

The FAIRMODE WG9 platform

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FAIRMODE WG9 OBJECTIVES

- For a given mitigation scenario (scen) and a base case (BC), models (M) provide different absolute results C^M_{scen}
- BUT, HOW MODELS BEHAVE ON DELTAS?

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

Policy Implication:

It is important to assess the robustness of deltas for urban air quality policies!

- 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?



Models and teams involved - Overview

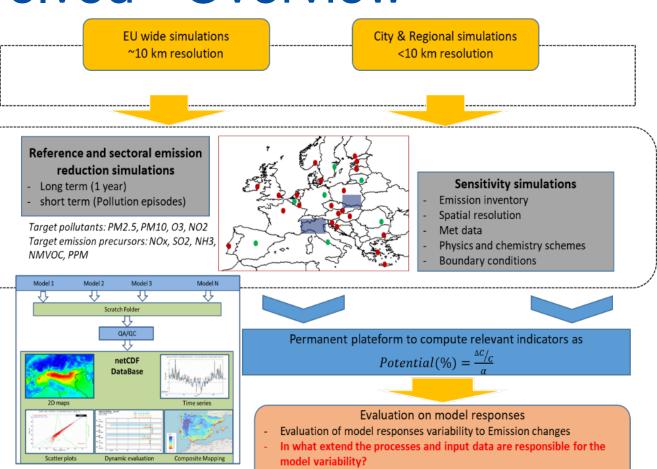
Constraints:

-Meteorology 2015

-Emission reductions 25 and 50%

-Target domains, periods (episodes)

Team name	- Country	Model Name
JRC	(EU)	EMEP
ZAMG	(AT)	WRF-Chem
Met Norway	(NO)	EMEP
Met Norway	(NO)	EMEP + uEMEP
Cyl	(CY)	WRF-Chem
NKUA	(GR)	WRF-Chem
DHMZ	(HR)	ADMS-Urban
DHMZ	(HR)	LOTOS-EUROS
LMD/IPSL	(FR)	WRF-CHIMEREv2020r1
UH-CACP	(UK)	WRF-CMAQ
CIEMAT	(ES)	IFS-CHIMEREv2017r4
ENEA	(IT)	WRF-MINNI
IRCELINE	(BE)	CHIMERE + RIO + ATMOSTREET



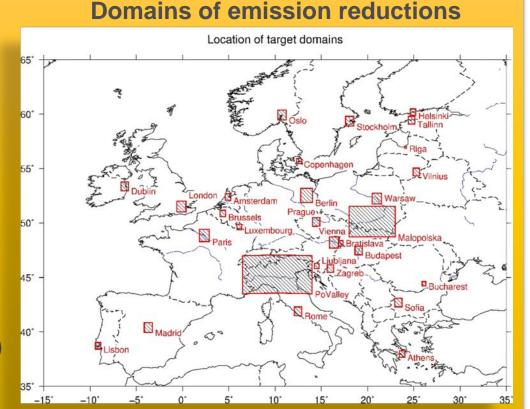
Two papers under review



The overall framework

Set-up

- Short term (ST) on episodes
 - Emissions reduced only during 2015 episodes from 00:00 to 23:00
- Long term (LT) simulations
 - Emissions reduced for the whole year 2015
- 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 (All together)
 - Ozone: NOx, VOC, ALL (All together)





The overall framework

Basis Indicators

> Absolute Potential defined as the reduction in μ g/m³ scaled by the reduction α of the scenario (25 or 50%) of a precursor from base case BC

• $APl = (C_{SCEN} - C_{BC})/(\alpha)$ ($APl \times \alpha$ is the delta of concentrations)

- > Relative Potential 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.
 - $RPl = (C_{SCEN} C_{BC})/(\alpha \times C_{BC})$
- Absolute Potency in µg/m³/(ton/day) 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)

