# Stage 4 Review and Assessment for the London Borough of Richmond upon Thames



### **University of London**

April 2002

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#### Appendix E

#### 1 Model Uncertainty Assessment

#### 1.1 Introduction

This appendix describes the application of Bayesian Monte Carlo (BMC) analysis to the ERG model developed to predict present and future concentrations of annual average  $NO_2$  in London. Model uncertainties arise because of limited scientific knowledge, limited ability to assess the uncertainty of model inputs, for example, emissions from vehicles, poor understanding of the interaction between model and/or emissions inventory parameters, sampling and measurement error associated with  $NO_X$  and PM10 sites in London and whether the model itself completely describes all the necessary atmospheric processes. The application of the BMC technique here results in the reduction in uncertainties predicted through the additional information provided by the measurements themselves.

#### 1.2 Uncertainty Assumption in Model Input Parameters

Selection of the uncertainty of input variables are obtained through access to published literature, the opinions of experts in the field, and through the assessment of relationships used within the model. A summary of the assumptions made for the model are given in the table below:

Table 29 Uncertainty Assumptions  $(1 \sigma)$  use for the Uncertainty Predictions

	(%)
Road Traffic Emissions	30
Other Emissions	50
London + Rural NO <sub>X</sub> Contribution	10
Pollution Climate Mapping (NO <sub>X</sub> )	11
NOx-NO2 Relationship	10
Roadside Dispersion	20

#### 1.3 Bayesian Monte Carlo Analysis

In Monte Carlo analysis, the model is run with the input variables varied simultaneously and independently of each other and a resulting probability distribution of the output information, obtained. Bayes' theorem is then applied to derive a final uncertainty estimate, by assigning a high probability to those predictions that agree with the measurements and a low or zero probability to those, which do not. The application of probabilities to the model prediction uses the likelihood function (Equation 1) and results in the best estimate of overall model uncertainty.

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$$L(Y_k \mid O) = \frac{1}{\sqrt{2\pi\sigma_e}} \exp\left(-\frac{1}{2}\left[\frac{O-Y_k}{\sigma_e}\right]^2\right)$$
(1)

A mathematical summary of BMC is given below. From Bayes' theorem the final probability of model output is defined by equation 2 as

$$p(Y_k \mid O) = \frac{L(Y_k \mid O)p(Y_k)}{\sum_{j=1}^{N} L(Y_j \mid O)p(Y_j)}$$
(2)  
1.4 Results at Background

A BMC uncertainty analysis was carried out for annual average NO<sub>2</sub> concentration throughout London.

The prior and posterior distributions for an average of the measurement sites in London are included in Table 30. The application of BMC analysis reduces the final uncertainty giving a standard deviations in this case are 2.0 ppb (8.5 %).

The BMC analysis was then applied for 5 sites individually and the results summarised in Table 31. Again BMC analysis results in a significant reduction in  $\sigma$  providing a reduction in uncertainty. The average  $\sigma$  for the 5 sites was 1.8 ppb.

Table 30 Final uncertainty and measured annual mean  $NO_2$  concentrations (ppb) at all sites in London for 1998

Average Model		Uncertainty %	
Prediction (ppb)	σ (ppb)	-	Measured Result (ppb)
23.6	2.0	8.5	23.2

Site Location	Final Predict	Model ion (ppb)	Uncertainty %	Measured Results (ppb)
		σ (ppb)		
Bridge Place	30.6	2.2	7.2	30.2
Bexley 2	19.1	1.5	7.8	18
<b>Tower Hamlets 1</b>	24.1	1.8	7.5	24.6
West London	26.8	2.0	7.5	26.8
Sutton 2	18.6	1.4	7.5	19.8

**Table 31** Final uncertainty and measured annual mean NO2 Concentrations for separate

 Sites in London for 1998

#### 1.5 Results at Roadside

Predictions of the concentration of  $NO_2$  at roadsides throughout London have shown a high sensitivity to the pass/fail standard of 21 ppb. These predictions are crucial to the development of air pollution control, through local authority action plans, and it is therefore essential to completely understand the uncertainty associated with them. Only then will the strengths and weaknesses of the predictive process be understood enough for decision-makers to make informed policy judgements. It is the uncertainties associated with these predictions, which are the subject of this appendix.

Monte Carlo modelling techniques have been used to calculate the uncertainties associated with roadside  $NO_2$  predictions. It also includes a full sensitivity analysis to determine the most important input variables to the model. Specific tests include the uncertainties associated with flows and emissions from LGVs, HGVs and buses, vehicle speed, the dispersion model, and the pollution climate mapping technique, used for calculating background concentrations.

