



# EVALUATING THE ROBUSTNESS OF THE SHERPA AIR QUALITY MODEL THROUGH THE APPLICATION OF GLOBAL SENSITIVITY ANALYSIS TECHNIQUES

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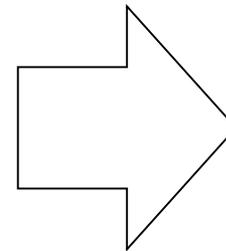
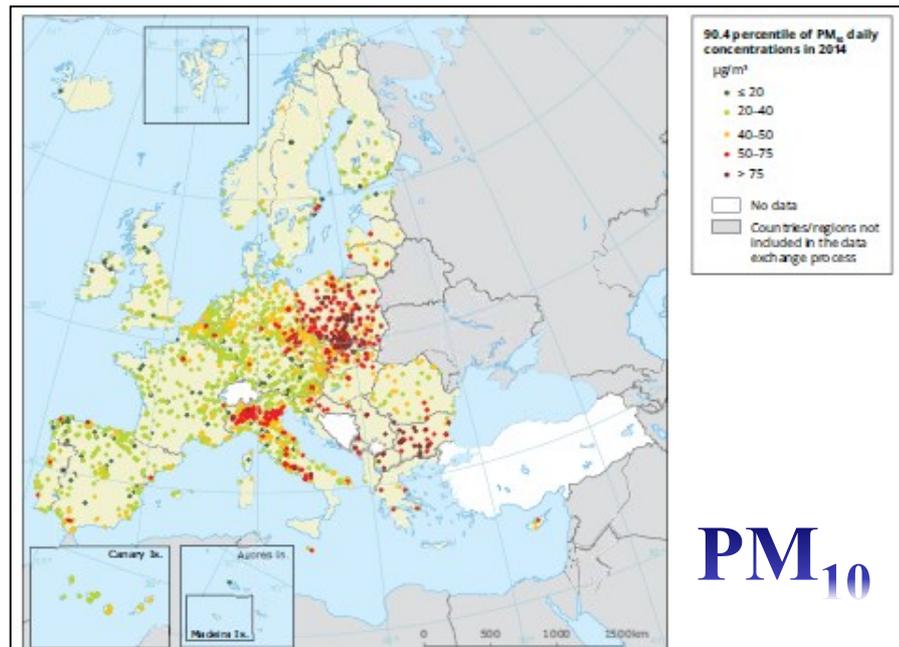
# Outline

1. Introduction of challenges and tools
2. How to apply the sensitivity analysis to an integrated assessment tool, SHERPA case
3. Results
4. Conclusions

# Background



Europe's air quality is slowly improving, but fine particulate matter nitrogen oxides and ground-level ozone in particular continue to cause serious impacts on health, especially in urban environments