•
$$APy = (C_{SCEN} - C_{BC})/(\alpha \times E_{BC})$$



Main conclusions of our paper (I)

- The evaluation of model outputs show a large variability of model performances for the base case, a high spatial resolution does not automatically improved the model performances even for pollutants influenced by local emissions like NO₂;
- The variability of model responses using delta-based indicators is higher for PM₁₀ than for Ozone;
- The variability of model responses is higher than the variability of base case concentrations and emissions;

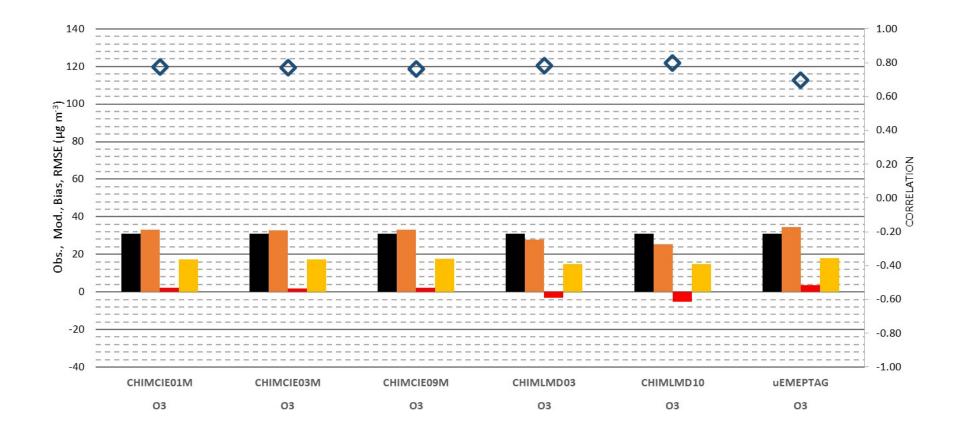


Main conclusions (II)

- Relative indicators like the Relative Potential (normalized by the concentration) and the Absolute Potency (normalized by the emission reductions) have a lower variability compared with the Absolute Potential (which is proportional to the delta of concentrations);
- For **ozone**, the analysis of linearity and additivity of model responses show a clear impact of **non-linear chemistry processes**, which produce a large deviation to linearity and additivity of emission reductions;
- For PM, the response is in general more linear and additive particularly, as expected, when reducing the primary emissions of particles which weakly perturb the chemical and physical processes involved in the PM formation;
- One should remain cautious in the interpretation of these **indicators**, because they are built on averages and ratio of values that can be **very low** with **different signs**.



