

# Graz air pollution

# Methods

1. Base additive
2. Base multiplicative
3. Base multlicative cutoffs
4. Features (f3)
5. f3\_cutoffs
6. Features2 (f4)
7. f4\_cutoffs
8. features3
9. crossed\_feaures

```
features=['Temp', 'RH', 'Pressure',  
          'Windddirection', 'Windspeed', 'Precip']  
  
features2=['weekday_Friday', 'weekday_Monday',  
          'weekday_Saturday', 'weekday_Sunday', 'weekday_Thursday',  
          'weekday_Tuesday', 'weekday_Wednesday']  
  
features3 = ['season_fall', 'season_spring',  
            'season_summer', 'season_winter']  
]
```

## 7. feature cross

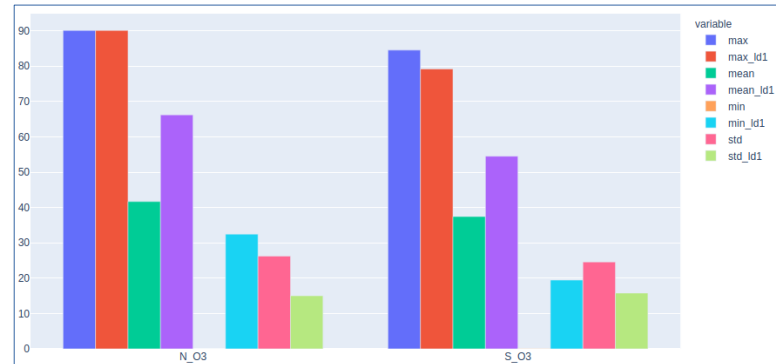
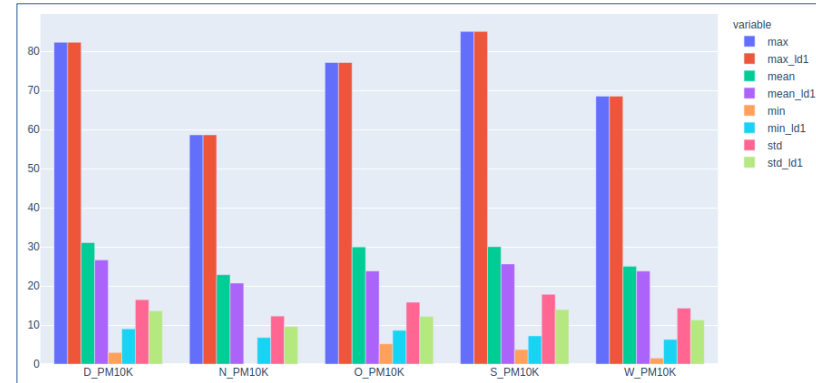
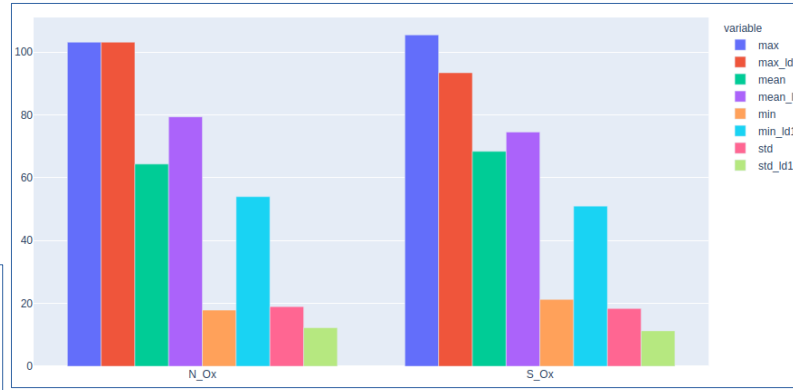
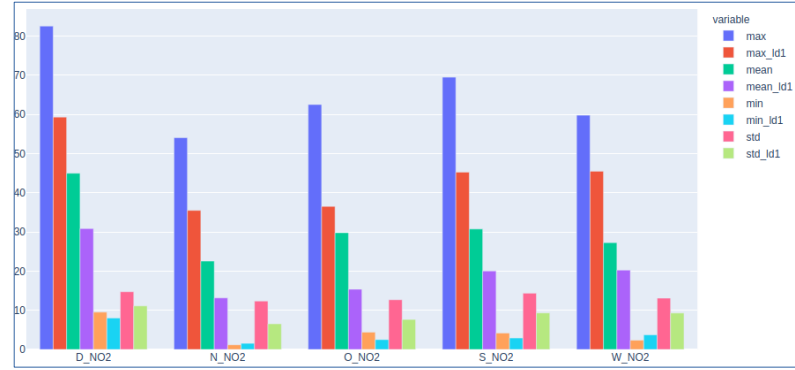
```
weekday_convert = {  
    'weekday_Sunday': 0,  
    'weekday_Monday': 1,  
    'weekday_Tuesday': 2,  
    'weekday_Wednesday': 3,  
    'weekday_Thursday': 4,  
    'weekday_Friday': 5,  
    'weekday_Saturday': 6  
}  
  
for c in features2:  
    df_org[c] = [weekday_convert[c] if i else i for i in df_org[c]]  
df_org['dayoftheweek'] = df_org[features2].sum(axis=1)  
df_org[['dayoftheweek']]
```

```
season_convert = {  
    'season_spring': 1,  
    'season_summer': 2,  
    'season_fall': 3,  
    'season_winter': 4  
}  
  
for c in season_convert.keys():  
    df_org[c] = [season_convert[c] if i else i for i in df_org[c]]  
df_org['season'] = df_org[season_convert.keys()].sum(axis=1)  
df_org[['season']]
```

