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EMS, Madrid, 1-5 Oct 2007

Wind power predictability: comparative study of forecasts with MM5 and WRF for Portuguese transmission system operator

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Goal

- Wind Power Forecast for TSO (Transmission System Operator)
 - Dispatching decisions (hourly basis)
 - Load scheduling strategy (daily basis)
- Motivation
 - MM5 runs 4 times per day (00,06,12,18) with 72h lead time.
 - Best forecast?
 - Improvement of GFS 0.5° (40km) over 1° (80km)?
 - Improvement of WRF over MM5?



Presentation Outline

1. Short Term Forecast (intra-daily)

- NWP + observations
- NWP time lagged ensemble

2. Medium Term Forecasts (daily)

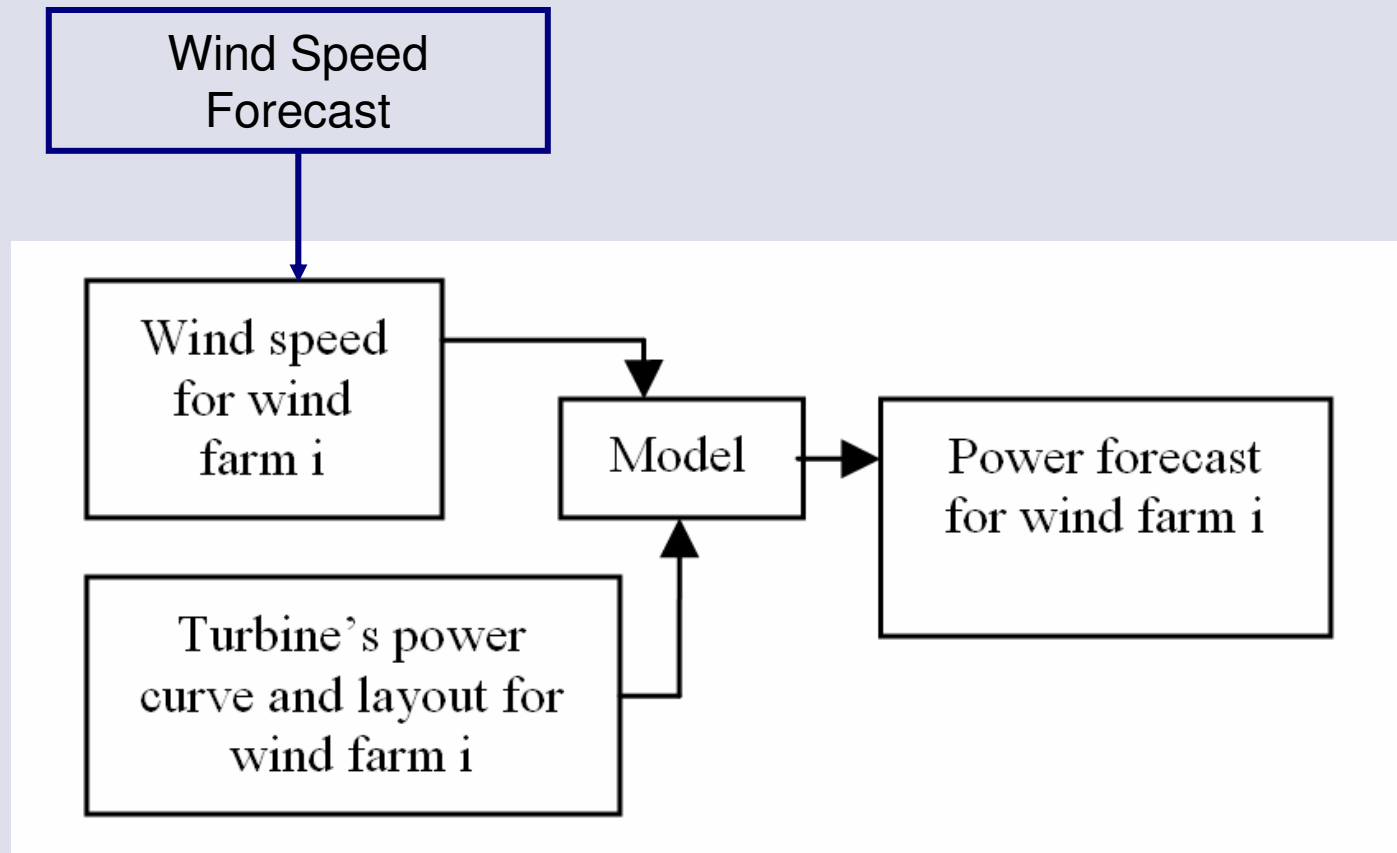
- NWP time lagged ensemble

3. Compare MM5 and WRF with GFS 1° and 0.5° resolution

Observations from online wind parks and weather stations, for 1st semester 2007.



