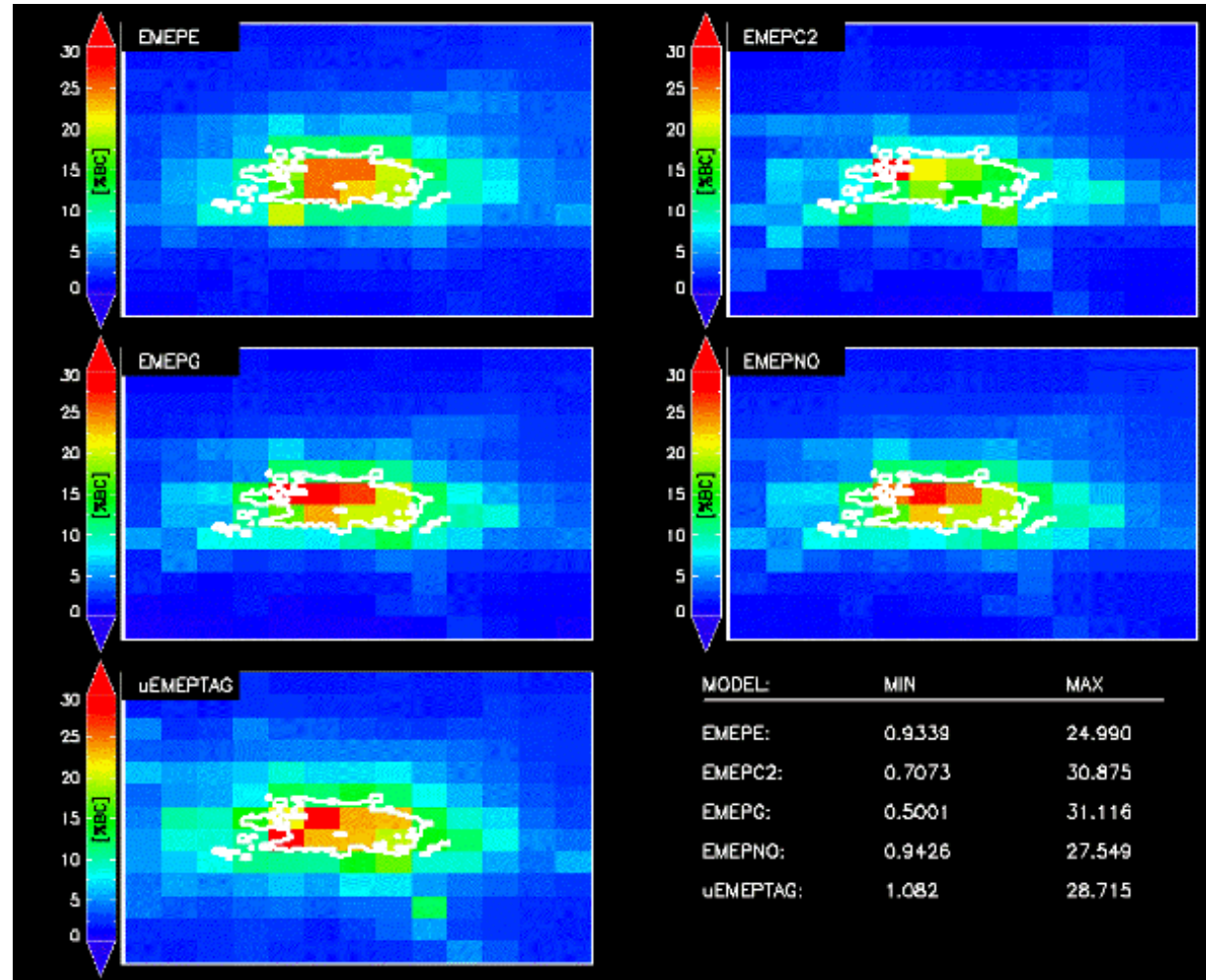
Spatial variability with 2D plots

PICT_2D_PM10_BER011_RPI_50%NOx



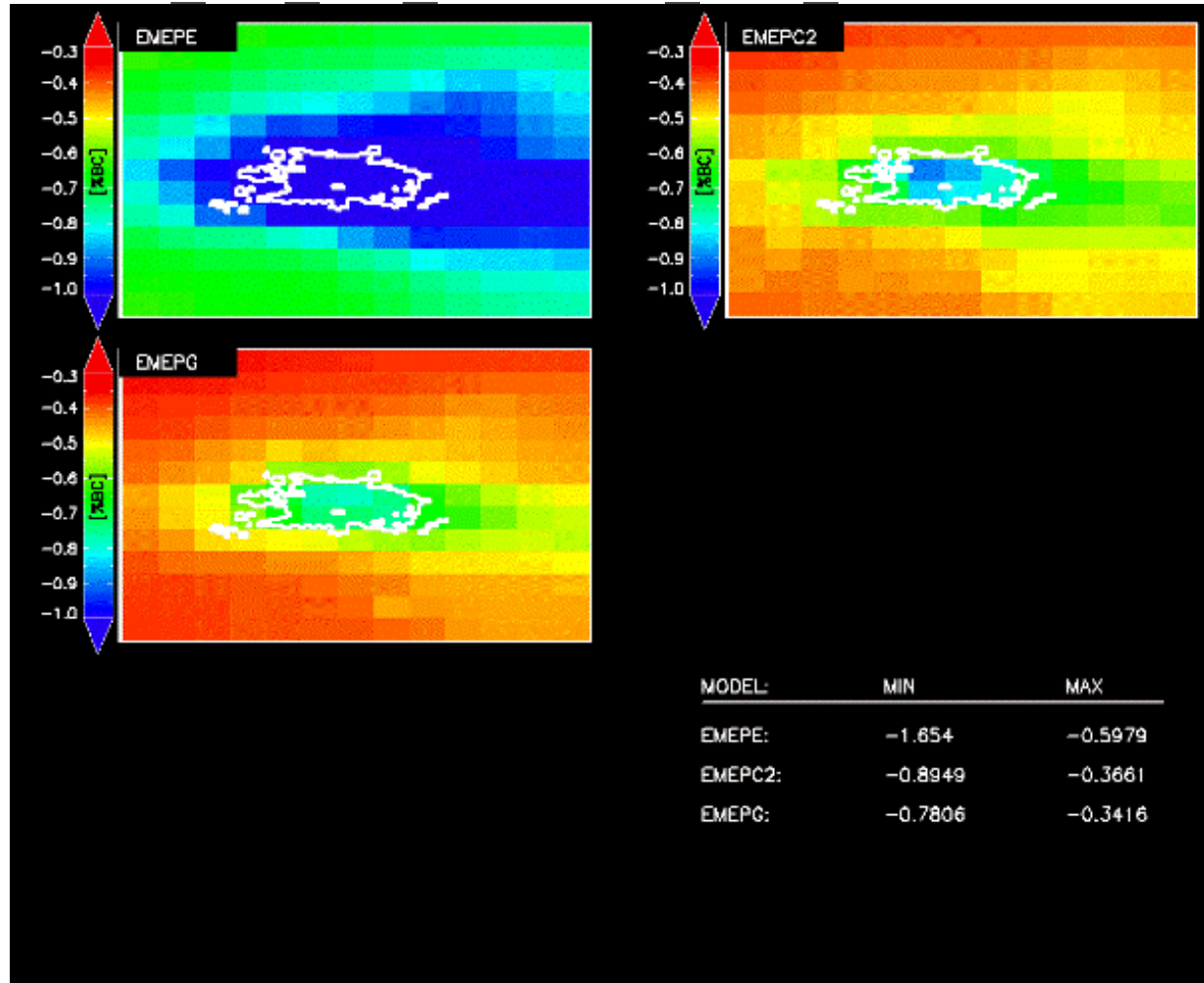
Spatial variability with 2D plots

PICT_2D_O3_BER011_RPI_50%NOx



Spatial variability with 2D plots

PICT 2D O3 BER011 RPI 50%VOC

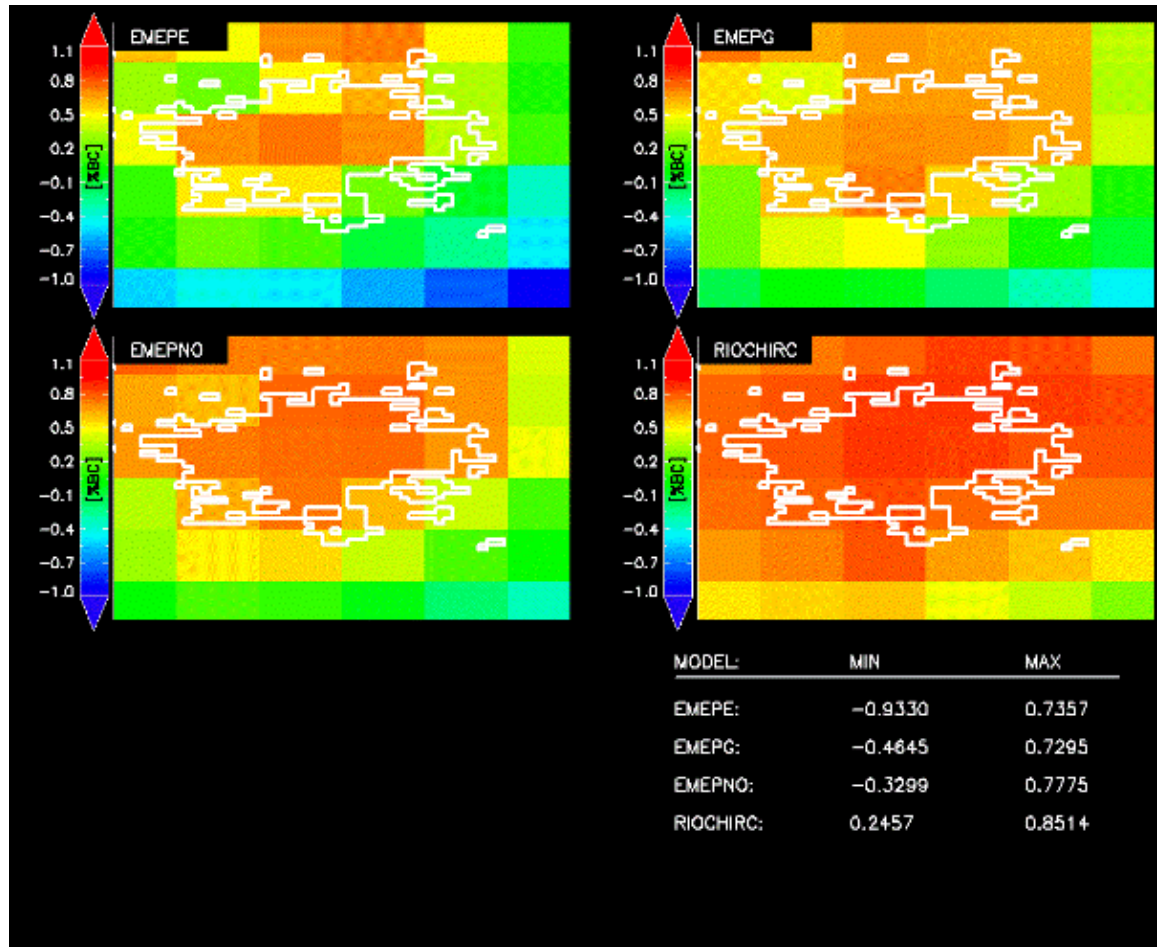


Chemical features

- Possible interactions of PPM emission reduction on O₃ through heterogeneous reactions (still in debate in the scientific community)
- NO_x emission reactions usually lead to PM reduction through the reduction of Ammonium nitrate by atmospheric reactions
 - ❑ But increasing O₃ due to titration effects lead to an enhancement of the atmospheric oxidizing capacity
 - ❑ ...then lead to more SOA formation and then a PM increase
 - (S, I)VOCs(g) + oxidants → (S, I)OVOCs(g) ↔ SOA(s)
 - ❑ SO₄ can increase also due to complex effects in aqueous chemistry with O₃ increase

Non linearities in models?