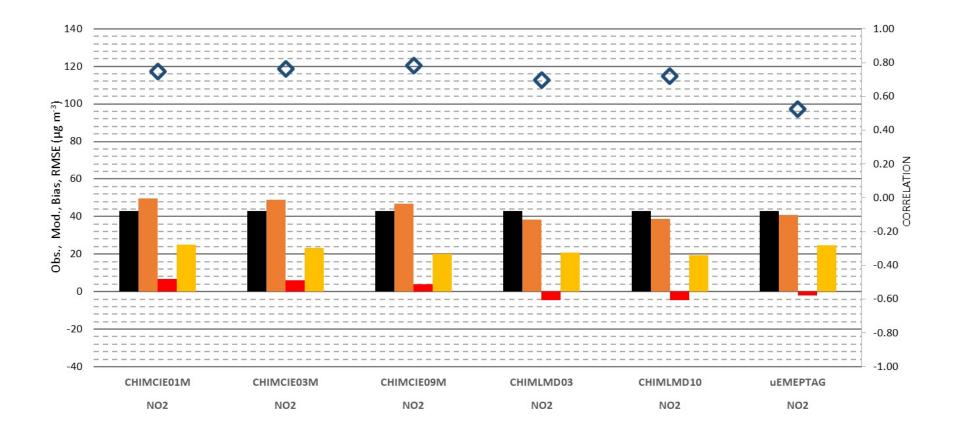
Evaluation over Paris (PM episode) – O3





■ Obs. ■ Mod. ■ Bias ■ RMSE ◆ Correlation

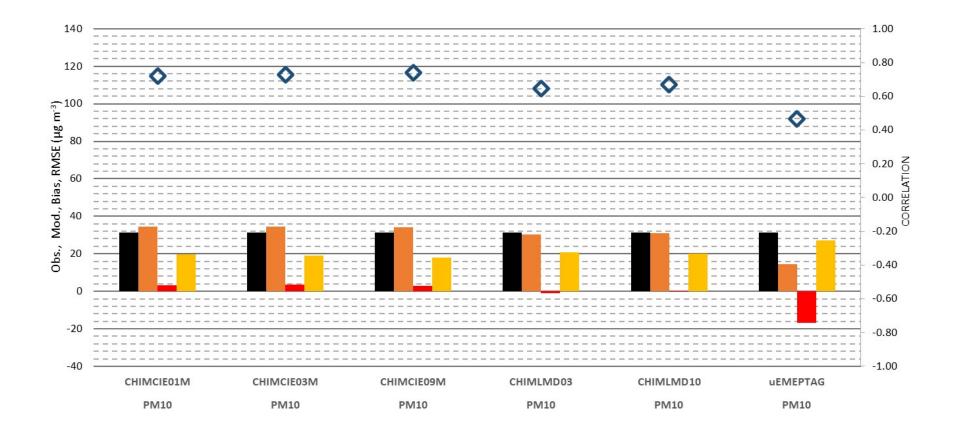
Evaluation over Paris (PM episode) – NO2





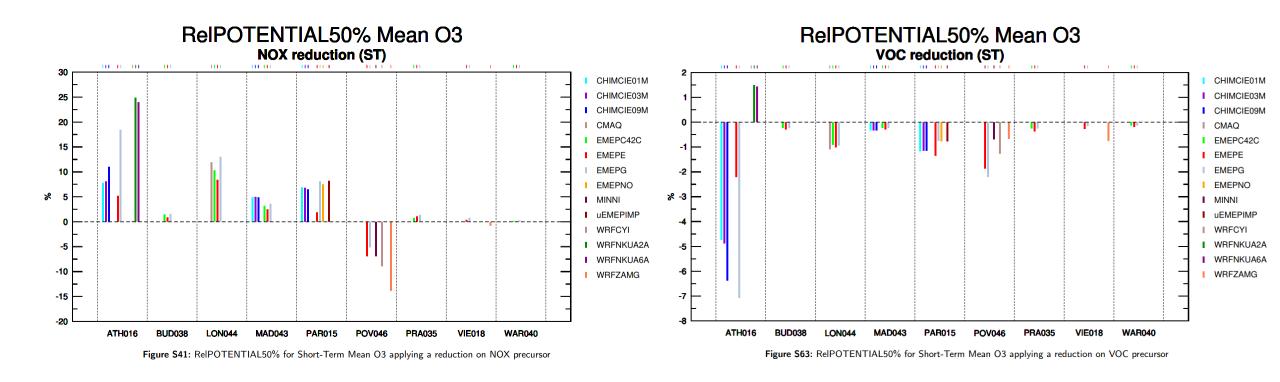
■ Obs. ■ Mod. ■ Bias ■ RMSE ◆ Correlation

