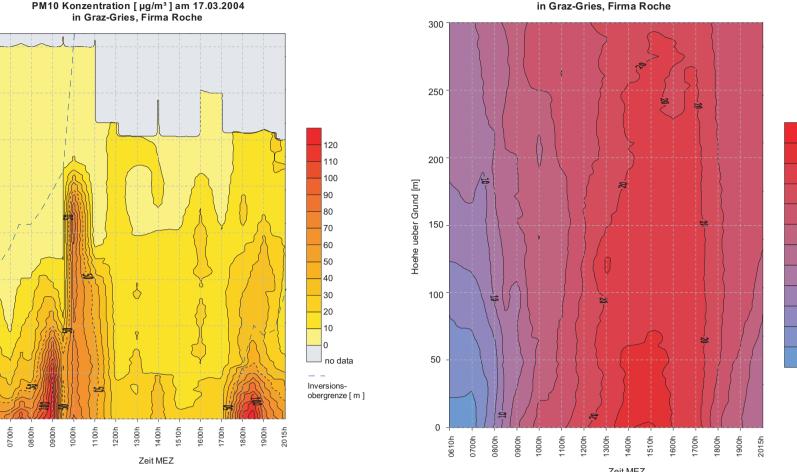


### **Primary particles**

Emerge from burning processes, mechanical abrasion of tyres, brakes, tarmac, etc. or natural sources (pollen, crushing rock, soil, etc.)



The aim of the prediction model is to give a forecast of the average PM10 load of the subsequent day. Multiple linear regression proved to be a reliable approach.

### Secondary particles Arise from aerially pollutants.

## **EU Directive**

According to the EU framework directive 96/62/EC the limit value for the daily average is  $50\mu g/m^3$  and must not be exceeded on more than 30 days a year. The corresponding threshold for the annual average is  $40\mu g/m^3$ .

# The PM10 Problem

For several years the PM10 concentration has been measured and analyzed in Europe.



*Figure 2.1: Graz at a period of stationary temperature inversion.* 

Mainly the adverse meteorological conditions are responsible for the high PM10 loads in basin areas south of the Alps.

*Figure 2.4: The balloon probe shows to what extent the dissolution* of inversion yields a decline of PM10. (Source ZAMG Styria.)

### 2.1.2 Traffic

Traffic plays a crucial role for the PM10 problem (exhaust, abrasion, dispersion of dust).

tes.

Figure 2.5: On Sun- and Ho-

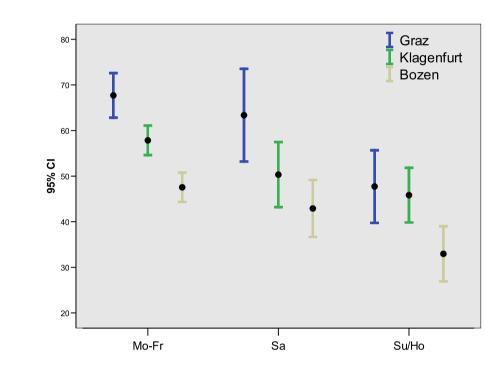
lidays the PM10 concentration

is considerably lower than on

working days (25–30%). This

may be explained by the redu-

ced traffic loads at the three si-



#### Wind, Precipitation and Frost 2.1.3

The emergence of wind and precipitation have positive effects on the PM10 concentration. Contrarily frost causes an increase of the PM10 values. This may be explained via heightened domestic fuel.

Our prediction models are based on linear regressions.

$$\sqrt{PM_{10}} = \sum_{k} b_k \cdot x_k + \sum_{l} b_l \cdot p_l + \epsilon \quad \text{with} \quad \epsilon \sim N(0, \sigma^2) \,.$$

A square root transformation of the response PM10 is necessary in order to assure that the model assumptions are not violated. For our models we use up to 7 input variables:

type	explanation
metric	PM10 24-h moving average from 12.00–12.00
categorial	Mo–Fr, Sa, Su/Ho
0/1	average temperature $> 0, \le 0$
categorial	October – March
metric	average wind speed of the subsequent day (to be forecasted)
0/1	precipitation of the subsequent day (to be forecasted)
metric	average temperature difference to a 300–400m higher reference test point (to be forecasted)
	metric categorial 0/1 categorial metric 0/1

The variables  $x_1 - x_4$  are available at the assigned time for prediction. The variables  $p_1 - p_3$  have to be forecasted.

#### **Quality of the model** 3.2

The models show a corrected  $R^2$  between 54% and 64%. The input variables have been selected via a stepwise procedure.