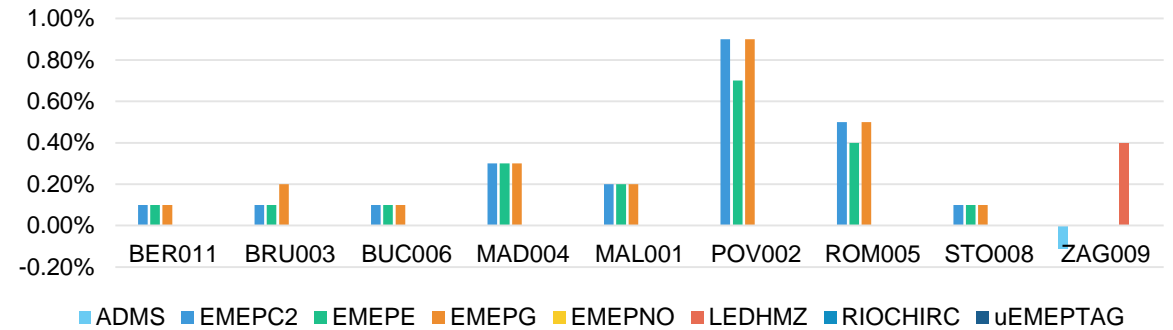
PICT_2D_PM10_BRU003_RPI_50%NOx



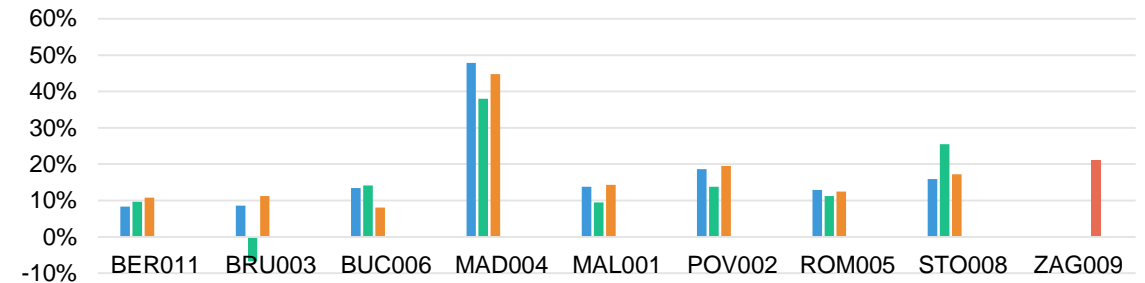
Linearities (PM10)

- Deviation % to linearity:
 - $as (AbsP50/AbsP25 - 1) \times 100$
 - “0%” means perfect linearity
- Perfect linearity for PPM, as it is considered as chemically inert (deviation < 1%)
- Usually higher efficiency when NO_x or NH₃ are reduced by 50% compared to 25%
 - *NH₃ or NO_x becoming no longer on average in excess and then being limiting in nitrate formation*

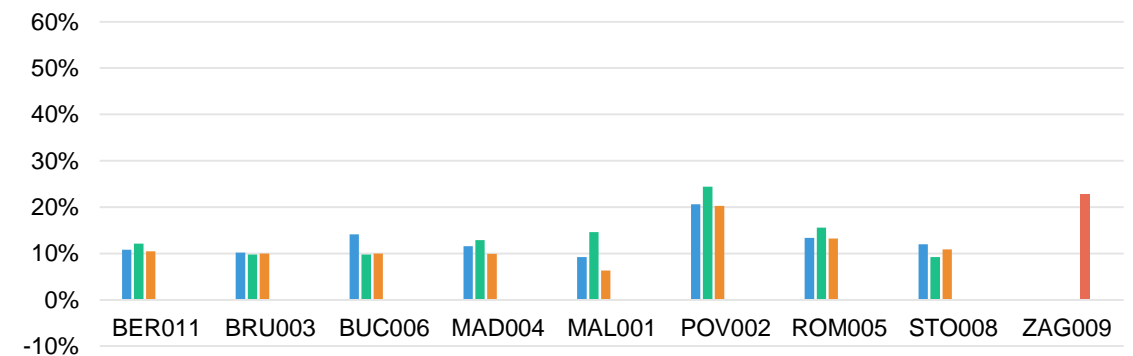
PM10#PPM AbsP50/AbsP25 deviation to linearity



PM10#NO_x AbsP50/AbsP25 deviation to Linearity

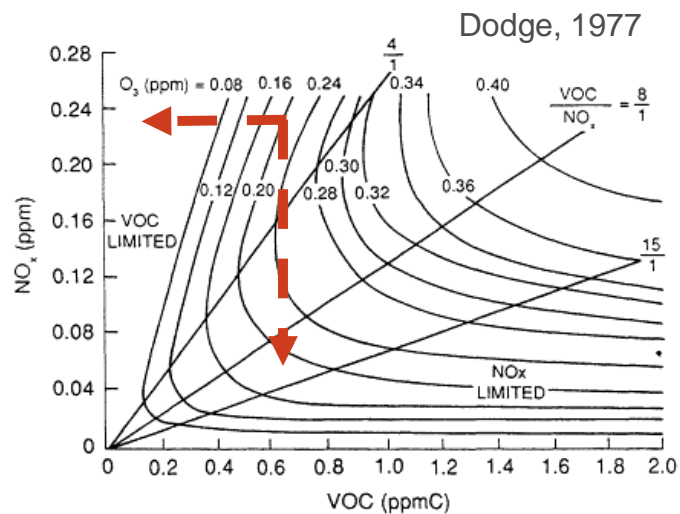


PM10#NH₃ AbsP50/AbsP25 deviation to linearity

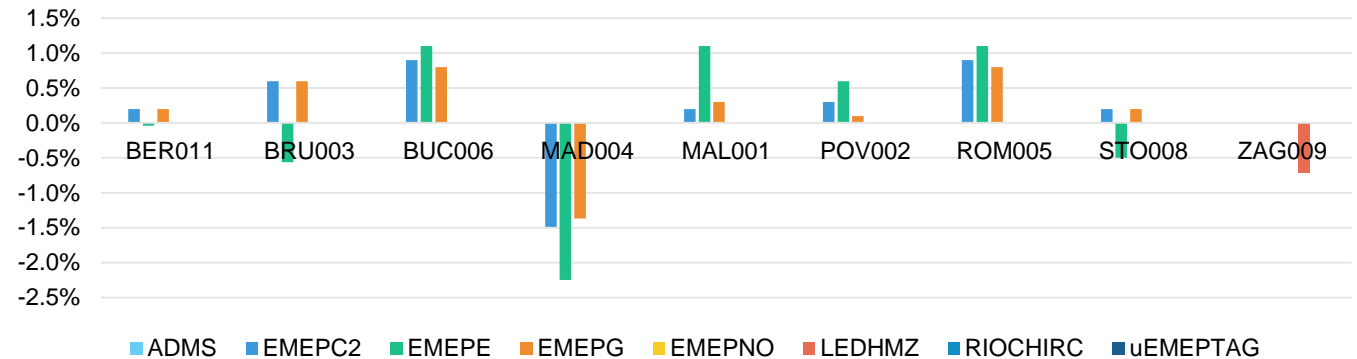


Linearities (O3)

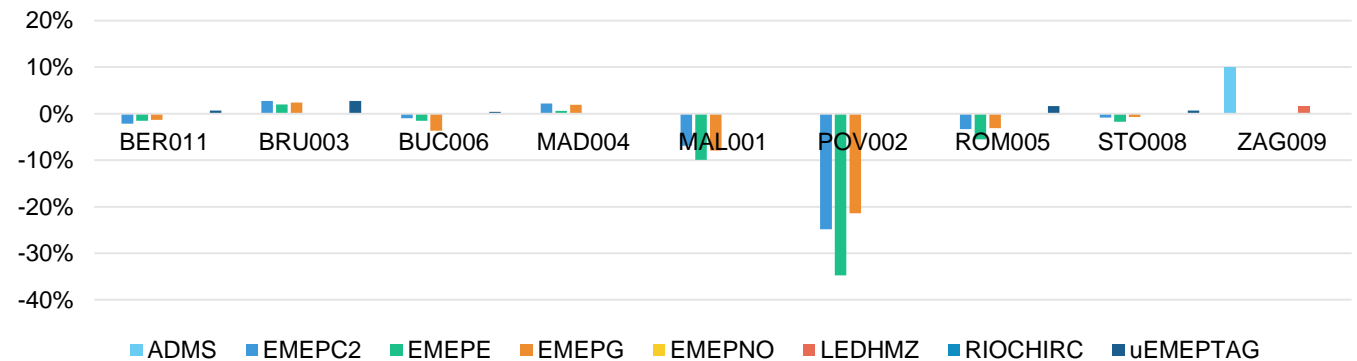
- Linearity for VOC emission reduction
- More or less linear in urban areas (VOC limited)
- Non linear for large regions (POV and MAL)