Evaluation over Paris (PM episode) – PM10



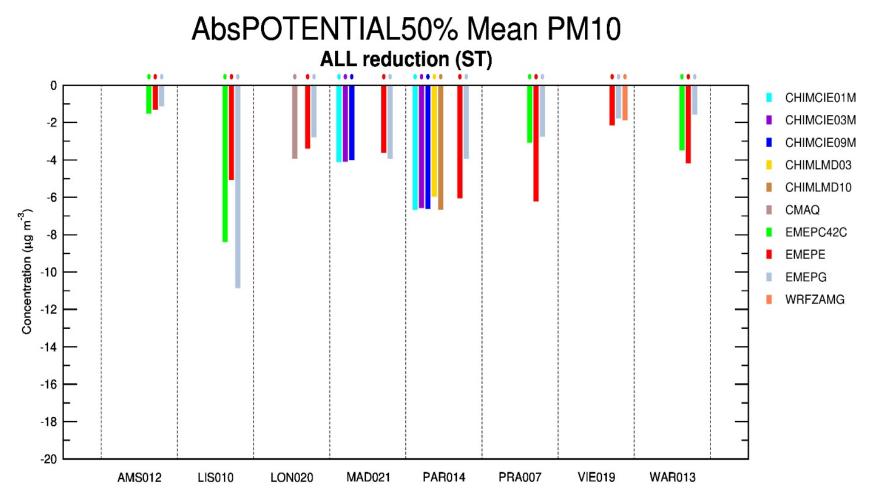


Relative Potential for O3 for NOx and VOC reduction