Transmission System Operator (TSO)





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REN Power Forecast

- Portugal:
 - 1 939 MW installed by July 2007 (147 parks).
 - 700 MW (13 parks) being telemeasured
 - 3 345 MW already licensed
- Persistence:
 - To improve short time scales
 - To correct initial numerical forecasts
- Plan Outages (wind farms, lines, ...)

<http://www.ren.pt/sections/exploracao/dpe/default.asp>



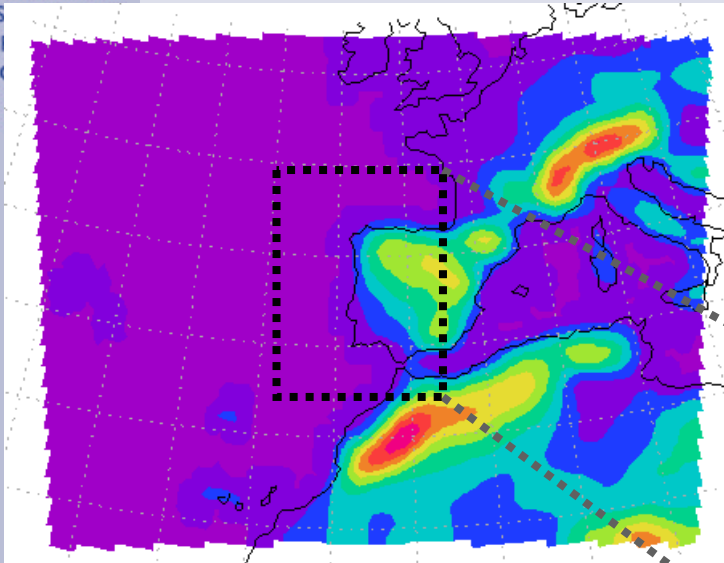


IST-MM5

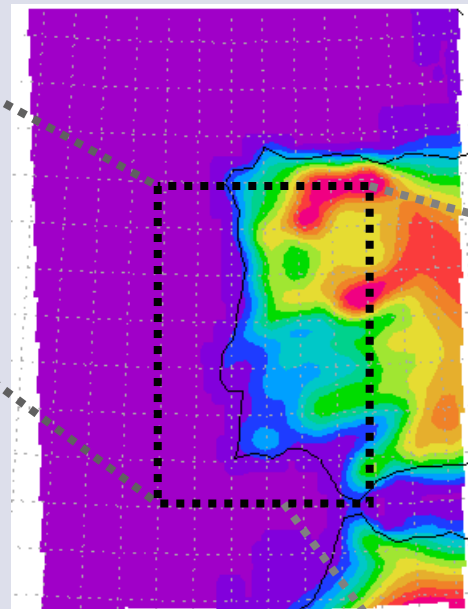
<http://meteo.ist.utl.pt>

Operacional since 2000.

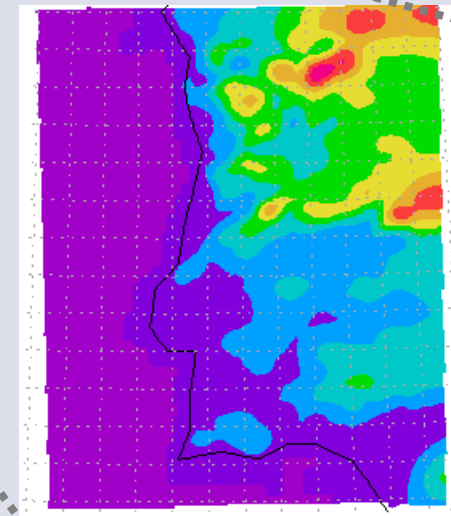
Online since 2001.