O3#VOC AbsP50/AbsP25 deviation to linearity

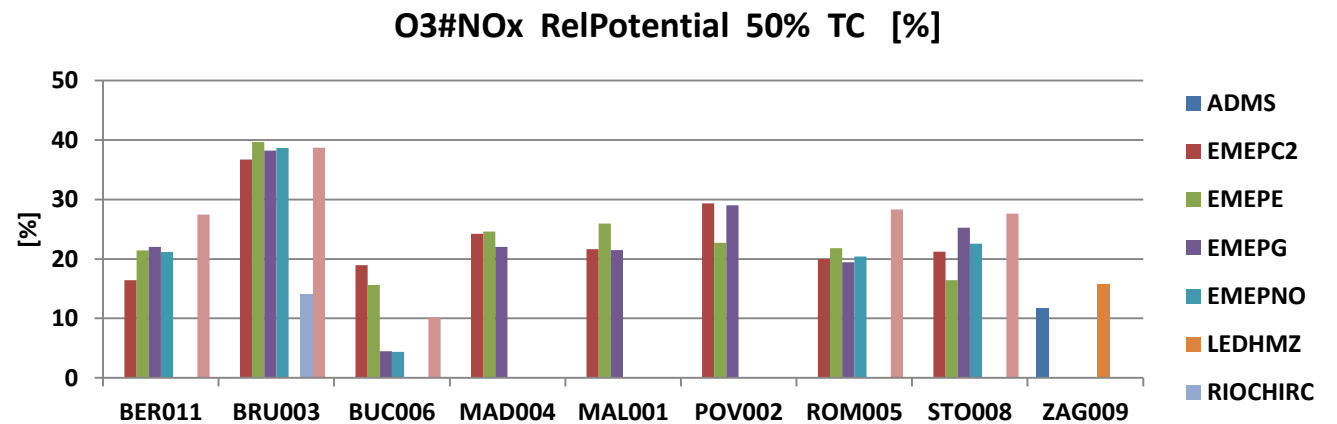
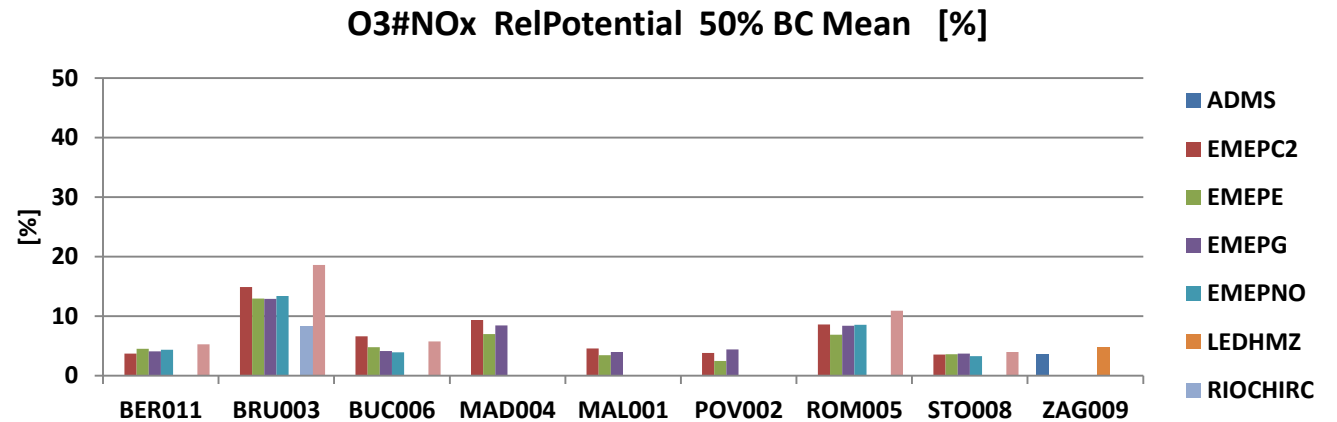


O3#NOx AbsP50/AbsP25 deviation to linearity



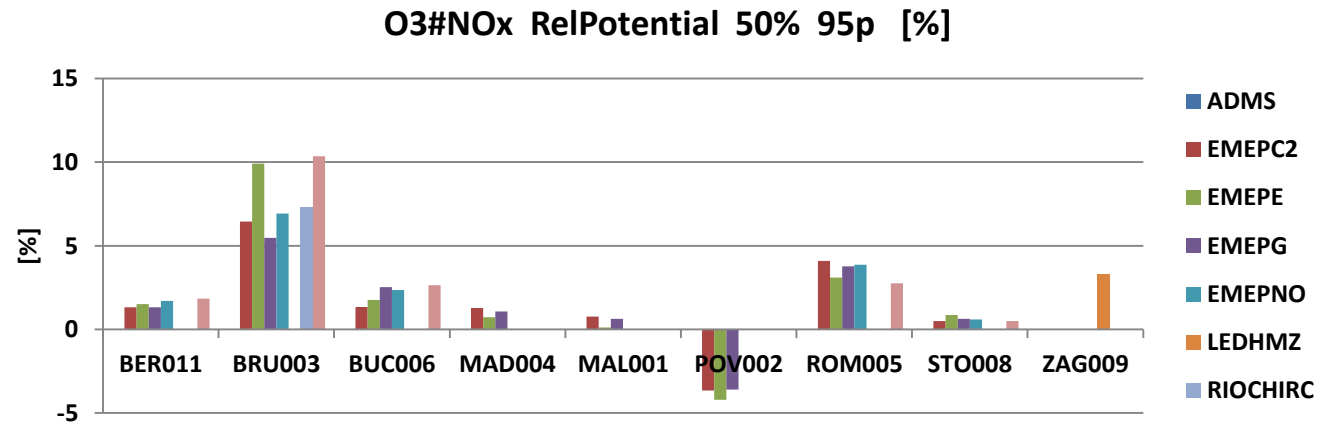
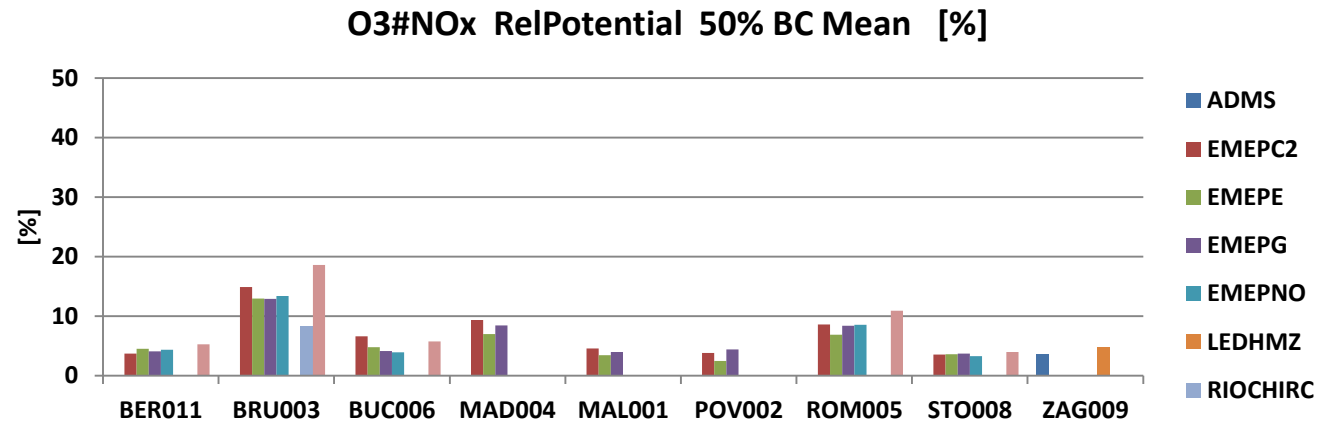
Tcentre versus Mean on O3

- Big jump on relative potential from Mean to Tcentre



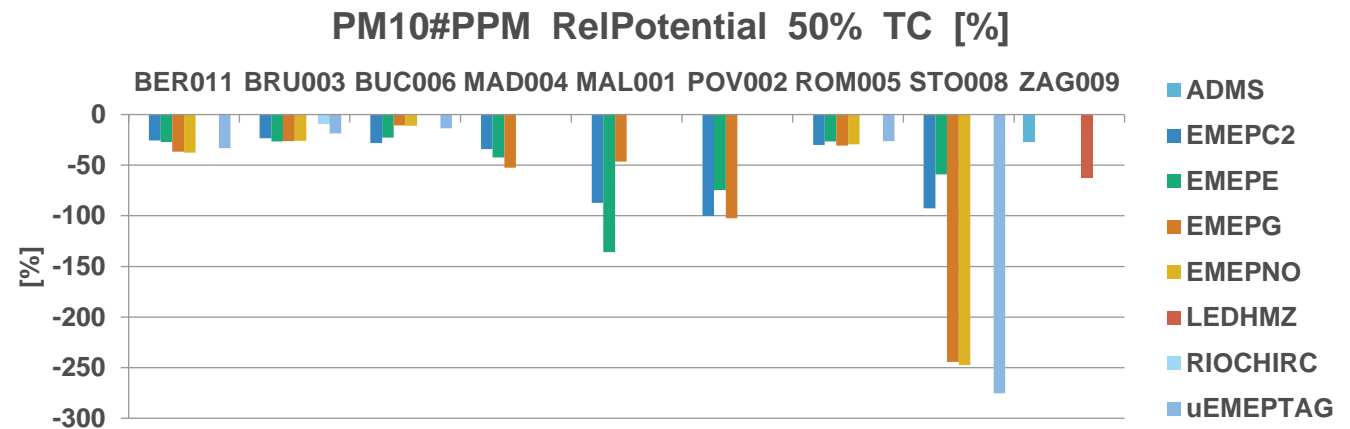
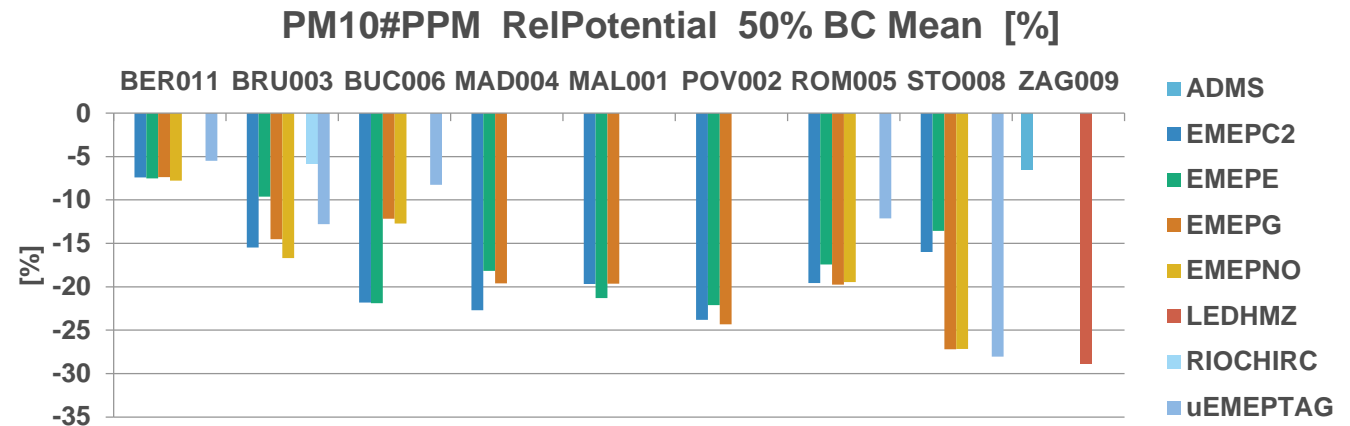
Tcentre versus Mean on O3