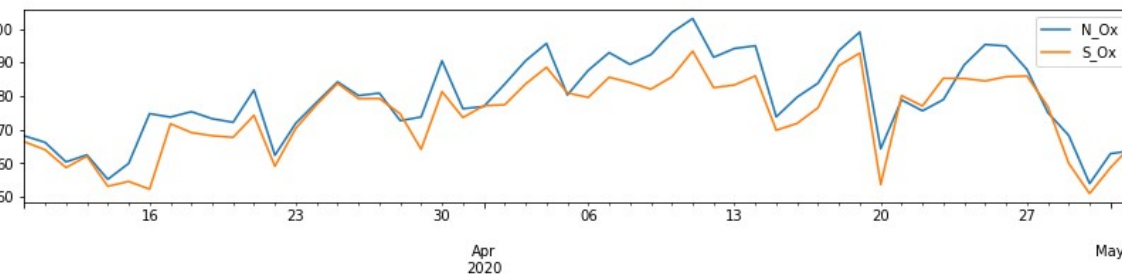
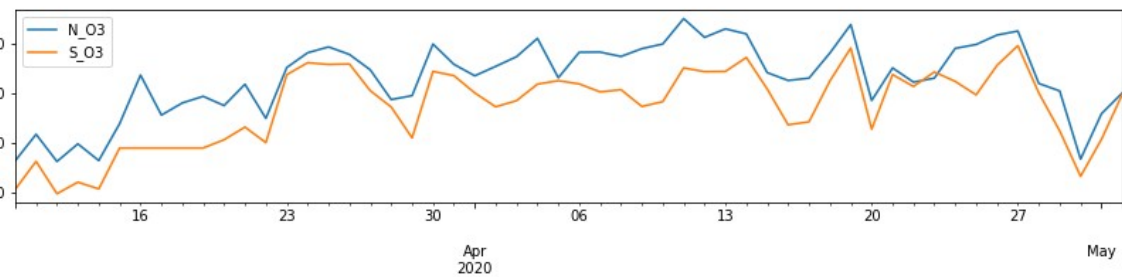
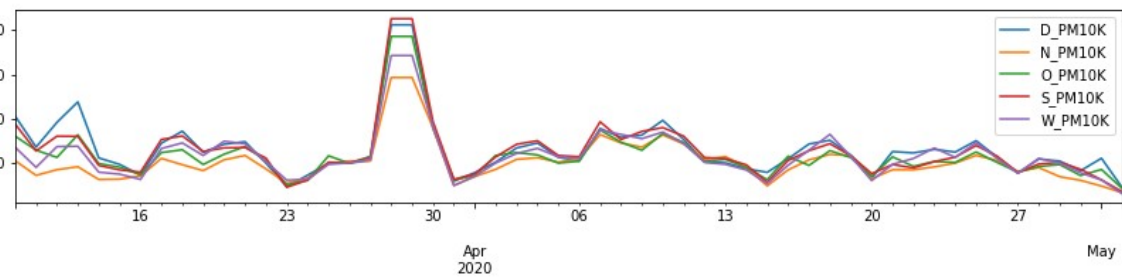
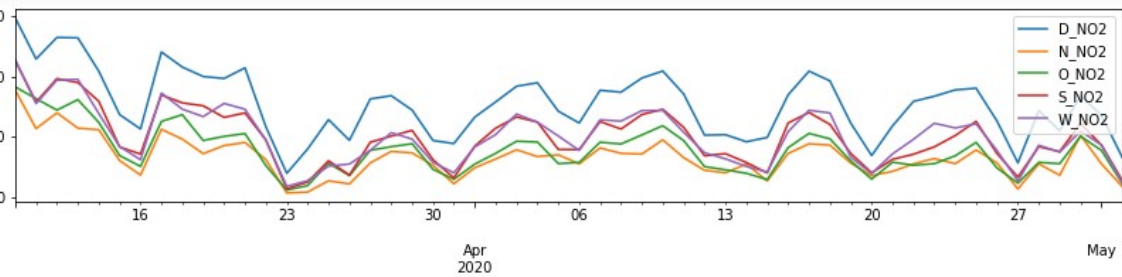
```
df_org['cross1'] = df_org['season']*df_org['dayoftheweek']  
df_org['cross2'] = df_org['season']**2 + df_org['dayoftheweek']**2  
df_org['cross3'] = df_org['season']**2  
df_org['cross4'] = df_org['dayoftheweek']**2  
crosses = ['cross1', 'cross2', 'cross3', 'cross4']  
df_org[crosses]
```