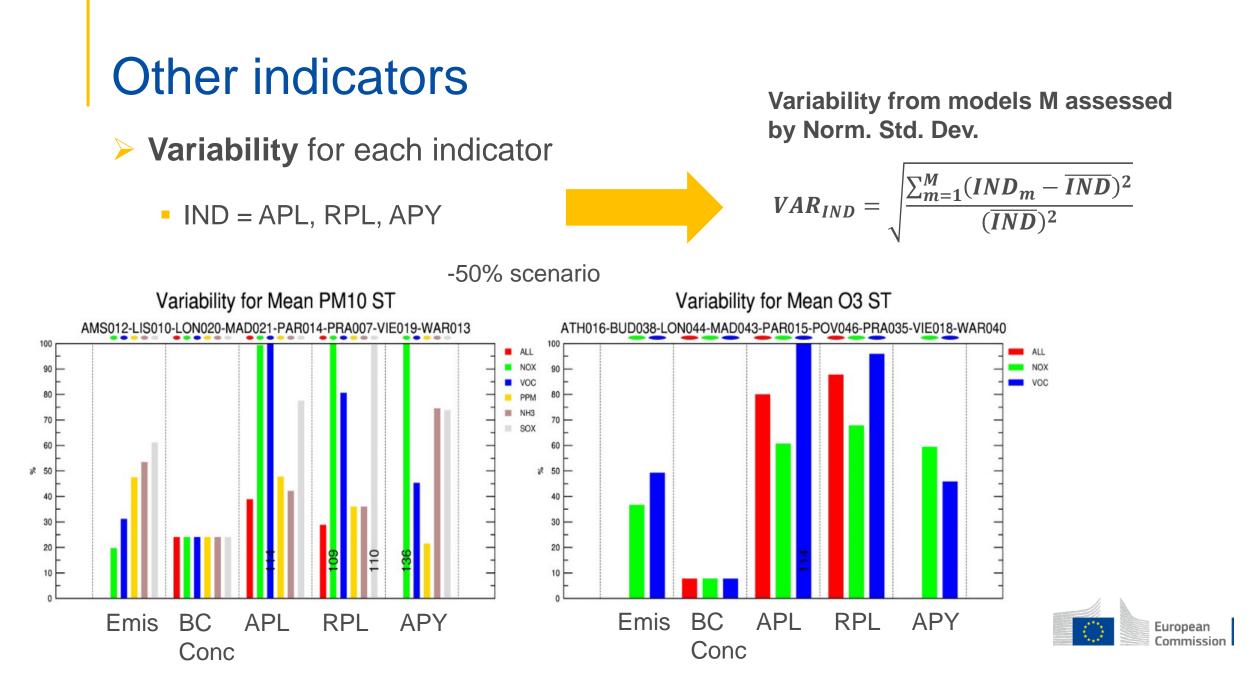




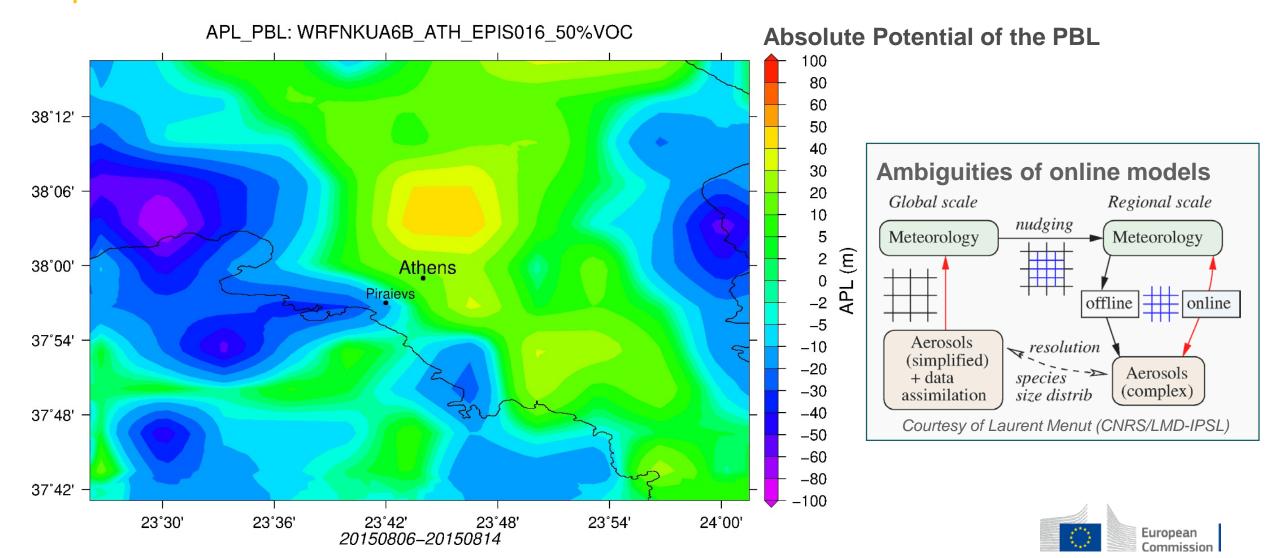
Absolute Potential for PM10 with ALL pollutant reductions