- The sign can change for P95 of course for Ozone when including large zones (Po Valley)



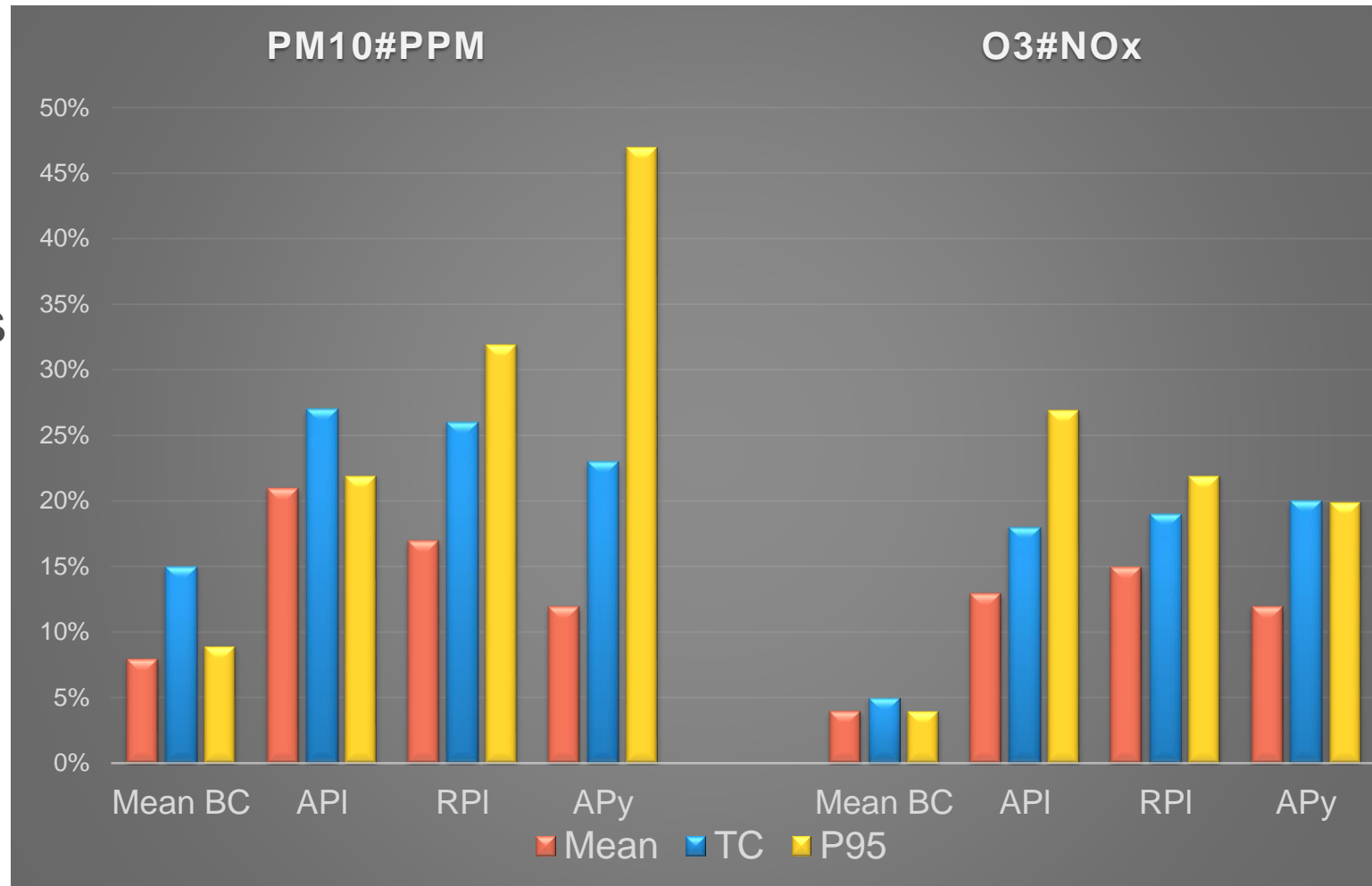
Tcentre versus Mean on PM10

- A clear increase of the relative potential



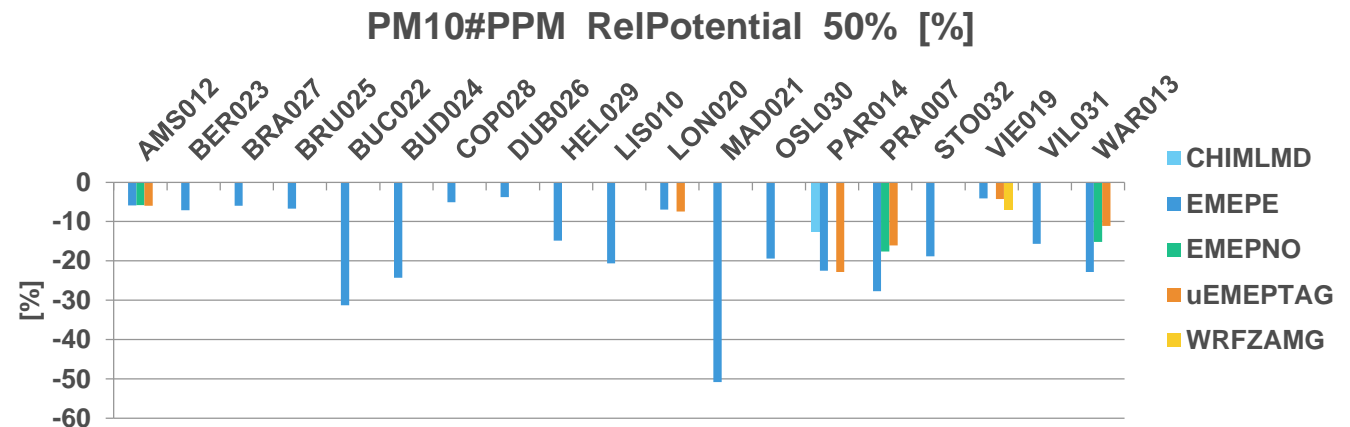
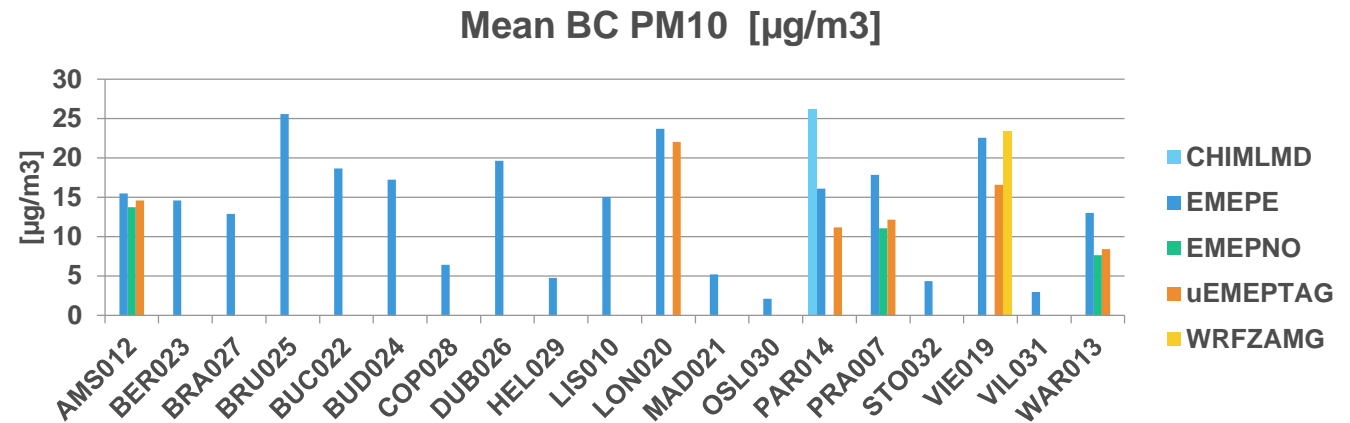
Variability change **Mean** versus **TC** versus **P95**

- Clear increase of variability of indicator scaled by concentrations or emissions



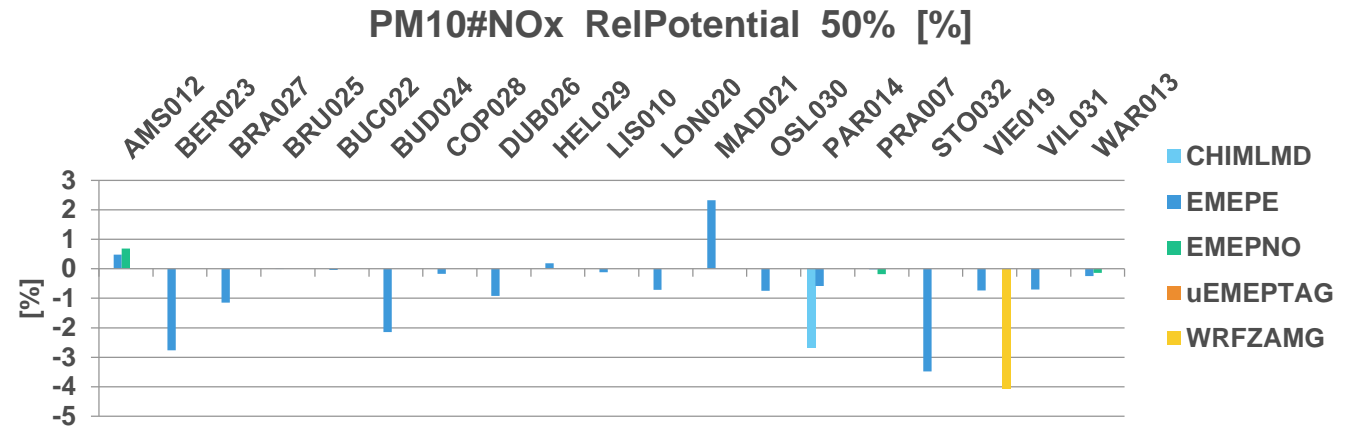
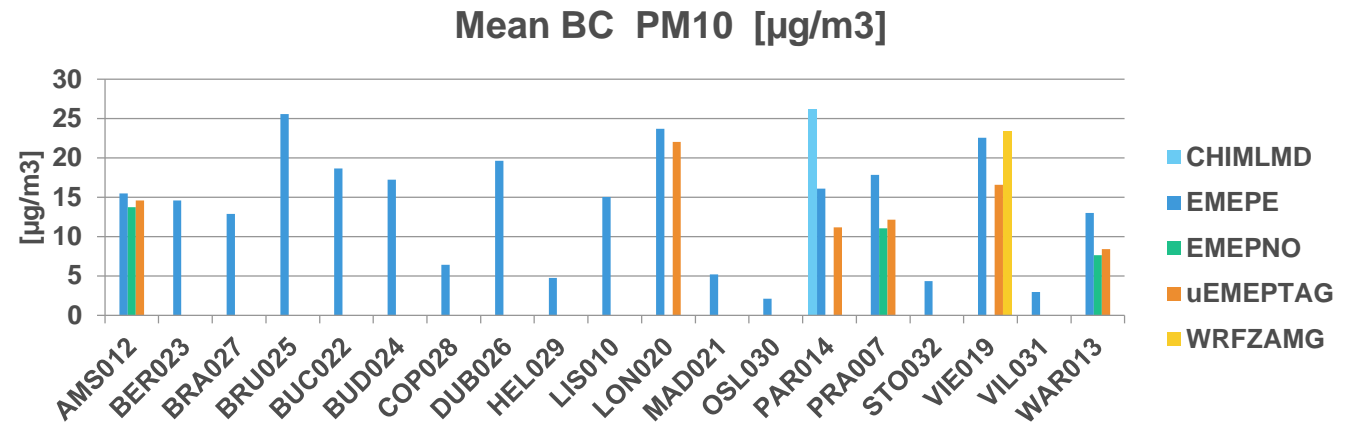
A look at short term