# Lockdown1 vs. train

# Lockdown1 vs. train

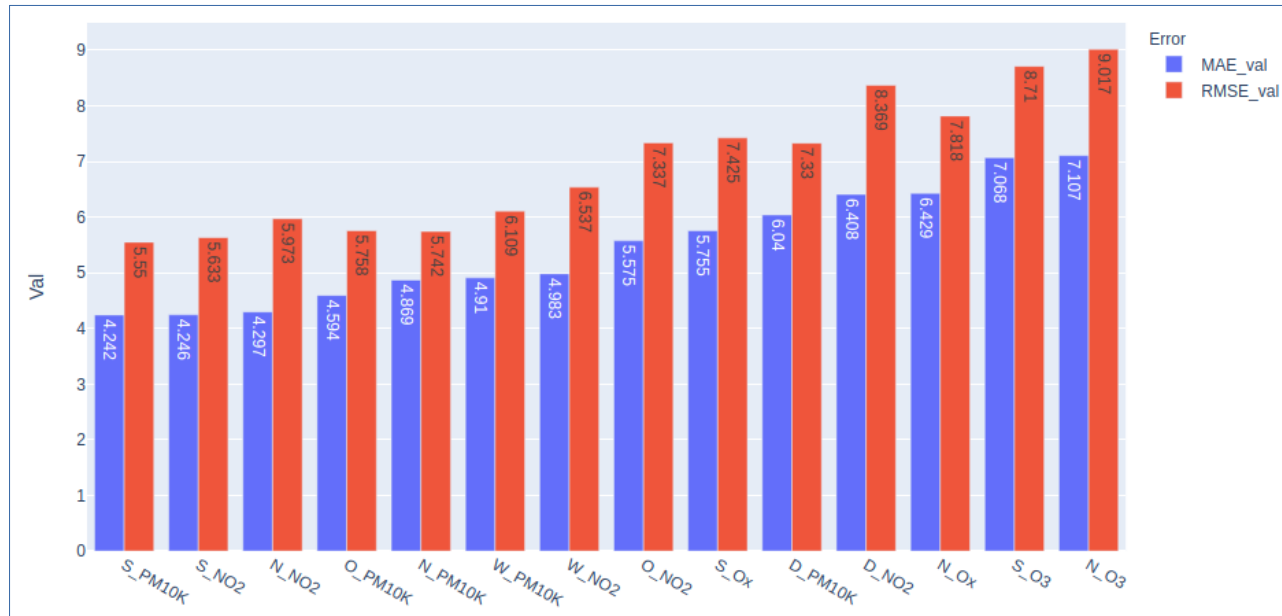


- Test data for NO2 is more simple than PM10
  - $\text{max\_test} \sim \text{max\_train}$



# Min

	RMSE	MAE	RMSE_val	MAE_val
D_NO2	RMSE_f4	MAE_f4	8.368847	6.407923
N_NO2	RMSE_new_base_add	MAE_base_add	5.973273	4.297227
O_NO2	RMSE_new_base_add	MAE_new_base_add	7.337199	5.575182
S_NO2	RMSE_f4	MAE_f4	5.633155	4.246237
W_NO2	RMSE_f4	MAE_f4	6.536686	4.982993
D_PM10K	RMSE_new_base_multi	MAE_new_base_multi	7.329732	6.039981
N_PM10K	RMSE_crossed_f5	MAE_crossed_f5	5.742252	4.869489
O_PM10K	RMSE_new_crossed_f	MAE_new_crossed_f	5.758059	4.593578
S_PM10K	RMSE_new_crossed_f	MAE_new_crossed_f	5.549522	4.242137
W_PM10K	RMSE_new_crossed_f	MAE_new_crossed_f	6.108925	4.909600
N_O3	RMSE_crossed_f5	MAE_features2_multi	9.016616	7.106509
S_O3	RMSE_crossed_f5	MAE_crossed_f5	8.710238	7.068301
N_Ox	RMSE_f3_cutoffs	MAE_f3_cutoffs	7.817738	6.428780
S_Ox	RMSE_tunning_params	MAE_new_crossed_f	7.425076	5.755069