**Assessed against EU Limit Value**

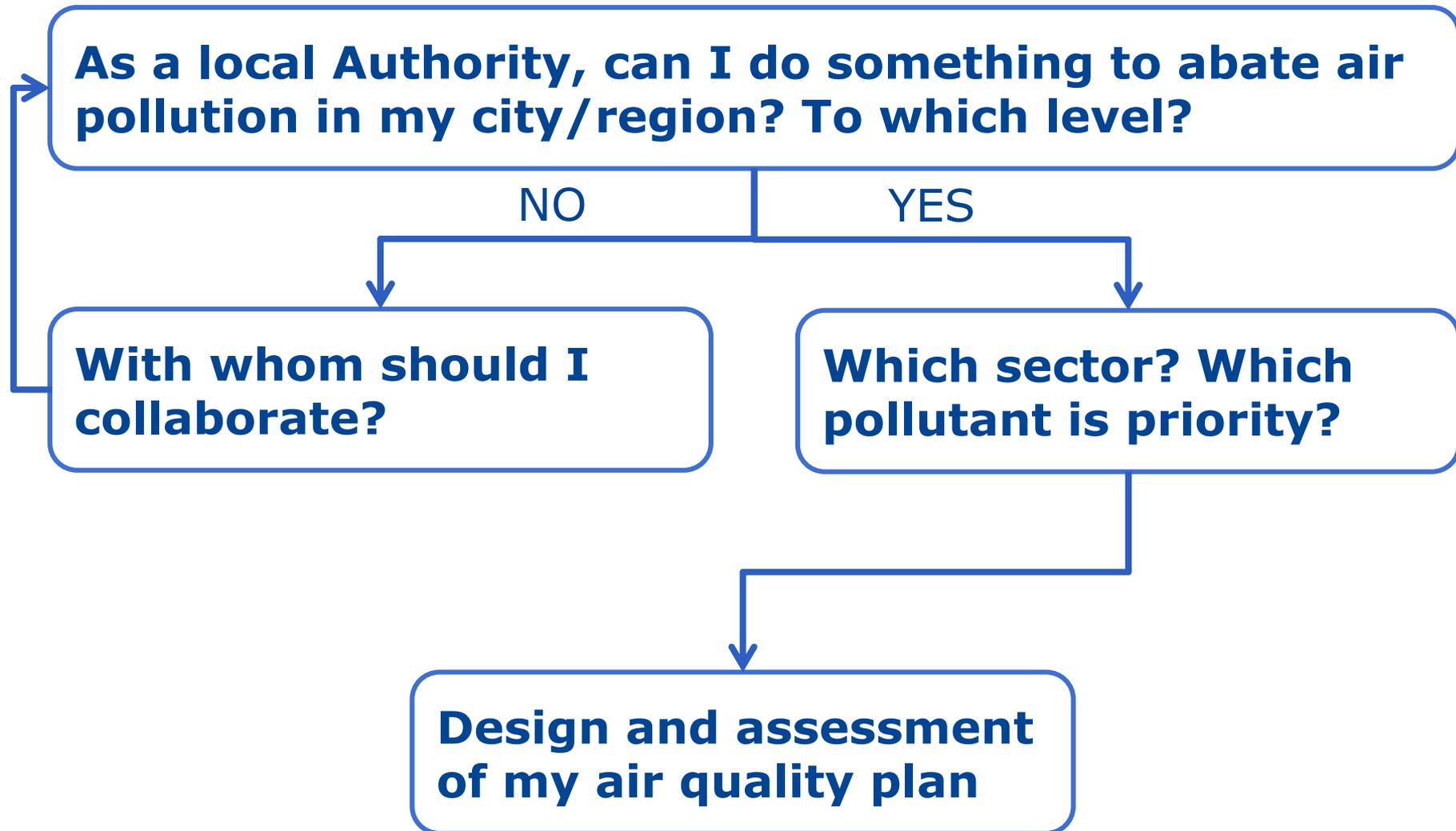


**Assessed against WHO Guidelines**



- 400.000 premature deaths in the EU28 (PM<sub>2.5</sub>)
- Most country (cities) fail to fulfill the EU limit values
- Compliance is an Air Quality Directive obligation → air quality plans