- So far 5 cities with at least 3 different model settings for a PPM emission reductions



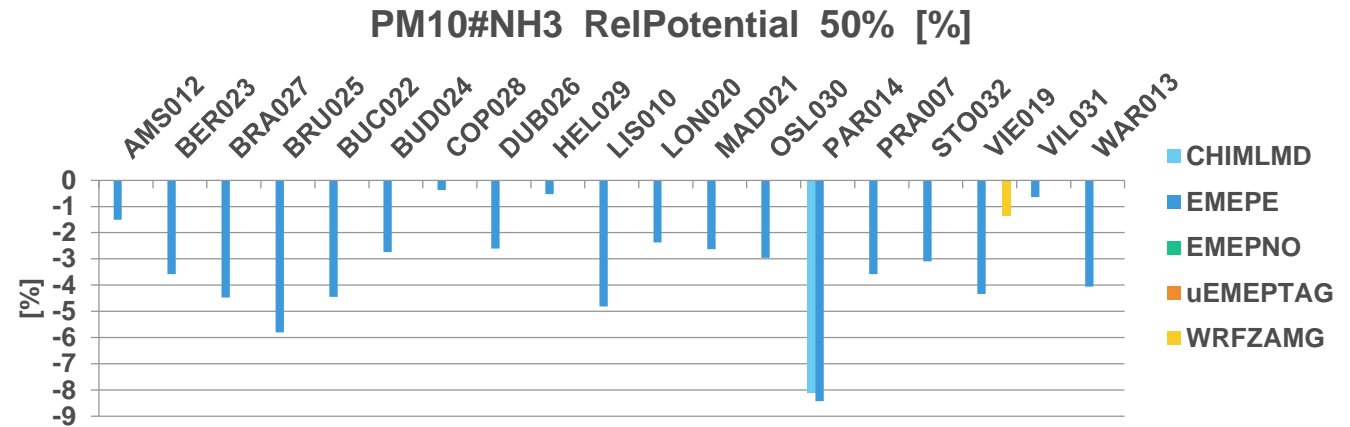
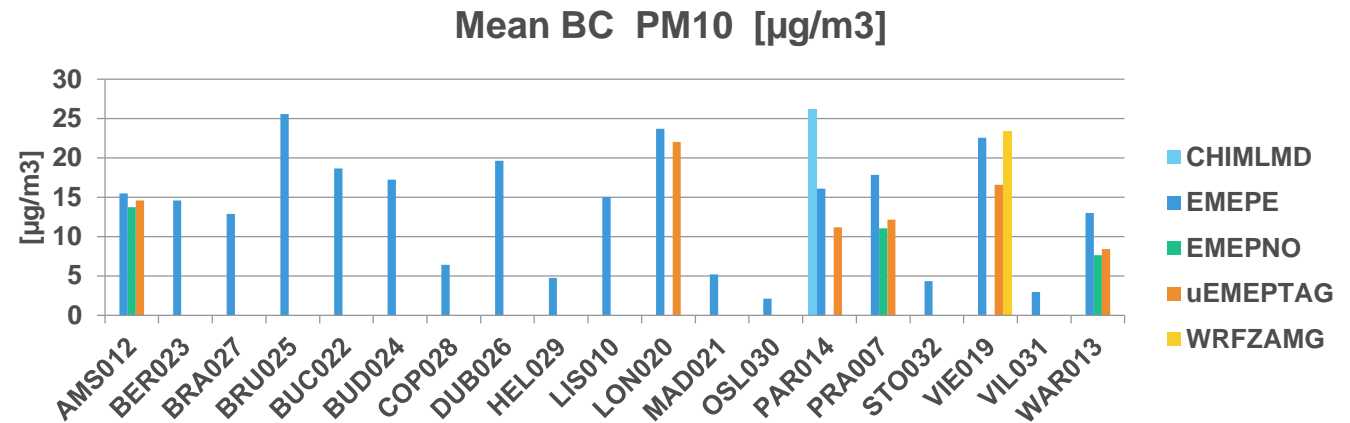
A look at short term

- So far 0 city with at least 3 different model settings for a PPM emission reductions



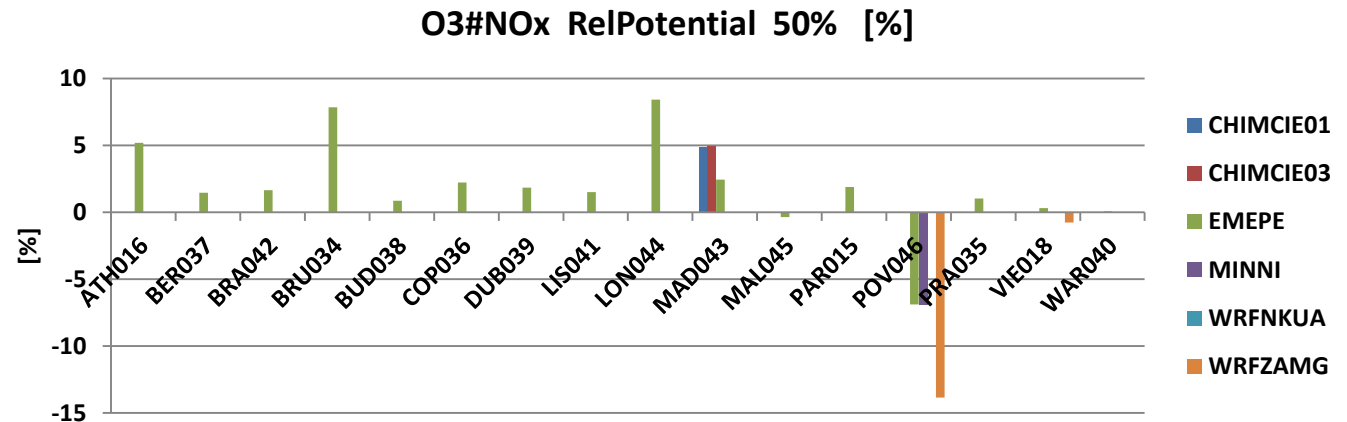
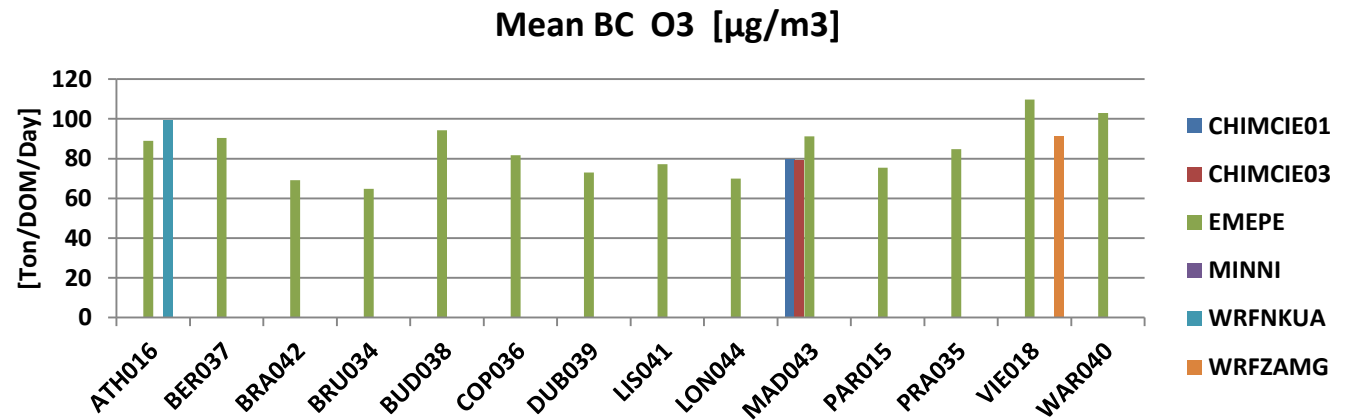
A look at short term