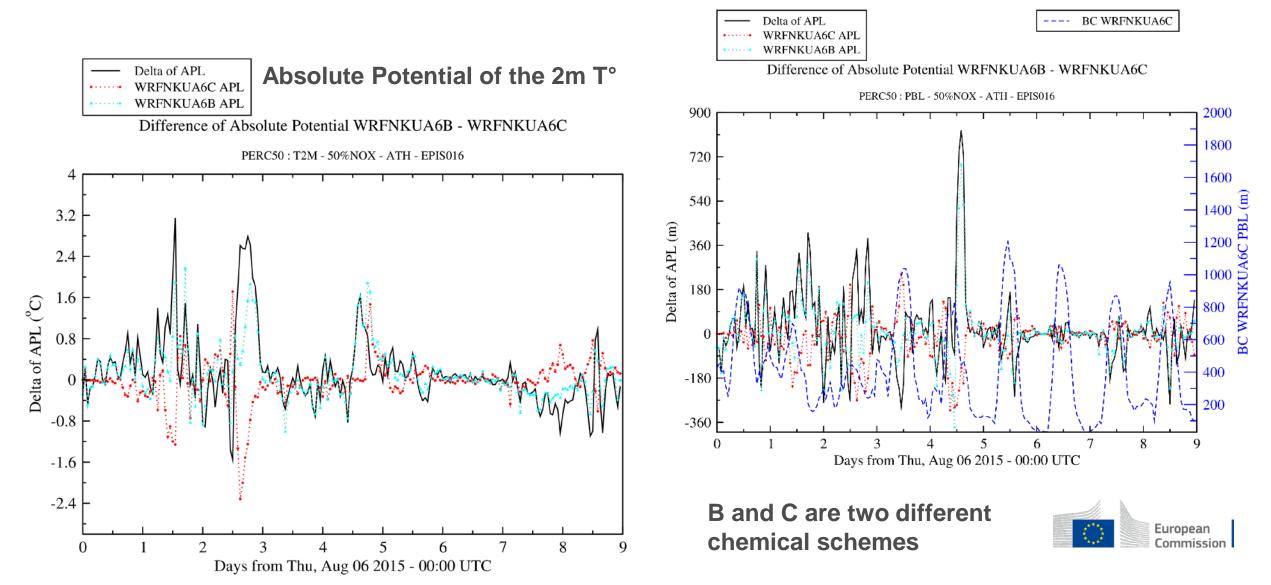


WRF online radiative feedbacks

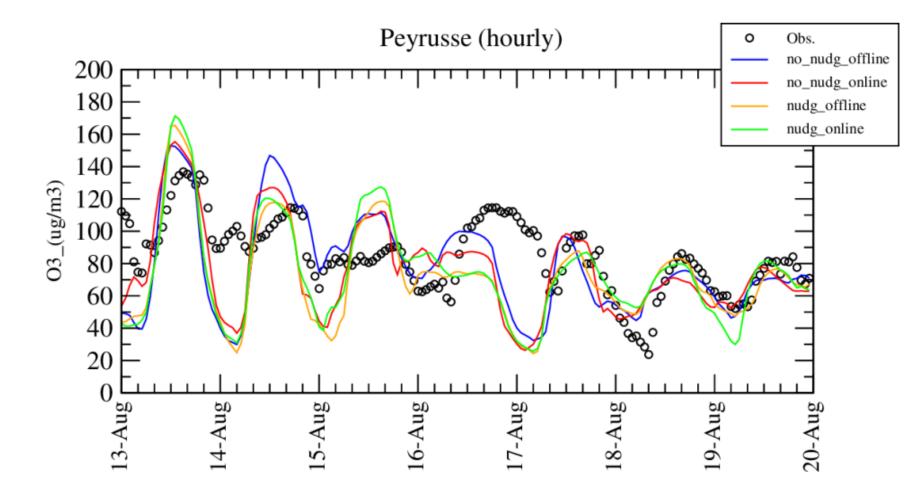


WRF online radiative feedbacks

Absolute Potential of the PBL



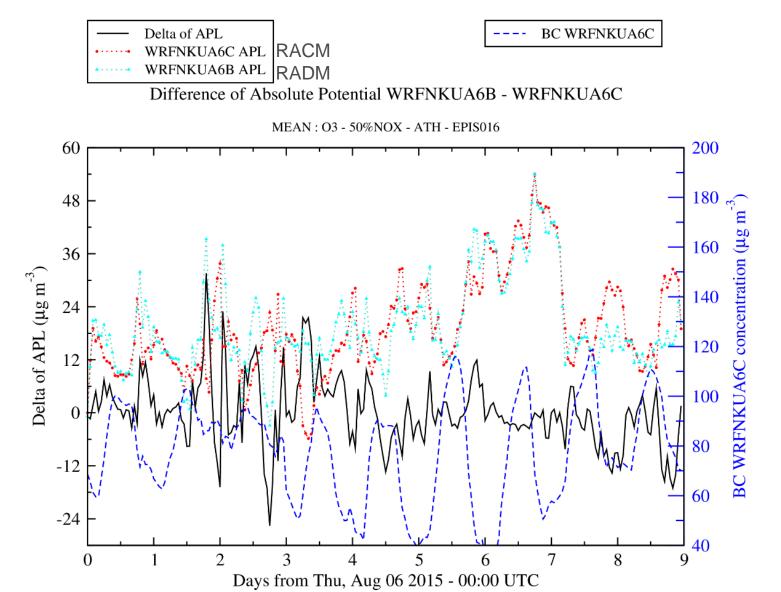
Online coupling with or without nudging in CHIMERE-WRF on a base case