# A possible scheme for air quality plans...



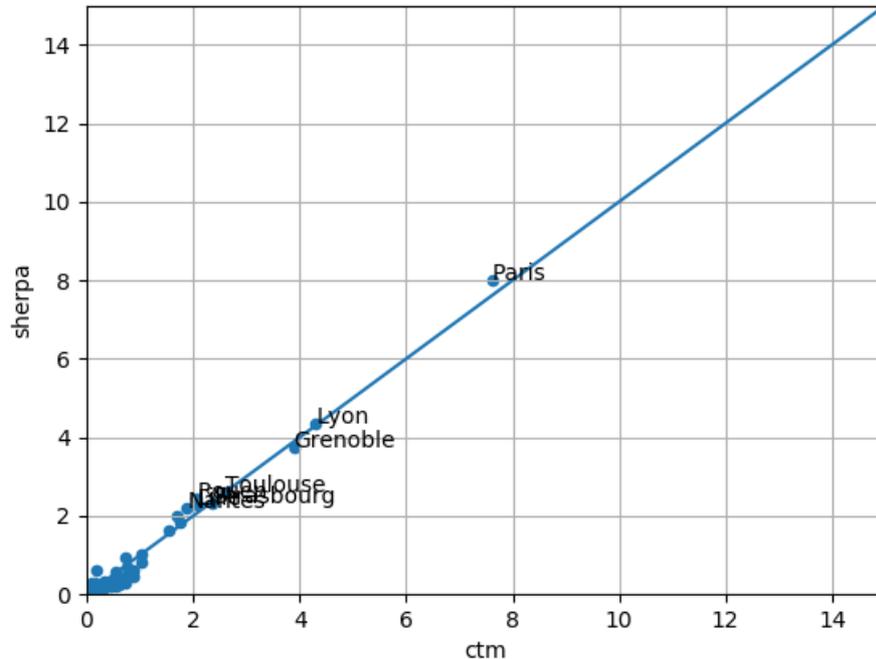
# The SHERPA screening tool



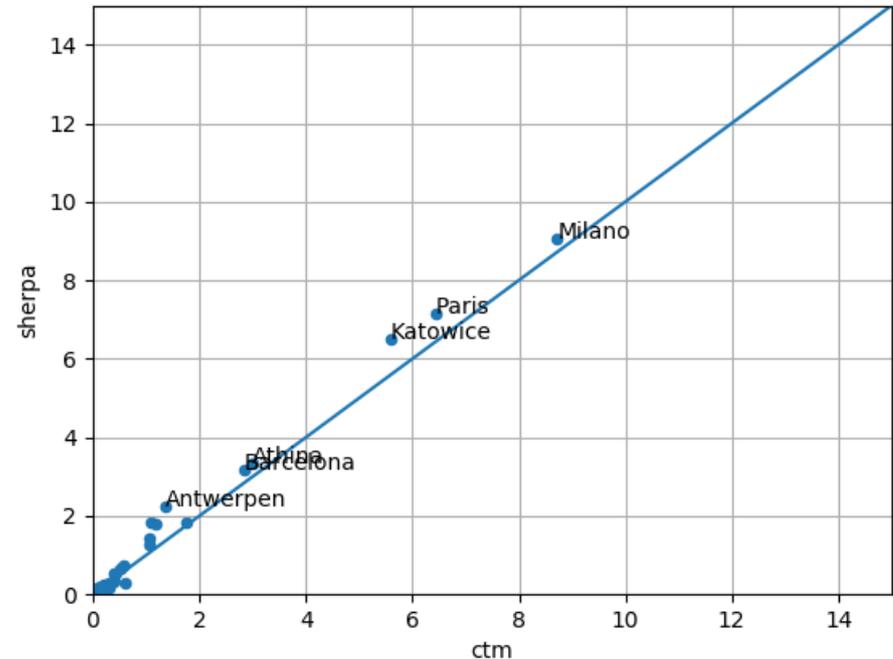
## Screening for High Emission Reduction Potentials on Air quality



comparison of delta values, PM<sub>2.5</sub> [mg/m<sup>3</sup>], scenario 22



comparison of delta values, PM<sub>2.5</sub> [mg/m<sup>3</sup>], scenario 34



CPU time: **minutes** to simulate 1 year

- Urban background levels
- Yearly PM<sub>2.5</sub>, PM<sub>10</sub>, NO<sub>2</sub>

# How to verify the robustness of the model ?



- Is my model robust for policy applications ?
- Which are the main sources of uncertainty?
  - Model coefficients?
  - Input data
  - ...
- Where should I put my efforts to further improve the model ?