➤ Idem

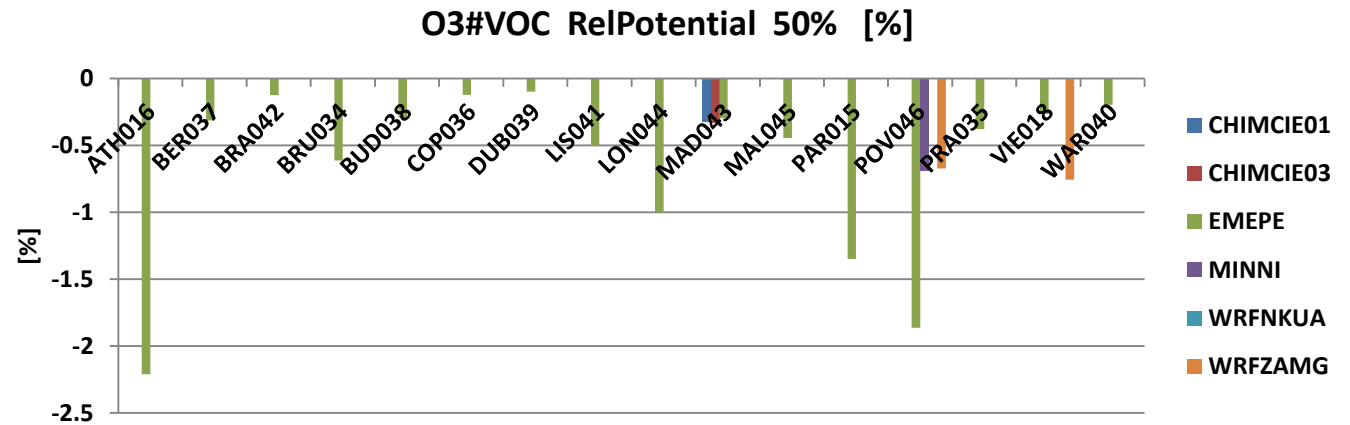
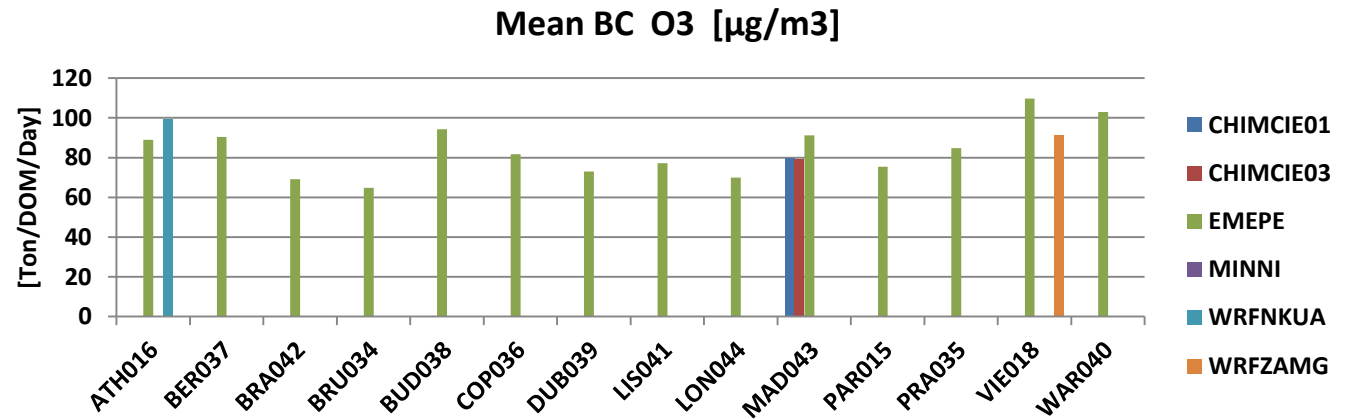


A look at short term for O3 episodes

- Weak impact of resolution in Madrid case with a slight increase of the RPI from 1 to 3km
- Increase everywhere except over the Po Valley



A look at short term for O3 episodes



Linearities over the Po valley for Short Term versus Long Term

- Long Term (LT) versus Short Term (ST) O₃ episodes
- Averaging remove more local non linearities in time and space

Deviation to linearity of API50%/AP25% for O₃#NO_x



Concluding remarks

- A sufficient database to play with but we need more scenarios on Short Term and more emission reductions
- Indicators: API, RPI, APy
- Many simulations remains to be performed particularly for episodes
- Variability based on NSD:
 - Less variability on O3 Mean than PM10 Mean (4 versus 8 % on average for LT)
 - Variability of RPI << API (less clear for O3) - 10 to 25% depending on the indicator
 - Extent of the target domain where emissions are reduced could be important
 - More variability looking at specific location (TC) and shorter timeframe (ST)
- Indicators are sensitive to seasons
- Non linearity effects clearly highlighted for secondary species
- **Open exercise: other modelling teams are still welcome 😊**

Ideas for the next steps (I/II)

- **Continue to populate the database**
 - Yearly simulation for variability assessment and seasonality
 - Short term simulations for variability analysis and prepare next phases
- More data will be necessary to analyse the ST simulations (more hourly concentrations for many species at least at the first level)
- *Other indicators to calculate?*
- *API, RPI, APy, NSD for variability of indicator: Other?*
- *Are taking the mean, TC, P95 adapted?*

Ideas for the next steps (II/II)

If a team performed all requested simulations for this first step then prepare the next phase to understand the delta of concentrations

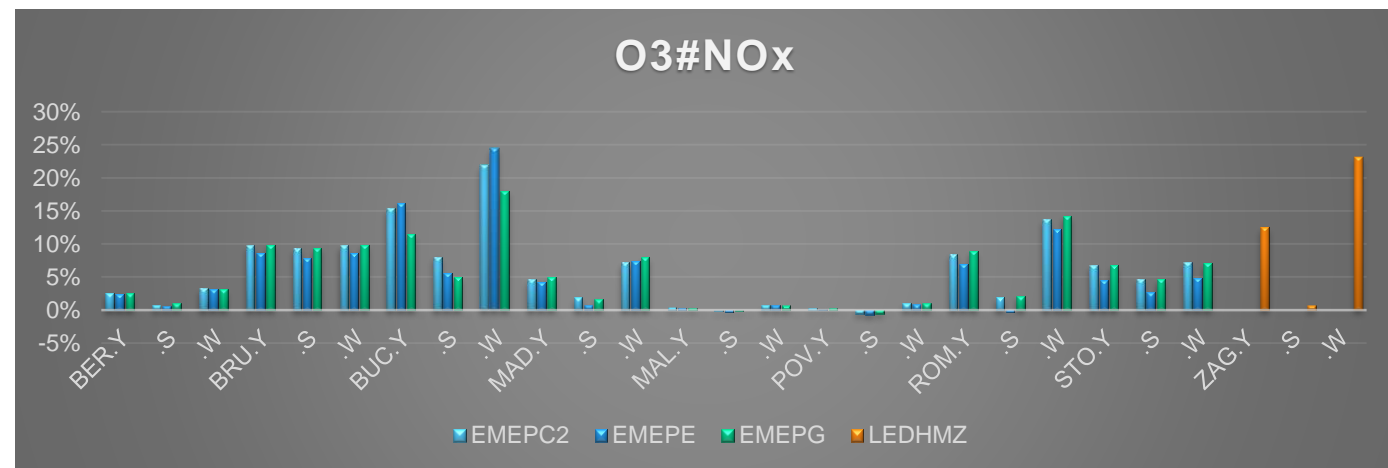
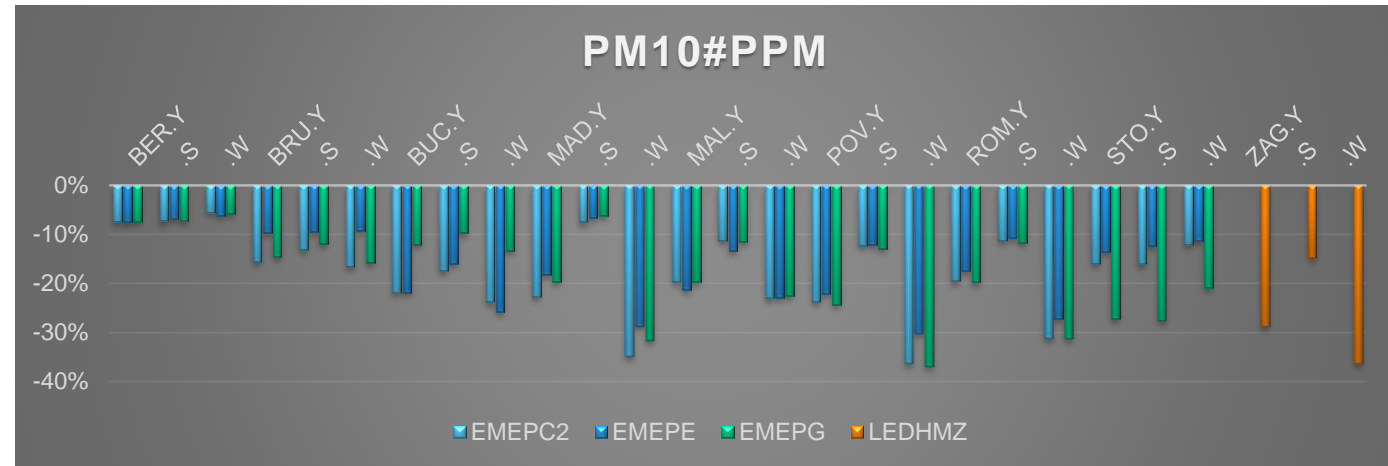
- Creation of sub-groups to understand this large variability with Short Term simulations over episodes
 - One or several models change only one setting (on a target area) that could be:
 - i. Emissions (including the impact of vertical profile distribution, biogenic emissions: NO and VOC)
 - ii. Horizontal resolution
 - iii. Physics and chemistry schemes
 - iv. Lateral and boundary conditions (link with vertical diffusion schemes) particularly for O₃
 - v. General model setting (Domain nesting strategy, numerical schemes)
 - ✓ Other?
 - We need leaders to frame each activity (i, ii, iii, iv, v,...)

The end

Additional slides

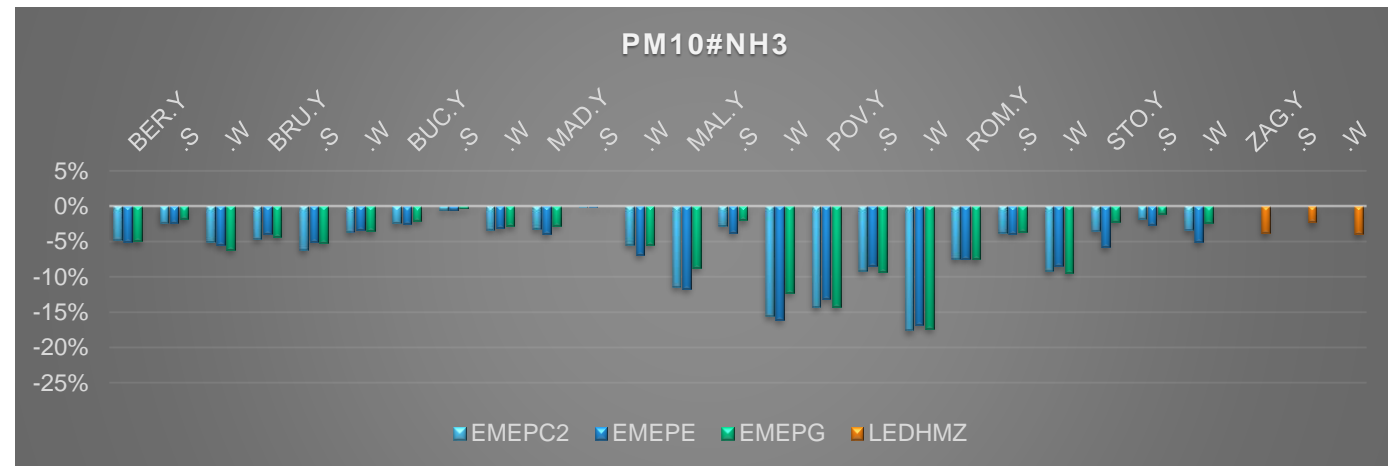
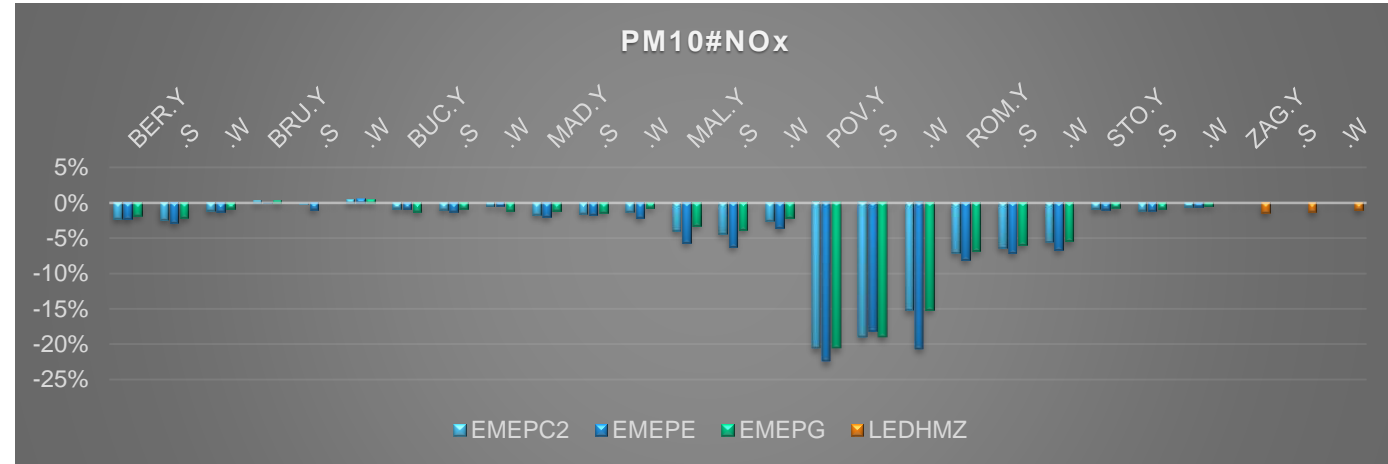
Seasonality of RPI on mean values

