

UNCERTAINTY AND VALIDATION OF URBAN SCALE MODELLING SYSTEMS APPLIED TO SCENARIO ANALYSIS IN TUSCANY, ITALY

Matteo Carpentieri¹, Paolo Giambini¹, and Andrea Corti²

¹Dipartimento di Energetica “Sergio Stecco”, Università degli Studi di Firenze, Italy

²Dipartimento di Ingegneria dell’Informazione, Università degli Studi di Siena, Italy

INTRODUCTION

The work presented in this paper has been carried out in the framework of the MoDiVaSET project (MODellistica Diffusionale per la VALutazione di Scenari Emissivi in Toscana). The project, funded by the Regional Administration of Tuscany, is aimed at developing modelling techniques for simulating atmospheric dispersion of PM₁₀, NO_x and SO_x in the metropolitan area of Florence, Prato and Pistoia.

The objective of the project is to develop an integrated meteorological and dispersion modelling system which can be used by administrators and policy makers in order to evaluate different future emission scenarios. With this aim, the project also included several long-term (1-year long) dispersion modelling applications and a detailed evaluation study, which is often neglected in similar applications, despite its importance.

The study area is 49×40 km² sized. Emissions were retrieved from the Tuscan Regional Emission Source Inventory (*RSE-RT, 2001*); three source categories have been initially considered: main point sources (87 industrial stacks), main line sources (A1 and A11 motorways) and all remaining ones (treated as 1×1 km² grid area sources).

AIR QUALITY MODELLING

Dispersion models

The simulations were mainly carried out using ADMS-Urban. In order to compare results from different models, simulations were also performed by means of CALGRID (grid source), CALPUFF (point sources), CALINE4 (line sources), and SAFE_AIR_II (point and line sources). Further modelling options (full chemistry simulation, street canyon, ...) are being investigated by using CAMx, OSPM and other models.

The full 2002-year time period according to a 1-hour time step was chosen, thereby all models were applied in a long-term mode.

Measurements from six meteorological stations within the study domain (Baciacavallo, Monte Morello, Empoli-Ridolfi, Montale, Firenze-Ximeniano and Peretola Airport) were used as input, while vertical profiles of wind and temperature were retrieved from the RAMS forecasting system archive of CNR-IBIMET/LAMMA (see also Corti, A. *et al.*, 2006). A suitable scaling to the 1×1 km² final working resolution was then performed by using the CALMET meteorological model. The results were directly used as input for the CALGRID and CALPUFF simulations, while a further elaboration proved to be necessary for the other models. The domain was divided in 32 sub-domains for the CALINE4 simulations, using a single CALMET point for each sub-domain. 8 of these points were also used as input for the SAFE_AIR_II own meteorological preprocessor. The same methodology was applied to ADMS-Urban, but using only one CALMET point.

Fig. 1; Monitoring stations in the MoDiVaSET domain.

The statistical indices used for the validation exercise are derived from the BOOT software (Hanna, S.R., 1989) and the Model Validation Kit (MVK, Olesen, H.R., 1995, 2005). Furthermore, two other indices originally proposed by Poli, A.A. and M.C. Cirillo (1993) were used. The resulting statistical set is similar to that applied by Canepa, E. and P.J.H. Builtjes (2001) for the validation of the SAFE AIR code in complex terrain. These statistical indices are: mean (MEAN), bias (BIAS), fractional bias (FB), standard deviation (SIGMA), fractional standard deviation (FS), linear correlation coefficient (COR), fraction within a factor of 2 (FA2), normalised mean square error (NMSE), weighted normalised mean square error of the normalised ratios (WNNR), and normalised mean square error of the distribution of normalised ratios (NNR).

Chang, J.C. and S.R. Hanna (2004) introduced acceptability criteria for some of the statistical indices provided by the BOOT software, basing on an extensive literature review. They proposed the following criteria for a “good” model: $FA2 > 0.5$; $-0.3 < FB < 0.3$; $NMSE < 4$.

A first simple comparison between the results was done using the calculated annual average concentrations. as reported in Table 1, for some of the monitoring stations.

Table 1. Comparison between NO_2 , PM_{10} and SO_2 annual mean concentrations calculated by CALGRID-CALPUFF-CALINE4 (CGPL), CALGRID-SAFE AIR (CGSA) and ADMS-Urban (ADMS), and annual mean concentrations measured by the monitoring stations (Meas); [$\mu g/m^3$]

Station (NO_2)	CGPL	CGSA	ADMS	Meas
FI – Bassi	30.3	26.1	18.8	37.8
FI – Boboli	26.6	26.1	20.3	30.7
FI_Montelupo – Pratelle	11.1	8.2	4.2	28.6
PT – Montale	20.9	15.2	10.6	32.2
Station (PM_{10})	CGPL	CGSA	ADMS	Meas
FI – Bassi	3.6	3.2	2.4	42.6
FI – Boboli	2.9	2.4	2.7	37.6
FI_Scandicci – Buozzi	3.1	2.3	2.6	42.7
PT – Montale	1.7	1.2	1.1	53.6
Station (SO_2)	CGPL	CGSA	ADMS	Meas
FI – Bassi	2.4	1.9	1.3	3.8
FI – Boboli	2.1	1.4	1.5	2.9
FI_Empoli – Ridolfi	1.0	0.8	1.3	4.8
PT – Montale	0.8	0.5	0.4	3.1

The statistical indices described previously were then applied to the calculated/measured annual mean concentrations. Unfortunately, the low number of measuring points reduced the significance of the statistical analysis. Results are reported in Table 2.