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## Steps performed

### Select sources of uncertainty

- Source receptor model coefficients ( $\alpha$  and  $\omega$ ...see next slide)
- Input (emissions)
- Policies (decisions on the measures)

### Propagate the uncertainty with the model

- Simulating with SHERPA the combinations from the previous sources of uncertainties

### Compute the sensitivity analysis indicators (see next slides)

- How much SHERPA results depend on the different sources of uncertainty?

# Sources of uncertainty: model coefficients



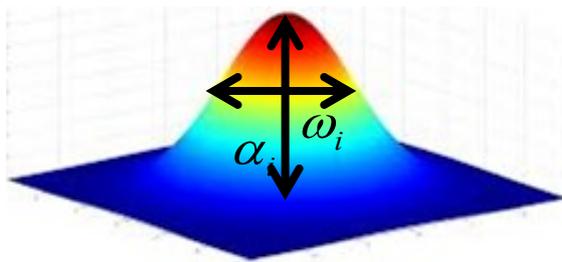
## 1. Source receptor equation

$$\Delta C_i = \sum_j^{N_{prec}} \sum_k^{N_{cell}} a_{i,j,k} \Delta E_{j,k}$$

Concentration change                      Emissions change

## 2. Coefficients of the source receptor:

$$a_{i,j,k} = \alpha_{i,j} (1 + d_{ik})^{-\omega_i}$$



## 3. Intervals for uncertainty sampling

| Coefficient | Nominal | Std   |
|-------------|---------|-------|
| W_NOX       | 1,97    | 0,02  |
| W_NH3       | 1,60    | 0,02  |
| W_PPM       | 2,32    | 0,018 |
| W_SO2       | 1,34    | 0,01  |
| a_NOX       | 0,05    | 0,005 |
| a_NH3       | 0,07    | 0,01  |
| a_PPM       | 1,97    | 0,04  |
| a_SO2       | 0,01    | 0,004 |

# Sources of uncertainty: emissions and policies



Input considered are emissions of:

- NO<sub>x</sub> (30%)
- NH<sub>3</sub> (50%)
- PPM (50%)
- SO<sub>2</sub> (10%)

Policies have been defined between Current Legislation and Maximum Feasible Reductions:

- 25%
- 50%
- 75%
- 100%

# Propagate uncertainty and compute indicators



|           |           |           |           |           |           |           |           |
|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
| 1.9728270 | 1.5775009 | 2.3517550 | 1.3347041 | 0.0481775 | 0.0546555 | 2.0381985 | 0.0103756 |
| 1.9615385 | 1.5843100 | 2.3227385 | 1.3426583 | 0.0492404 | 0.0879418 | 1.9297483 | 0.0022419 |
| 1.9615385 | 1.5775009 | 2.3517550 | 1.3347041 | 0.0481775 | 0.0546555 | 2.0381985 | 0.0103756 |
| 1.9728270 | 1.5843100 | 2.3517550 | 1.3347041 | 0.0481775 | 0.0546555 | 2.0381985 | 0.0103756 |
| 1.9728270 | 1.5775009 | 2.3227385 | 1.3347041 | 0.0481775 | 0.0546555 | 2.0381985 | 0.0103756 |
| 1.9728270 | 1.5775009 | 2.3517550 | 1.3426583 | 0.0481775 | 0.0546555 | 2.0381985 | 0.0103756 |
| ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         |
| ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         | ⋮         |
| 1.9401762 | 1.5955876 | 2.3189332 | 1.3345709 | 0.0482430 | 0.0860395 | 1.9534254 | 0.0086029 |
| 1.9401762 | 1.5955876 | 2.3189332 | 1.3212162 | 0.0400640 | 0.0860395 | 1.9534254 | 0.0086029 |
| 1.9401762 | 1.5955876 | 2.3189332 | 1.3212162 | 0.0482430 | 0.0503413 | 1.9534254 | 0.0086029 |
| 1.9401762 | 1.5955876 | 2.3189332 | 1.3212162 | 0.0482430 | 0.0860395 | 1.9433326 | 0.0086029 |
| 1.9401762 | 1.5955876 | 2.3189332 | 1.3212162 | 0.0482430 | 0.0860395 | 1.9534254 | 0.0195561 |