Courtesy of Laurent Menut (CNRS/LMD-IPSL)

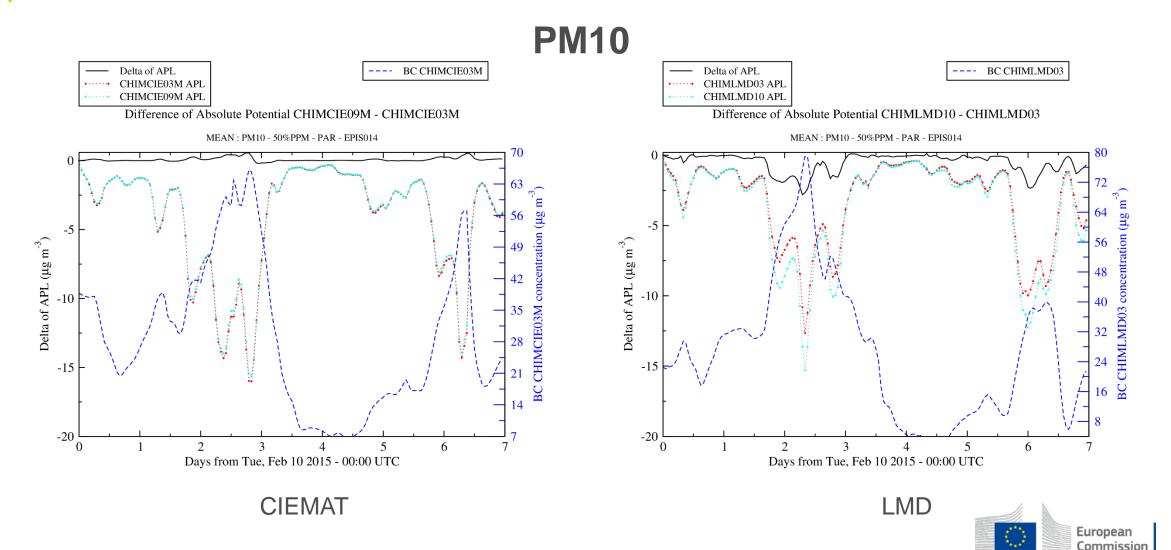


Impact of the chemistry (RACM vs RADM) over Athens

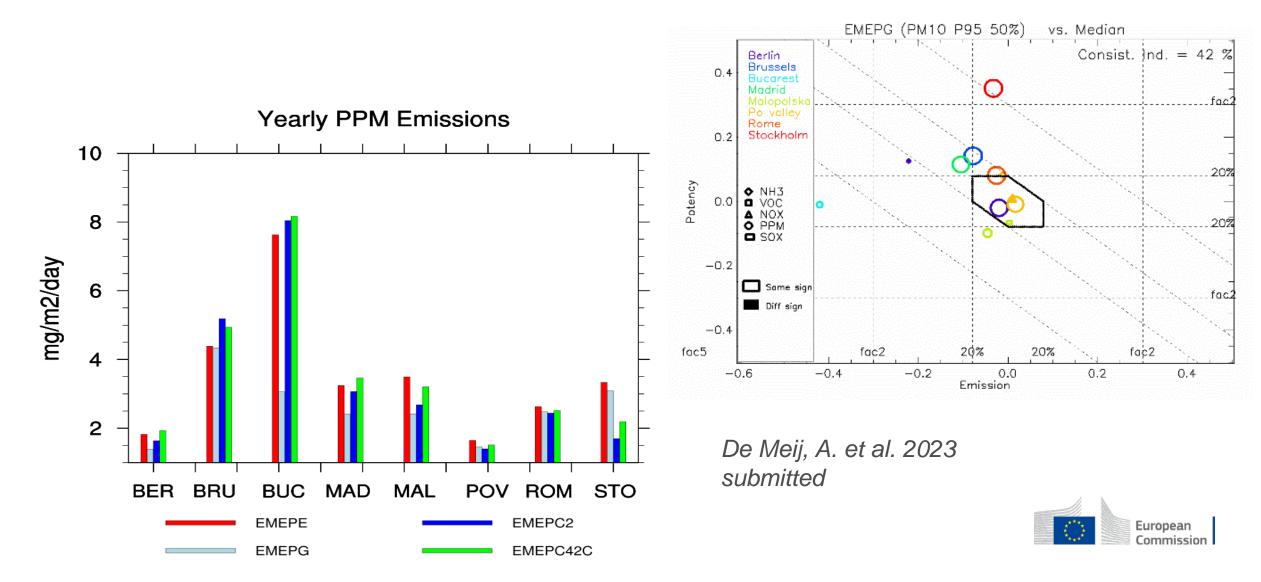


European Commission

Impact of the resolution: CHIMERE over Paris (PM episode) – From 3 to 10km



Impact of emissions on the Potency



Conclusions

> A good plateform to experiment the impact of emission reductions on delta

- Study on the relevance of indicators
- Development of a visualization plateform
- A first flavor of variability is provided with some conclusions (impact of processes, emissions, resolutions)
- An opportunity for modellers to challenge their set-up (bug detection thanks to dynamic evaluation)

➤ Two papers under review



Next steps to be discussed

> Are the teams still interested to go on feeding the database?

> Add more constraints (for instance same emissions)?

Cut emissions by sectors and pollutants to be more realistic?

Simulation of episodes (less costly)?

Make use of observations?

A modelling team could change one setting flags at a time to generate a new model member (vertical distribution of emission, vertical mixing scheme, vertical resolution)

> Make use of SHERPA as a reference for LT?

> Development of an online version of the visualization tool (> Kees)





Thank you for your attention