In *Monte Carlo* analysis, the input variables are varied simultaneously and independently of each other, and the effect on important outputs assessed. The model uncertainty, relating to the input parameters, is calculated by treating them as random variables. By studying the resulting probability distribution of the output (i.e. the concentration or emission estimate), information is obtained regarding the model uncertainty.

The original study has focused on Marylebone Road for a base year of 1997 for meteorology and atmospheric chemistry and uses the London Transportation Studies (LTS) traffic model. Further uncertainty assessments have also been undertaken for an "average road' in central and outer London, as well as a 'Motorway' in outer London.

The sensitivity analysis revealed that roadside  $NO_x$  predictions are mostly sensitive to the assumptions regarding HGV emissions and flows and the dispersion model

used to predict roadside concentrations. For the prediction of  $NO_2$ , the  $NO_x$ - $NO_2$  relationship used is the most important factor. Table 32 below shows how each input data or modelling method affects the final concentration, for the Marylebone road example.

Table 32 The Relative Importance of Model Parameters in Predicting NO2	at Marylebone
Road	

Model Parameter	Relative Importance 2005	Relative Importance 1997
	(% of mean at $2\square$ )	(% of mean at $2\square$ )
NO <sub>X</sub> -NO <sub>2</sub> relationship	13.9	11.9
HGV emissions	7.9	8.1
Dispersion model	7.3	6.8
HGV flow	5.5	5.5
LGV emissions	4.2	4.7
LGV flow	4.2	4.7
Vehicle speed	3.6	2.1
Background mapping	1.8	1.7
Bus emissions	1.2	0.9
Bus flow	0.6	0.4

For 1997, NO<sub>x</sub> was predicted to be 258 +/- 83 ppb and NO<sub>2</sub> 47 +/- 10 ppb, at two standard deviations – equivalent to the 95 % confidence interval. These statistics assume that the resultant distribution is normal.

The overall uncertainty of NO<sub>2</sub>, which corresponds to 22 %, is less than that for NO<sub>X</sub> (32 %). This feature is a result of the non-linear NO<sub>2</sub> relationship, which is quite insensitive to NO<sub>x</sub> concentrations, implying that a stated NO<sub>X</sub> uncertainty is a better indication of the quality of a prediction.

Measurements for the Marylebone Road site for  $NO_x$  and  $NO_2$  are within the uncertainty limits calculated here.  $NO_x$  was between 213 and 229 ppb and  $NO_2$  between 44 and 48 ppb for 1997. The range reflects the two different monitoring techniques used at the Marylebone site.

Similarly, for 2005, NO<sub>x</sub> is estimated to be 117 +/- 35 ppb and NO<sub>2</sub> 33 +/- 7 ppb, at two standard deviations – equivalent to the 95 % confidence interval. It can therefore be concluded that with a probability of 95 % the true value lies within the ranges given above. This would indicate that, despite the calculation of uncertainty associated with the 2005 predictions, the NO<sub>2</sub> concentration always exceeds 21 ppb and therefore Marylebone Road will exceed the AQS objective. This may not always be the case however and with a prediction whose range straddles 21 ppb, a decision must be made concerning the approach to be taken. For example, a prediction of 20 +/- 2 ppb could be considered a pass or a fail.

It is further concluded that the prediction of  $NO_2$  concentrations in London depend most on the  $NO_x$ - $NO_2$  relationship used and the traffic data for HGVs. It is flows of, and emissions from, HGVs and buses that become more important in the future, as emissions from these vehicles will make up a greater proportion of the total.

The results from the analysis of a further three roads is given in Table 33. These represent an average road at a central and outer location and an average motorway in outer London. The flow and percent HGV for the average road was derived from all 10,000 roads in the LTS 91 network.

Road Type/Location	Total vehicle flow	Percent HGV	Uncertainty (% of mean at 2σ)
Average road (central	17,000	9	16
London)			
Average road (outer	17,000	9	18
London)			
Motorway (outer London)	80,000	9	21

Table 33 NO2 Uncertainty Estimates for Typical Roads in London in 2005

## Our best estimate of the uncertainty in annual mean $NO_2$ predictions is therefore +/- 16-21 % at two standard deviations.

It has not been possible to quantify the uncertainty of PM10 predictions in the same way as NO<sub>2</sub>. This is because the uncertainty of the measurement techniques themselves and the sources and sinks of particles has not been well established. *However, it is reasonable to expect that the uncertainty in PM10 predictions is larger than NO*<sub>2</sub>.

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