Combination of model coefficient, emissions and policy perturbation

SHERPA

→

|          |
|----------|
| 35.79223 |
| 34.99098 |
| 35.88423 |
| 35.75833 |
| 36.4947  |
| 35.7454  |
| ⋮        |
| ⋮        |
| 36.25576 |
| 35.23081 |
| 35.33825 |
| 36.18899 |
| 38.02076 |

Concentrations

**SENSITIVITY INDEX:** how much of the output variance depends on the variance of  $Y$  when perturbing  $i$

$$S_i = \frac{V[E(Y|X_i)]}{V(Y)}$$

**TOTAL EFFECT SENSITIVITY INDEX :** as the  $S_i$ , but considering the interactions terms

$$T_i = 1 - \frac{V[E(Y|X_{\sim i})]}{V(Y)}$$



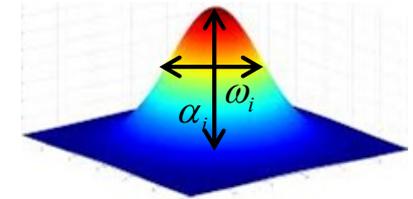
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# Model coefficient perturbation results



Analysis performed at the moment on 15 cities  
(here showing the results only for 3)



| Policy profile  | Total order Sensitivity Indices |               |               |               |               |               |               |               | Sum  |
|-----------------|---------------------------------|---------------|---------------|---------------|---------------|---------------|---------------|---------------|------|
|                 | $\omega$ -NOx                   | $\omega$ -NH3 | $\omega$ -ppm | $\omega$ -SO2 | $\alpha$ _NOx | $\alpha$ _NH3 | $\alpha$ _ppm | $\alpha$ _SO2 |      |
| 100-100-100-100 | 0.01                            | 0.01          | 0.11          | 0.01          | 0.24          | 0.09          | 0.16          | 0.35          | 0.98 |
| C: 75-25-25-25  | 0.04                            | 0.00          | 0.03          | 0.01          | 0.69          | 0.03          | 0.04          | 0.11          | 0.95 |
| B: 25-75-25-25  | 0.00                            | 0.05          | 0.06          | 0.01          | 0.13          | 0.45          | 0.09          | 0.20          | 0.99 |
| A: 25-25-25-75  | 0.01                            | 0.01          | 0.11          | 0.01          | 0.24          | 0.09          | 0.16          | 0.35          | 0.98 |

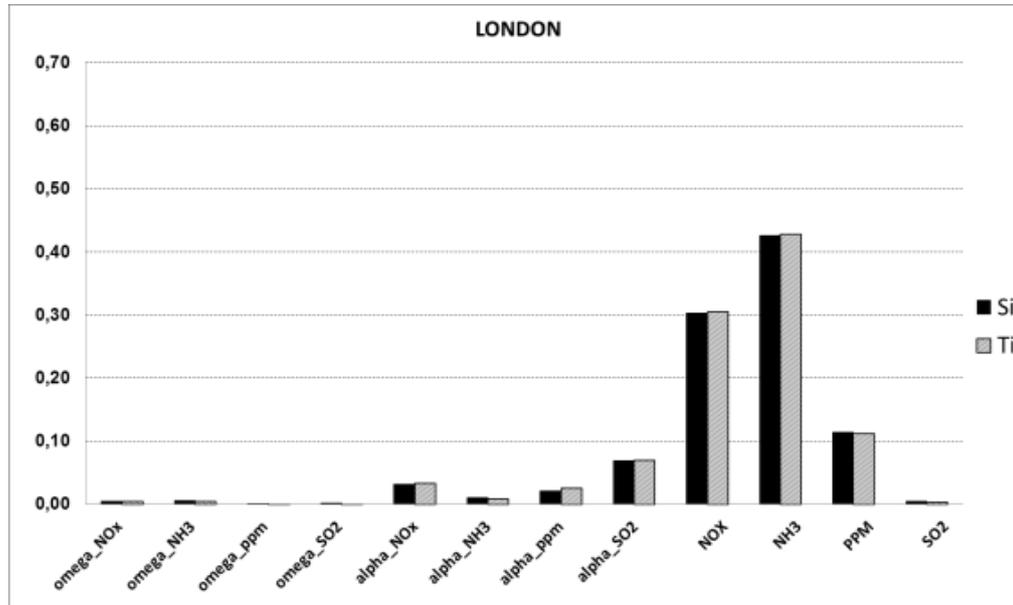
$\alpha$ 's show higher values: a good way to reduce the uncertainty on the  $\Delta C_i$  is to reduce uncertainty of the  $\alpha$ 's.