- > Variability for each indicator
 - IND = API, RPI, APY

Variability from models M assessed by Norm. Std. Dev.

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

Fest of linearity using the 50% and 25% runs. Deviation to linearity for API

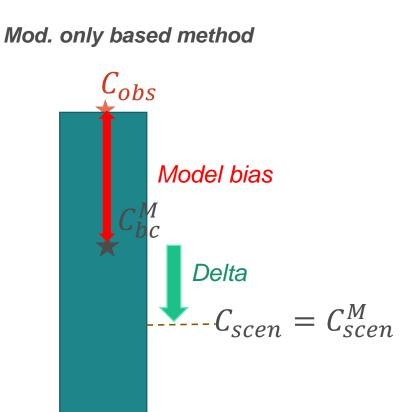


European

> Test of additivity using the ALL scenarios and "ADD" as the sum of individual precursors reductions. **Deviation to** $100 \times \left(\frac{IND_{ADD} - IND_{ALL}}{IND_{ALL}}\right)$ additivity for API, RPI

FAIRMODE CT9 CONTEXT TOPIC 2

- Many inter-comparison exercises of air quality models
- No recent exercises to assess the capacity of models to simulate "delta" (Formerly CityDelta, EURODELTA) particularly at more local scale
- Need to have a long term inter-comparison <u>platform</u> to continually assess model responses





FAIRMODE CT9 CONTEXT TOPIC 2

delta

- Many inter-comparison exercises of air quality models
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- Need to have a long term inter-comparison <u>platform</u> to continually assess model responses

- A Model Concentration Delta can be applied to an observation C_{obs} to evaluate a scenarios based on 'bc' reference and 'scen' simulations:
 - Absolute (for O3?): $C_{scen} = C_{obs} + \overline{(C_{scen}^M C_{bc}^M)}$
 - 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