- Yearly (Y), Winter (W as DJF), Summer (S as JJA)
- A clear winter(W) vs summer (S)
 - Negative Delta S < W for PM10 with PPM reduction
 - Positive Delta S < W for O3 with NOx reduction

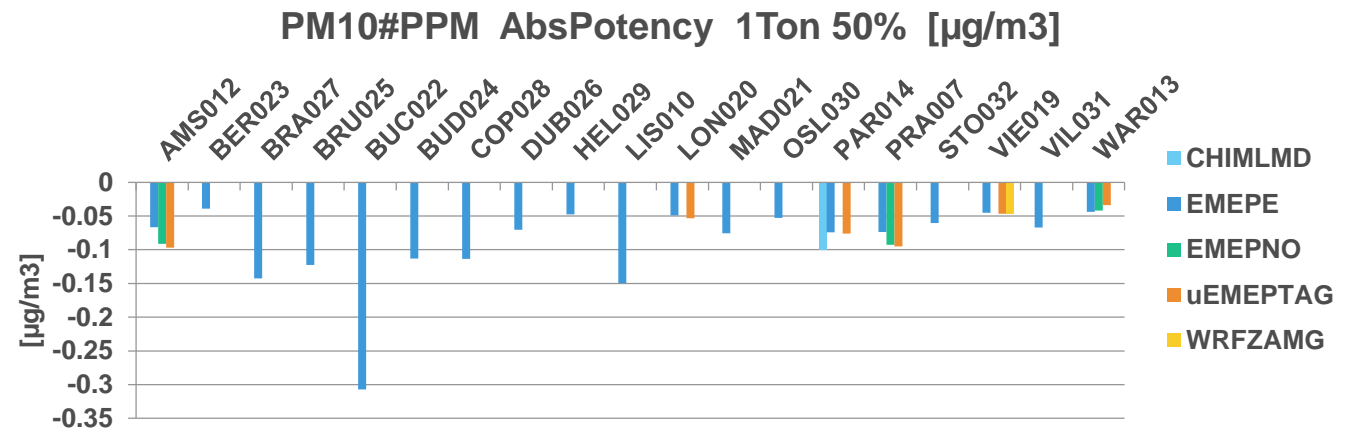
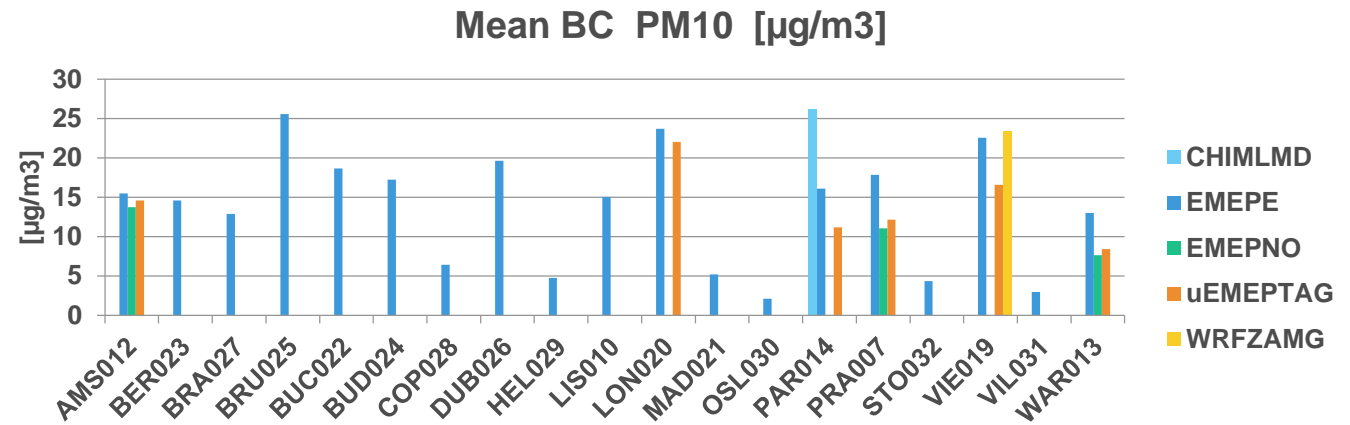


Seasonality of RPI on mean values

- Difficult to have a general behaviour, it depends on cities and chemical regimes