$\omega$ 's values have low 'influence' values

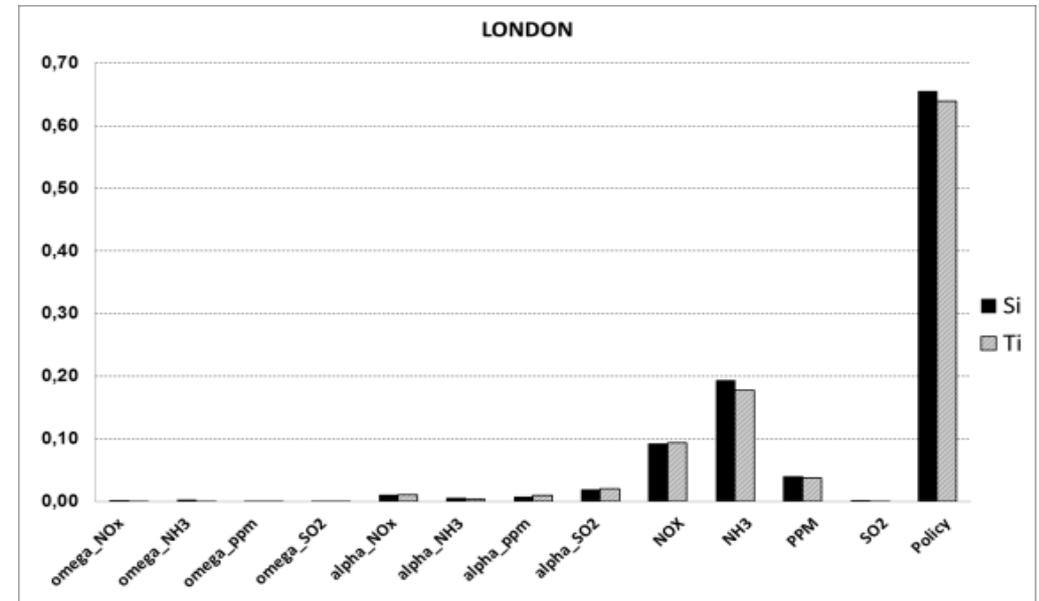
# Adding the input and policy perturbation