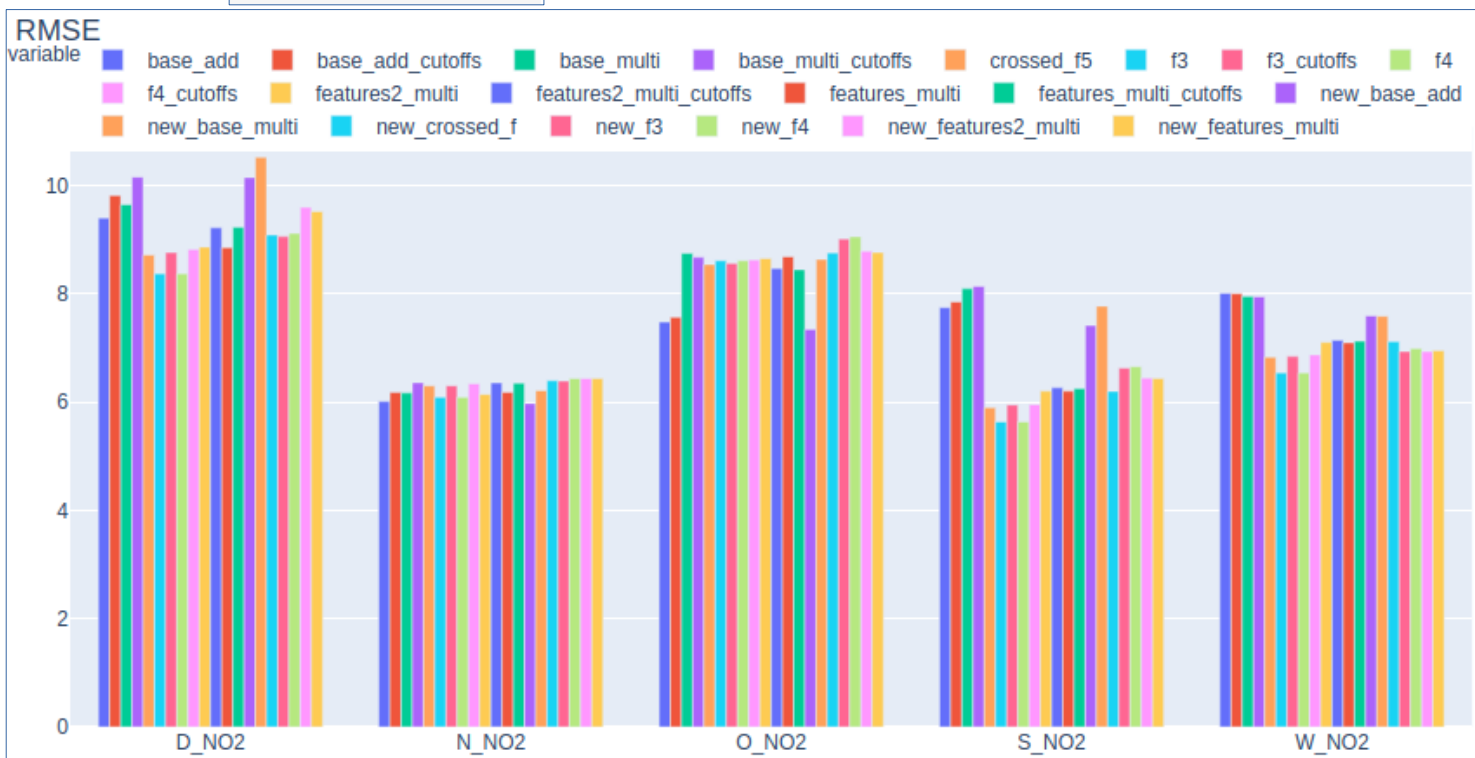
- Methods
  - 50%:50 for new and old data
  - *crossed\_features* dominate
    - S\_PM10
    - N\_PM10
    - O\_PM10
    - W\_PM10
  - *base\_add* fits well for N\_NO2, O\_NO2 and *base\_mul* for D\_PM10K
  -