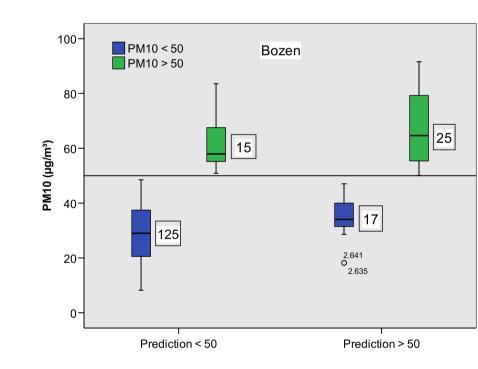


Figure 3.1: From 1.10.2005-31.3.2006 150 (=82%) out of 182 predictions have been categorized correctly (exceedance/no *exceedance). The* misclassified values were still close the classification limit  $(50\mu g/m^3).$ 

- low wind velocities
- rare days with precipitation
- stationary temperature inversions

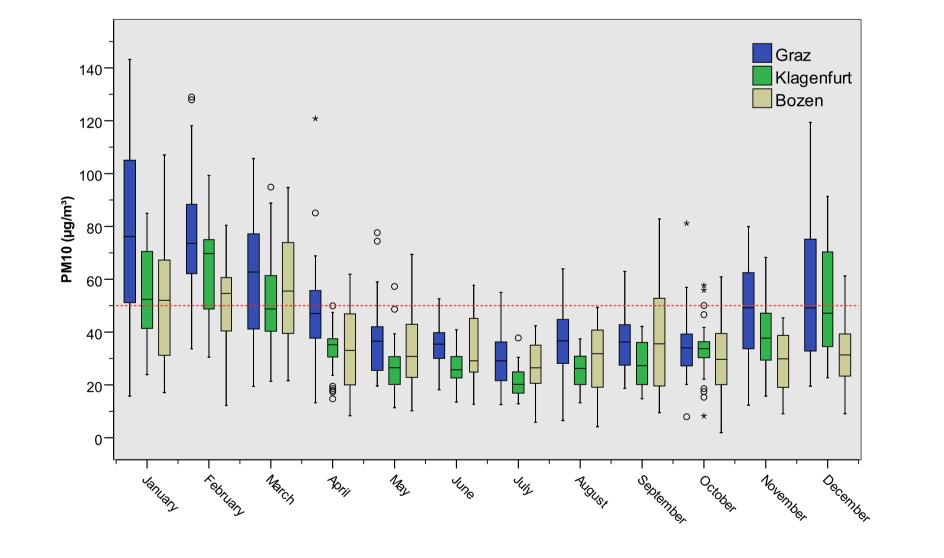
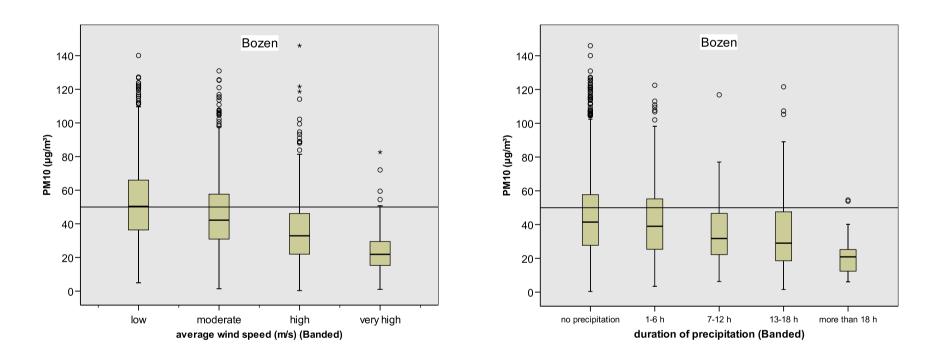


Figure 2.2: Most exceedances of the limit value occur during the *winter period (October till March).* 

#### **Exploratory Analysis** 2.1

#### **Temperature Inversion** 2.1.1

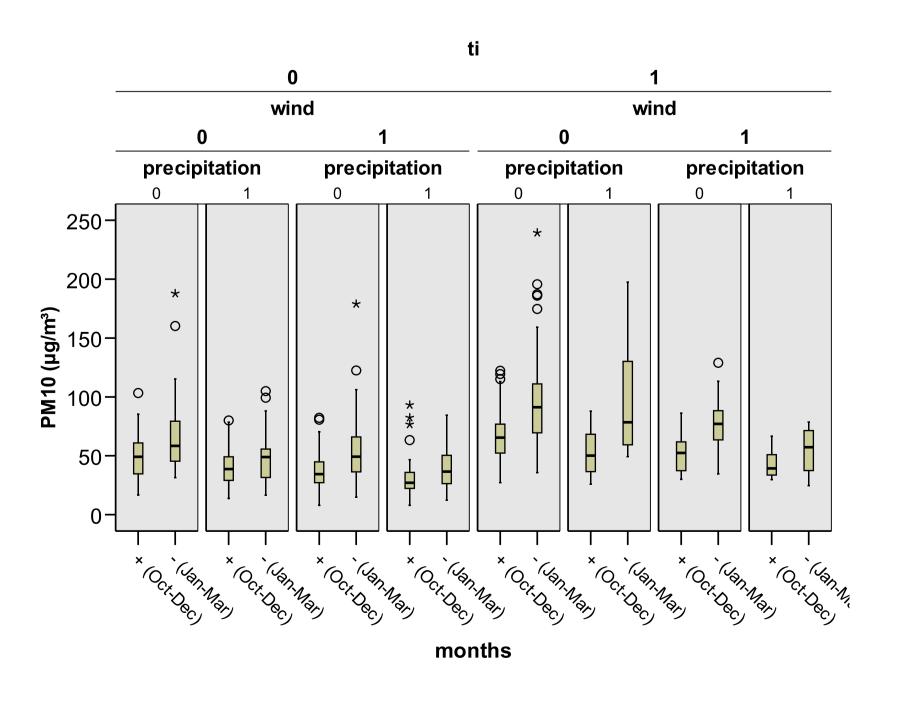
Due to the reduced air exchange we observe the highest PM10 concentration at stationary temperature inversions.