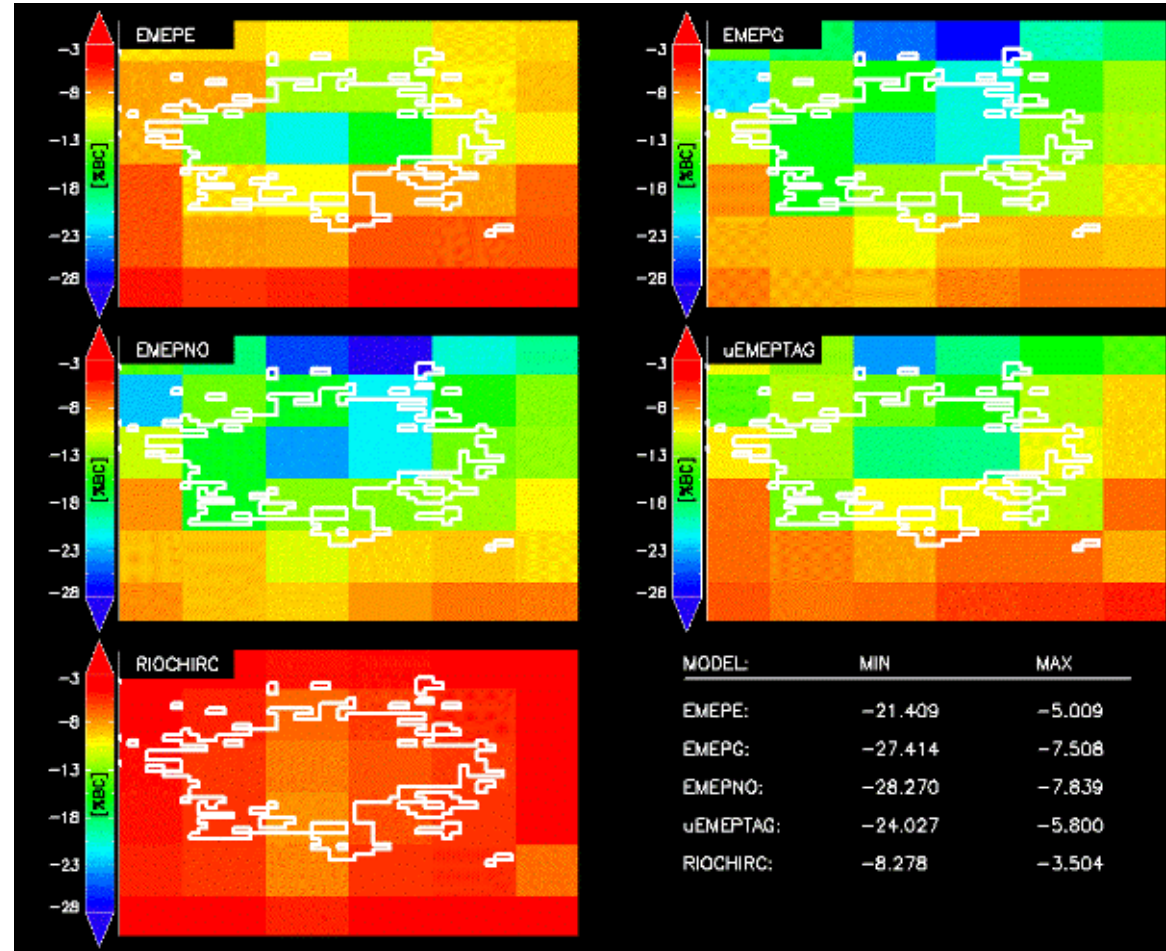


A look at short term



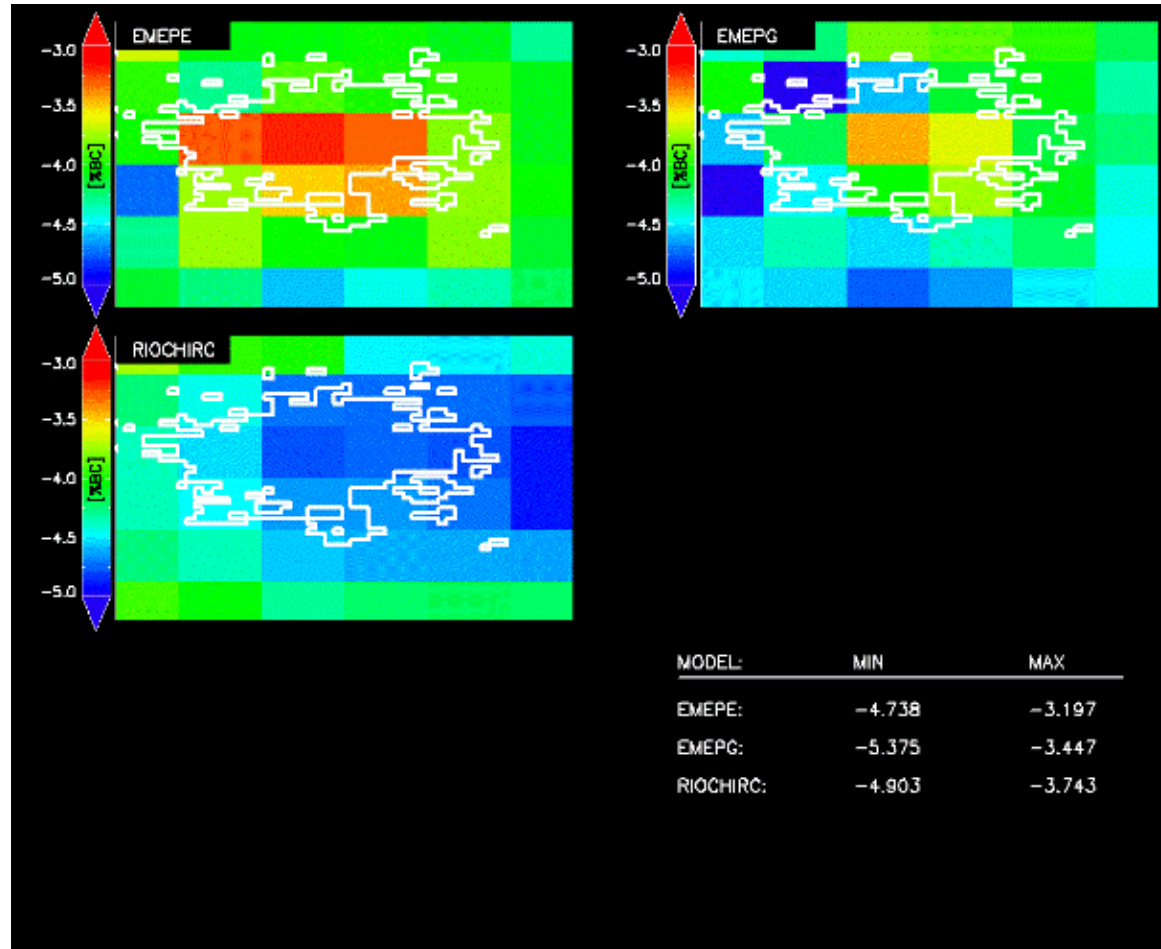
Non linearities in models?

PICT_2D_PM10_BRU003_RPI_50%PPM



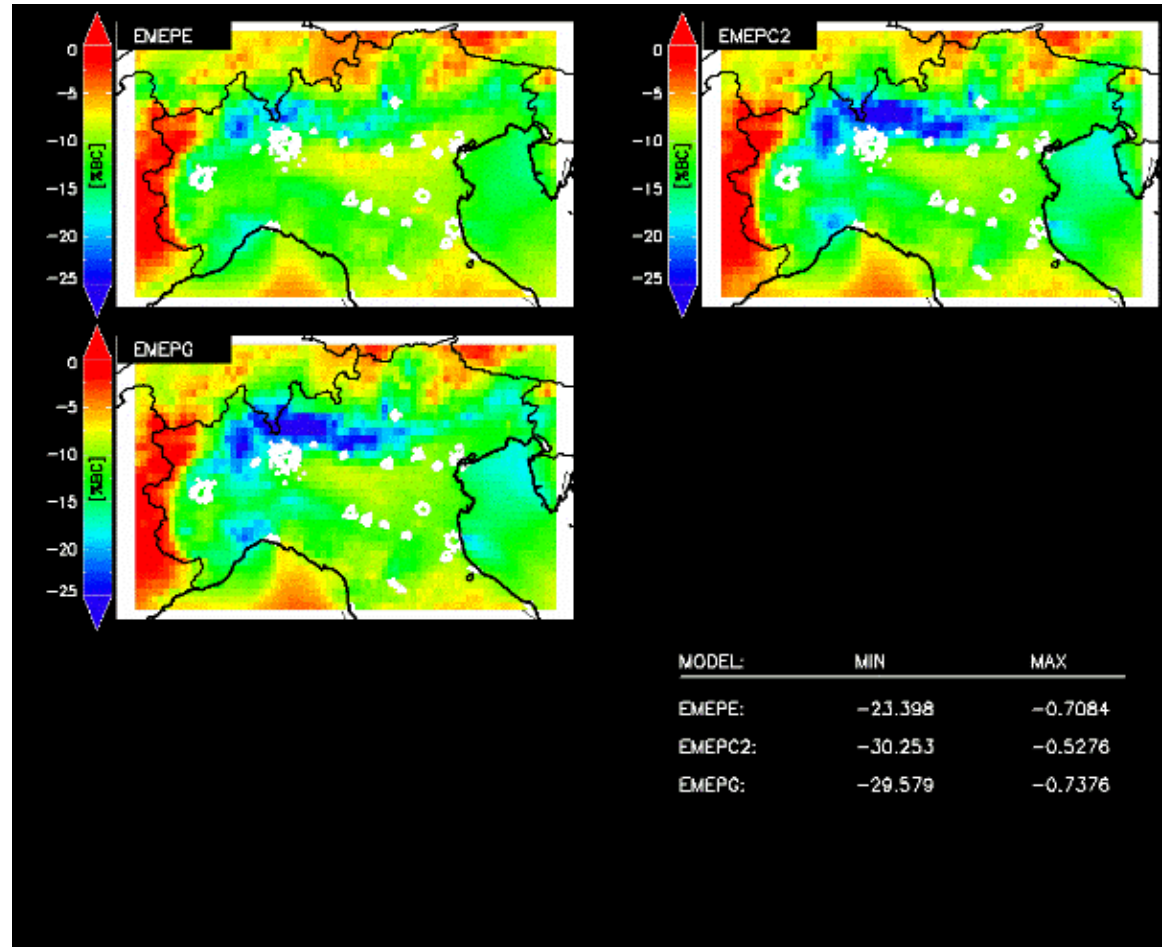
Non linearities in models?

PICT_2D_PM10_BRU003_RPI_50%NH3



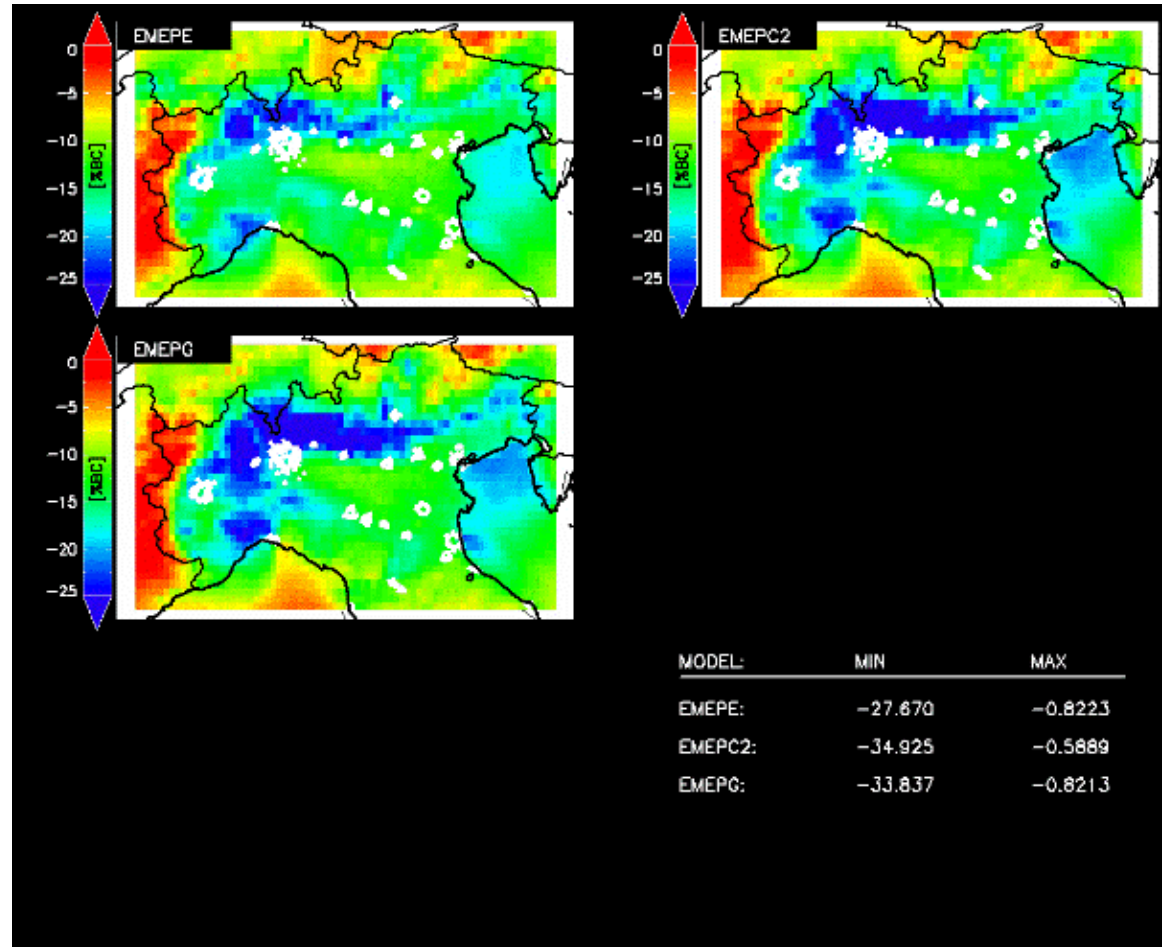
Spatial variability with 2D plots - Linearity

PICT_2D_PM10_POV002_RPI_25%NH3



Spatial variability with 2D plots - Linearity

PICT_2D_PM10_POV002_RPI_50%NH3



Indicators I/II

- k Zone of emission reductions
- n Precursor as NO_x, VOC, PPM, SO_x, NH₃
- m Pollutant: O₃, NO₂, PM₁₀,...
- s Simulation scenario named as 50%NO_x, etc. ... or base case BC
- α Emission reduction (25 or 50%)
- $E^{m,k,s}$ Total emission of precursor m over zone k for simulation s
- $e_{i,j}^{m,k,s}$ Emission of precursor m over zone k for grid cell $(i,j) \in A_k$ for simulation s
- $C^{n,k,s}$ Averaged concentration of pollutant n over zone k for simulation s
- $c_{i,j}^{m,k,s}$ Concentration of pollutant n over zone k for grid cell $(i,j) \in A_k$ for simulation s

Seasonality of RPI on mean values

- Ratio of RPI between NOx and NH3 reductions
- Usually same sign of concentration reductions except in Brussels
- Usually higher RPI from ammonia emission reduction scenario except in summertime