- Values
  - S\_PM10K

# RMSE

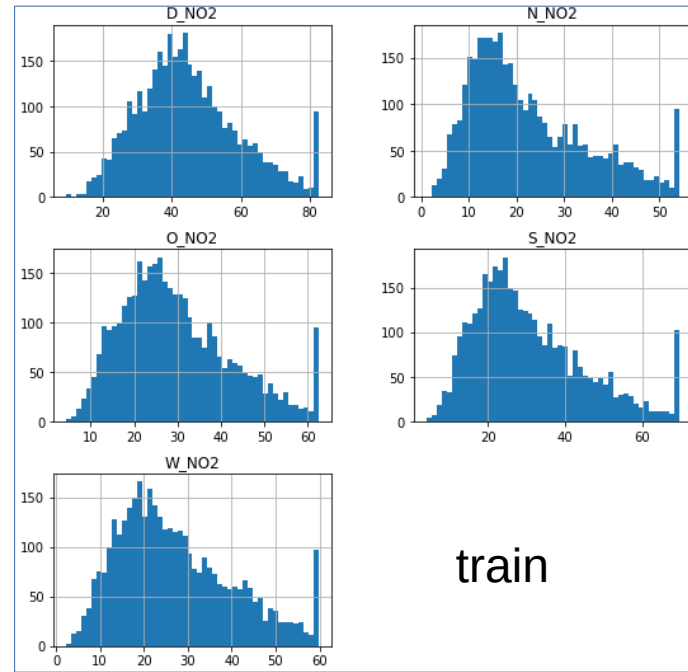
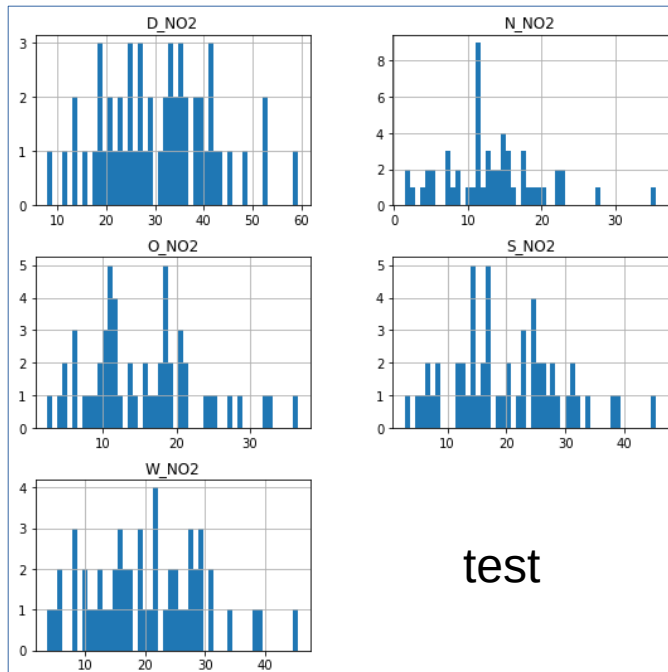
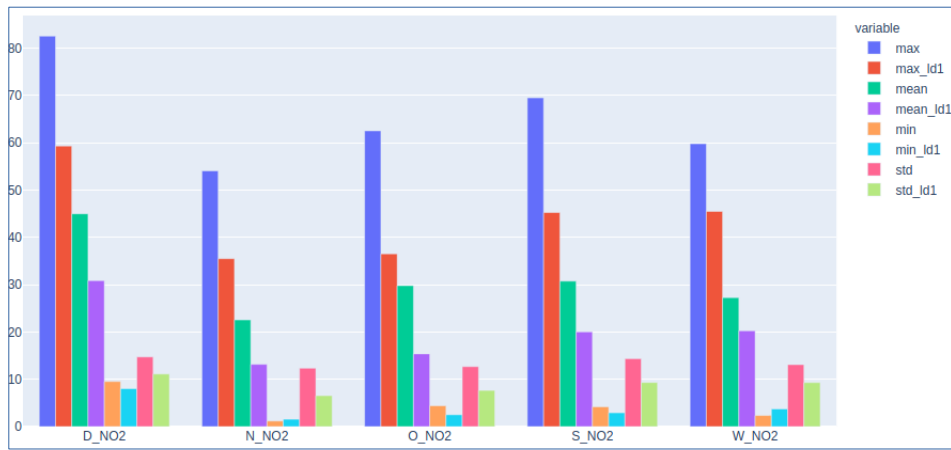
	0	1
D_NO2	8.369	f3
N_NO2	5.973	new_base_add
O_NO2	7.337	new_base_add
S_NO2	5.633	f3
W_NO2	6.537	f3

- N\_NO2
  - is already low error compared to other NO2
- N\_NO2 and O\_NO2: more features do not help
  - Best at new base additive
    - *tune the parameters of the algorithm/preprocess more data to get better results*
- D\_, S\_, W\_: best at f3 (=f4) from old data