The statistical analysis confirms the previous results. Good performances are obtained for nitrogen and sulphur oxides, while performance values for PM_{10} are rather low. The acceptability criteria proposed by Chang, J.C. and S.R. Hanna (2004) are not completely verified (although SO_2 and NO_2 values are quite close), but it should be noted that in this case

we are using concentrations paired in space. Further, Chang and Hanna's results are referred to research level measures, while in this case routinely monitored data are used.

Table 2. Statistical indices calculated for NO₂, PM₁₀, and SO₂ annual mean concentrations

NO ₂	mean	bias	fb	sigm	fs	cor	fa2	nmse	wnnr	nne
Meas	46.12	0.00	0.00	16.61	0.00	1.00	1.00	0.00	0.00	0.00
CGPL	24.58	21.54	0.61	7.60	0.74	0.25	0.60	0.65	0.66	0.41
CGSA	18.85	27.27	0.84	8.04	0.70	0.30	0.27	1.12	1.12	0.79
ADMS	20.34	25.78	0.78	10.95	0.41	0.62	0.53	0.89	0.90	0.81
PM ₁₀	mean	bias	fb	sigm	fs	cor	fa2	nmse	wnnr	nne
Meas	41.51	0.00	0.00	8.58	0.00	1.00	1.00	0.00	0.00	0.00
CGPL	2.70	38.82	1.76	0.75	1.68	0.11	0.00	14.11	14.11	12.96
CGSA	2.17	39.34	1.80	0.83	1.65	0.10	0.00	17.98	17.98	16.60
ADMS	2.65	38.87	1.76	1.54	1.39	-0.28	0.00	14.51	14.51	12.48
SO ₂	mean	bias	fb	sigm	fs	cor	fa2	nmse	wnnr	nne
Meas	3.07	0.00	0.00	0.35	0.00	1.00	1.00	0.00	0.00	0.00
CGPL	1.98	1.09	0.43	0.84	-0.84	-0.16	0.67	0.35	0.38	0.35
CGSA	1.41	1.66	0.74	0.66	-0.62	-0.09	0.50	0.77	0.77	0.72
ADMS	1.37	1.71	0.77	0.52	-0.40	-0.37	0.50	0.82	0.82	0.73

Uncertainty analysis

Uncertainty analysis methods can be classified in two categories. The widely used approach can be referred to as “bottom-up”, and it attempts to quantify the single error sources, and then to calculate the overall error by means of statistical techniques such as error propagation analysis, sensitivity analysis, sampling methods and Monte Carlo methods. This approach is the most applied in literature, although the error quantification is often arbitrary. For this reason, Colville, R.N. et al. (2002) introduced an alternative approach, referred to as “top-down”, which does not consider the single error sources, but the overall error is quantified by means of a high number of measures sufficiently representative of the phenomenon. This latter technique was used in this work: the uncertainty is quantified by means of the estimation of the model *precision*, calculated after removing the bias, and normalised using the appropriate limit value for the considered pollutant. Eventually the precision is calculated using the logarithmic mean square deviation of the modelled values with respect to the measured ones.

It was not easy to perform an uncertainty analysis, given the systematic underestimation resulting from the models applications. This is confirmed by the calculation of the “accuracy” as recommended by the European legislation (1999/30/EC and 2000/69/EC), which gives not acceptable results (not reported for brevity; again, for a complete overview see Carpentieri, M., 2006). More useful, in this case, is the methodology proposed by Colville, R.N. et al. (2002), because it allows the removal of the systematic underestimation effect (caused by other factors, see Conclusion).

The calculated “precision” values are reported in Table 3.

Table 3. Model precision calculated following the methodology by Colville, R.N. et al. (2002)

	NO2	PM10	SO2
CGPL	41 %	34 %	54 %
CGSA	37 %	37 %	61 %
ADMS	51 %	78 %	73 %

CONCLUSION

The obtained results point out the importance of including the following critical factors:

- Regional background concentrations: looking at the results a systematic bias between calculated and measured concentration proved to exist; the models systematically underestimated pollution levels; however, this did not affect the analysis of future scenarios (starting from the hypothesis that background concentrations do not change).
- Smaller scale effects: monitoring stations are often located in complex environments; this implies a decrease in the effectiveness of validation studies; a possible solution would be to include small scale effects (e.g. street canyon modelling) in order to increase the resolution of the models; otherwise, a detailed study on the representativeness of the monitoring sites appears to be necessary.
- Secondary pollution: primary PM₁₀ levels are only a small part of the total PM₁₀ concentrations; besides high regional background levels, much of the urban PM₁₀ is actually produced by chemical transformations and other physical mechanisms (for example, resuspension).

All these issues strongly affected the evaluation work. However, this does not alter the validity of the scenario analysis, because it is based on the differences between calculated primary pollutants concentrations deriving from the considered emissions. Modelling results can be trusted on the basis of the evaluation work, despite the critical factors listed above.

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