## London Results



Emission uncertainty is higher than Model coefficient uncertainty

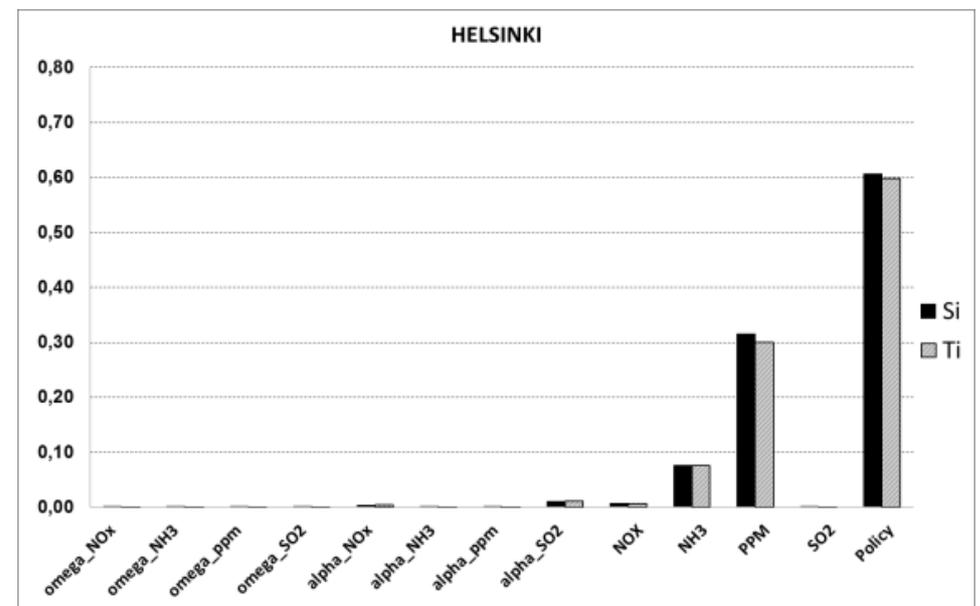
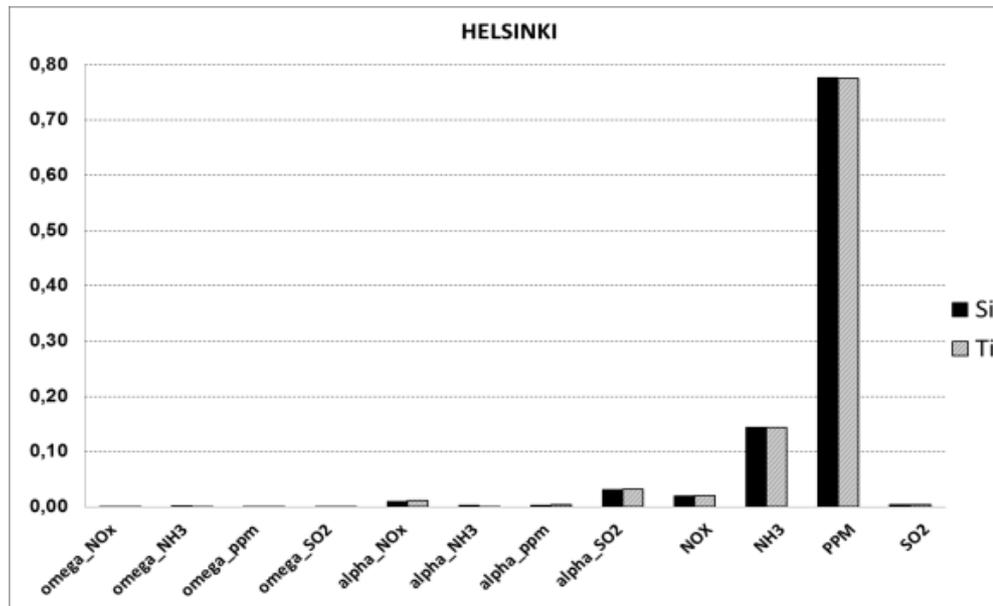