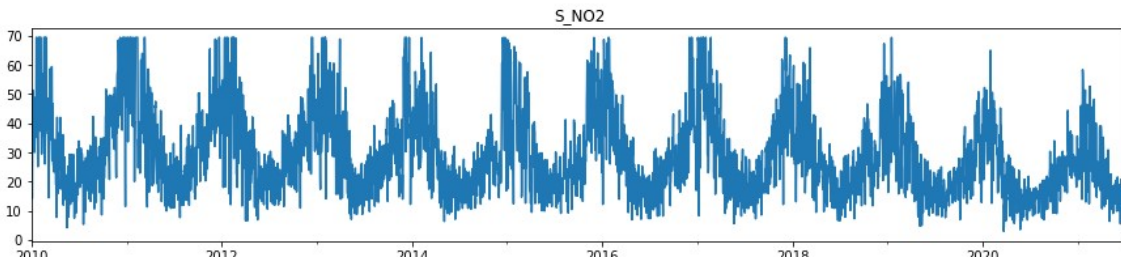
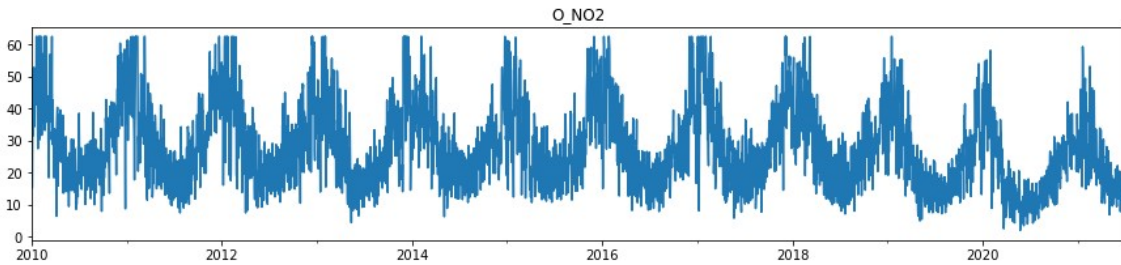
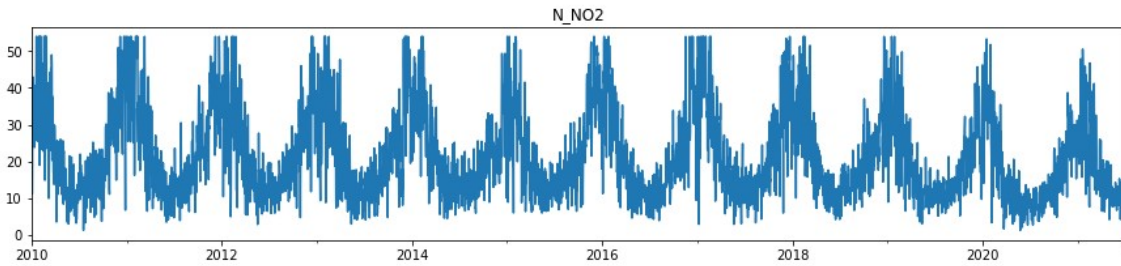
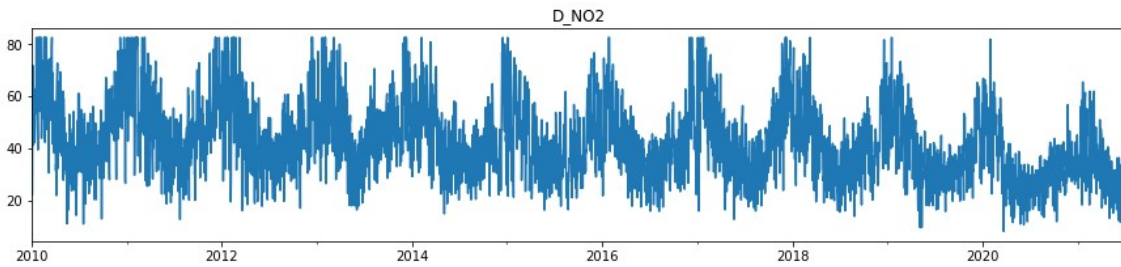
# Explanation