*Figure 2.6: The influence of wind (left) and precipitation (right) in* Bolzano. Here the wind velocity has the biggest influence on PM10.

#### 'Saturation Effect' 2.1.4

Under constant meteorological conditions the PM10 values become considerably higher in course of the winter period. A possible explanation might be that the defilement of deposited road grit is increasing during the winter months.



### **Test Run in Graz** 3.3

Our prediction model for Graz has been tested within three winter periods (2004/05–2006/07). The necessary meteorological forecasts were provided by the ZAMG Steiermark.

The forecasts can be found on http://www.feinstaubfrei. at/htm/ampel.php. Thereby our prediction model proved its worth as reliable monitoring tool.

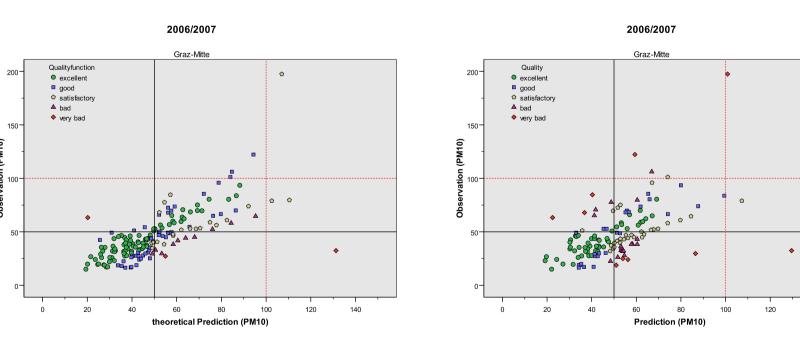


Figure 3.2: Comparison between observation and prediction. If the meteorological parameters are known the predictions are satisfactory in about 90% of the cases (left panal). During the test run where we used the meteorological forecasts approximately 80% (right panel) met our demands.

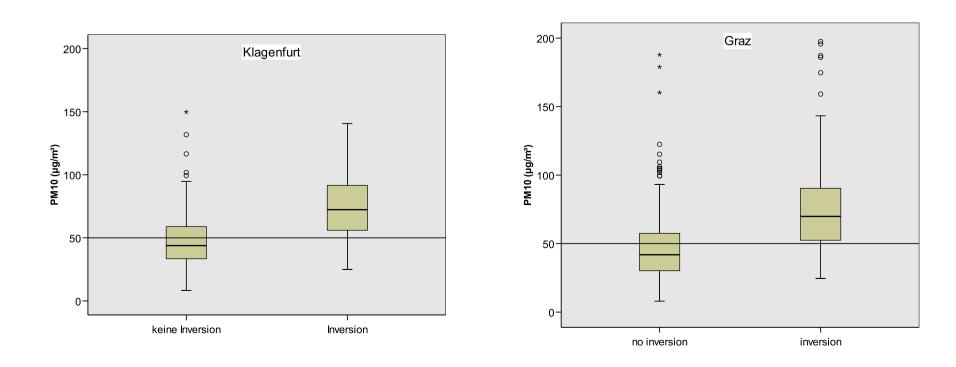


Figure 2.3: We measure temperature inversion with respect to Göriach and Kalkleiten (390/360m above ground). Temperature in*version is indicated if* temp(*Klagenfurt*)-temp(*Göriach*) *respectively* temp(*Graz*)-temp(*Kalkleiten*) is negative.

*Figure 2.7:* PM10 *load in Graz under specific meteorological scenarios: Inversion: 0=no inversion; Wind: 0=wind speed below median; Precipitation:* 0=no precipitation. Value=1 describes the complements.

### Literature

Hörmann, S., Pfeiler, B., Stadlober, E. (2005): Analysis and Prediction of Particulate Matter PM10 for the Winter Season in Graz, Austrian Journal of Statistics (34) 307–326.

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