As wished, the policy variability is higher than the uncertainties

# Perturbing the model, coefficient and policy



## Helsinki

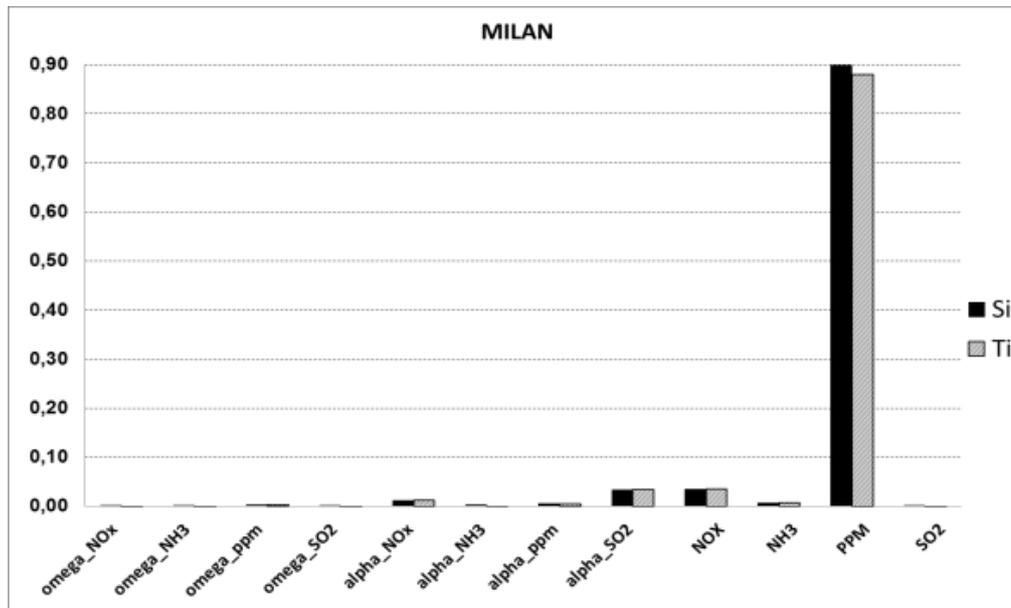


Same as London, with higher sensitivity to primary PM

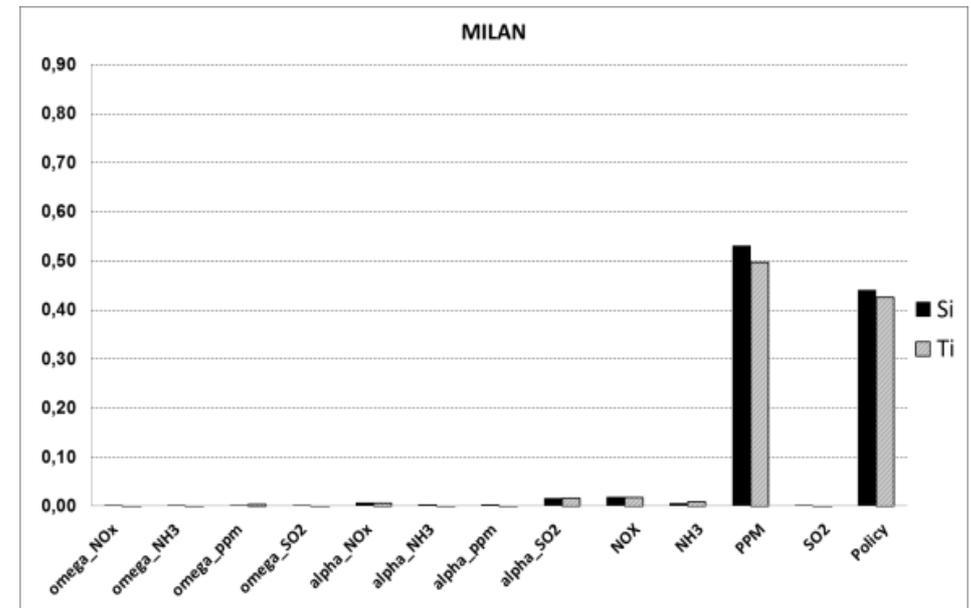
# Perturbing the model, coefficient and policy



## Milan



Same as London, with higher sensitivity to primary PM



Policy and emission uncertainty are at a similar level



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# Conclusions

Sensitivity analysis allows to:

- Understand the key source of uncertainty (model? input? policy?)  
(In our case input are very important!)
- Understand where to put effort in model development
- Identify that the 'policy' is an important factor (but in some cases further work is needed to improve input data)