- N\_NO2 and S\_NO2 with lower error:
- D\_NO2 bigger error
- 



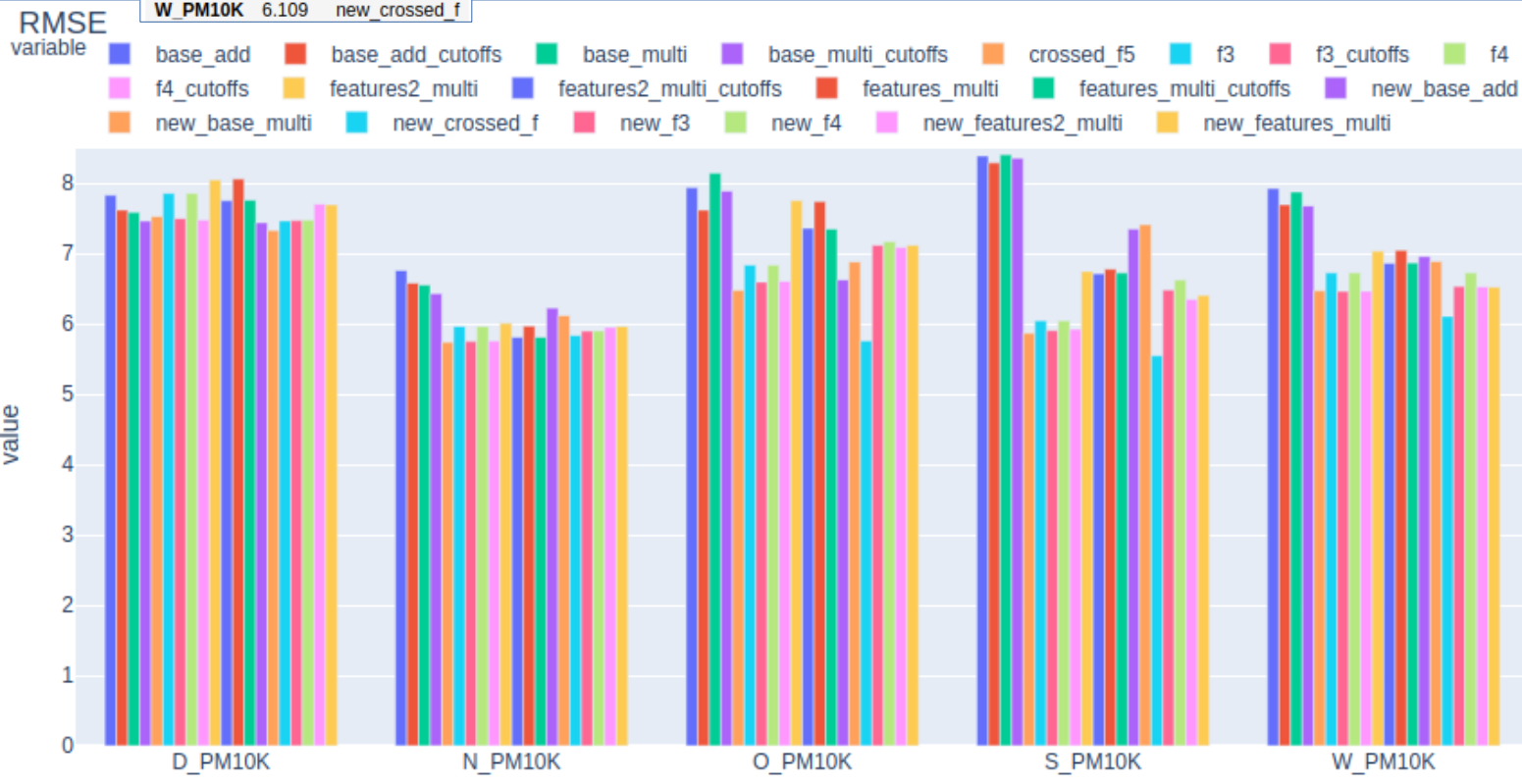


# Explanation



# RMSE

	0	1
D_PM10K	7.330	new_base_multi
N_PM10K	5.742	crossed_f5
O_PM10K	5.758	new_crossed_f
S_PM10K	5.550	new_crossed_f
W_PM10K	6.109	new_crossed_f



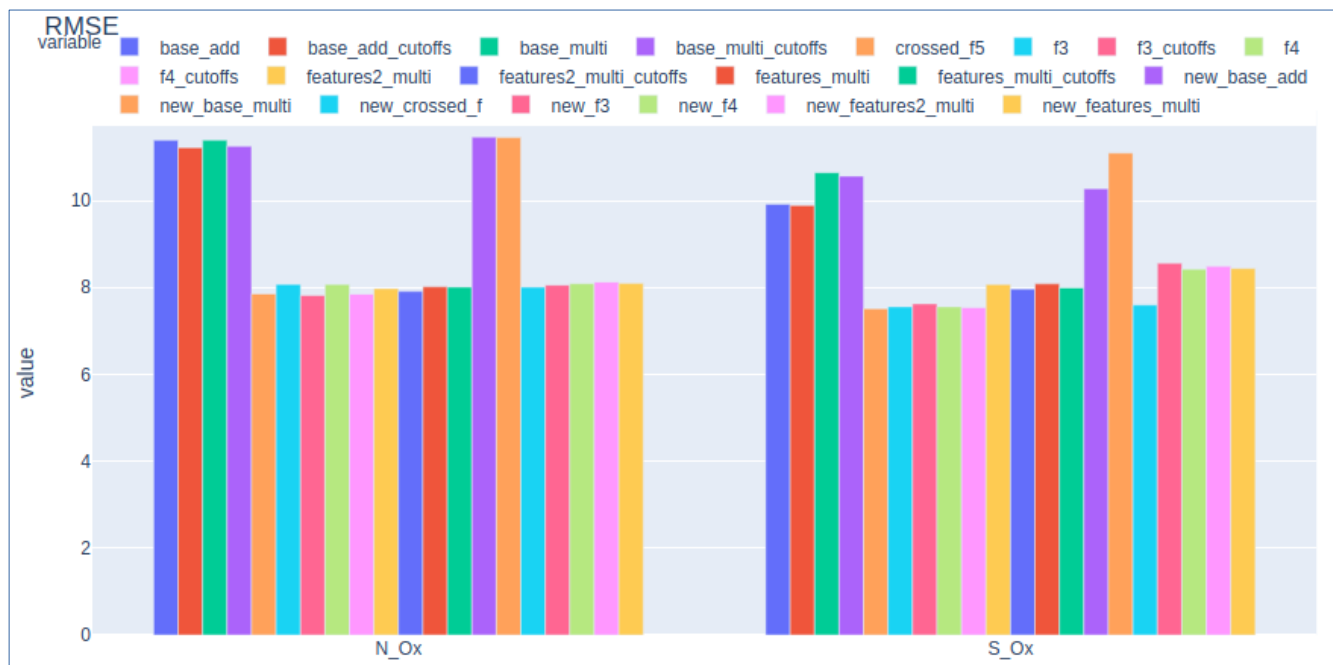
# RMSE

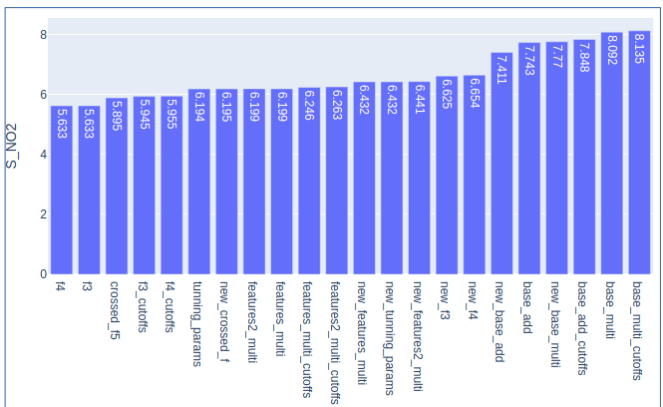
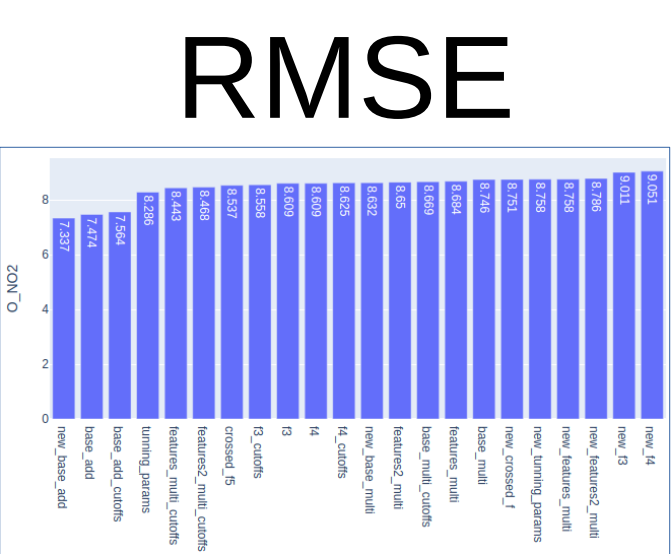
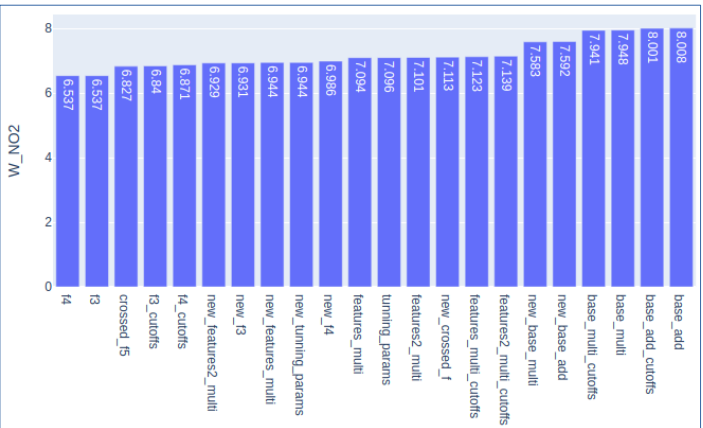
	0	1
N_O3	9.017	crossed_f5
S_O3	8.710	crossed_f5



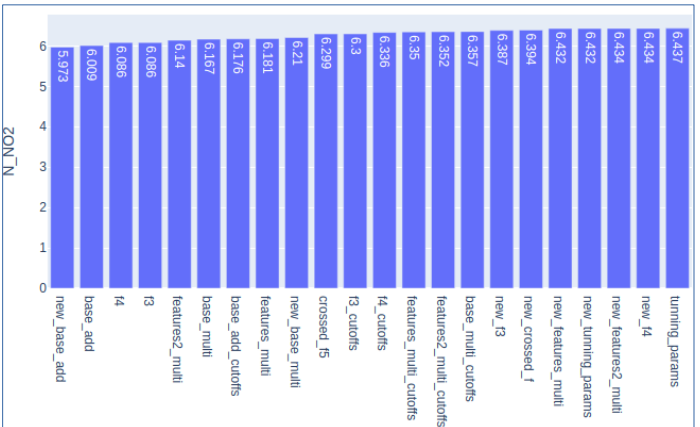
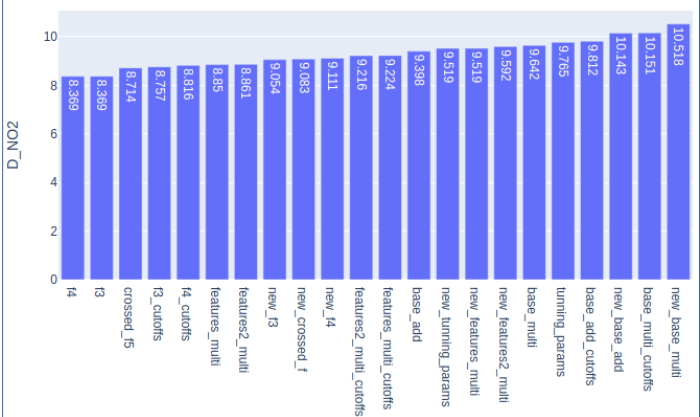
# RMSE

	0	1
<b>N_Ox</b>	7.818	f3_cutoffs
<b>S_Ox</b>	7.510	crossed_f5





# RMSE



# Next Steps

1. Cutoffs
2. Optimization for each area
3. Lockdown 2, 3?

Thank you!