40x50
dx = 81 km



55x40
dx = 27 km



82x55
dx = 9 km



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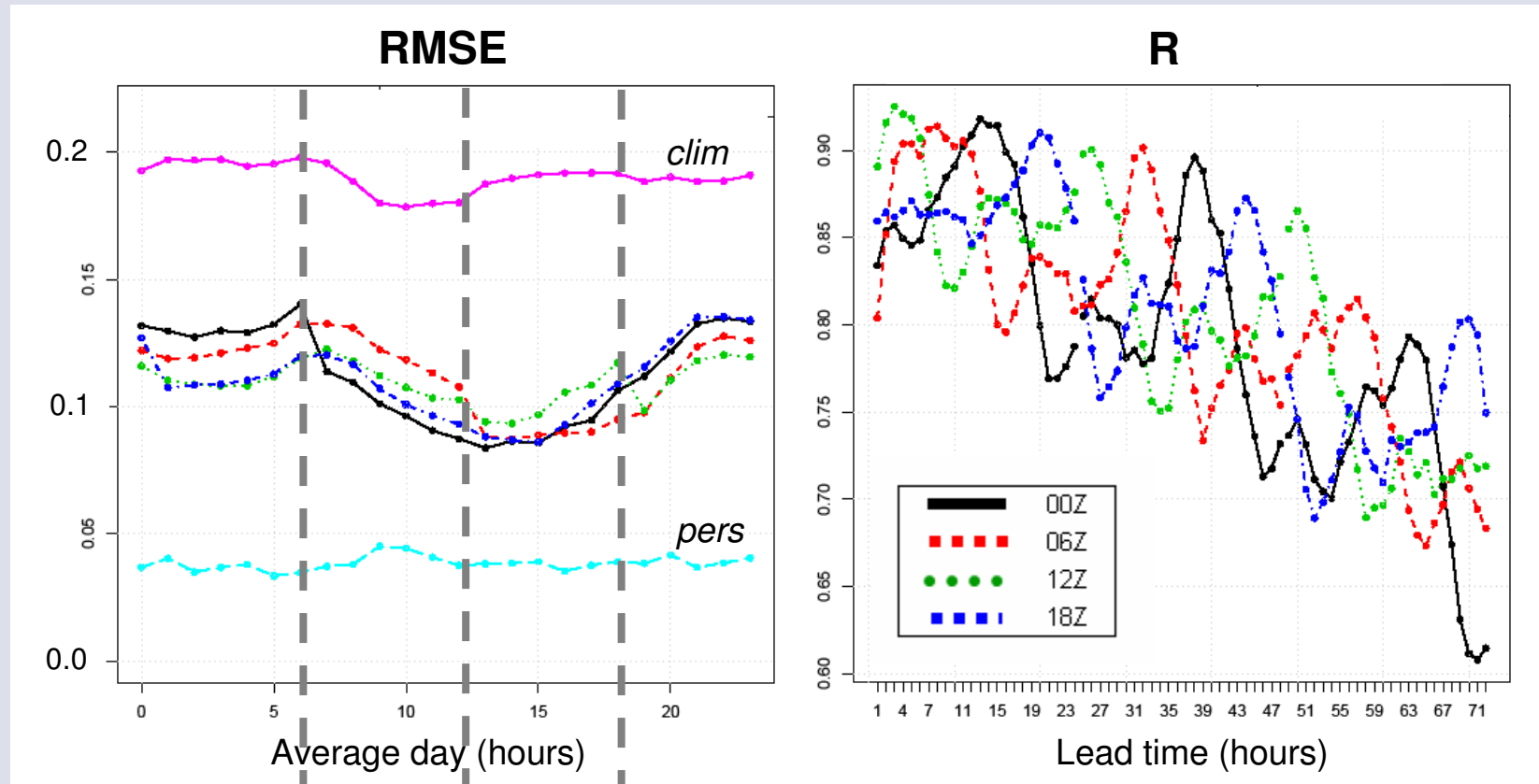
IST-MM5

- GFS initial and boundary conditions 1^o resolution
- USGS topography
- 27 vertical levels
- Forecasts 72h, 4 times per day (00,06,12,18Z)



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Compare MM5 (1^o GFS) 4x day⁻¹



Smallest RMSE: 18Z 00Z 06Z 12Z
6h 12h 18h

Within 24h, the best performance is achieved in **lead times from 6-12 h**. RMSE increases with lead time, mainly due **phase errors** (↓ R)



Short Term Forecasts

- **S1** : Most recent available forecast (minimum lead time). Because each simulation takes about 5 hours to complete, we use lead times from 7 to 12 hours;

$$\hat{Y}_{t+i} = F_{t+i},$$

	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
Run	18Z						00Z						06Z						12Z					
HZ	7	8	9	10	11	12	7	8	9	10	11	12	7	8	9	10	11	12	7	8	9	10	11	12

- **S2** : Most recent available forecast combined with observations at $t=0,6,12,18$ h, with linear weights.

$$\hat{Y}_{t+i} = (1 - \alpha_i)F_{t+i} + \alpha_i Y_{t+6},$$

$$t \in \{0, 6, 12, 18\}$$

$$i \in \{7, \dots, 12\}$$

$$\alpha \in \{0.9, \dots, 0.1\}$$

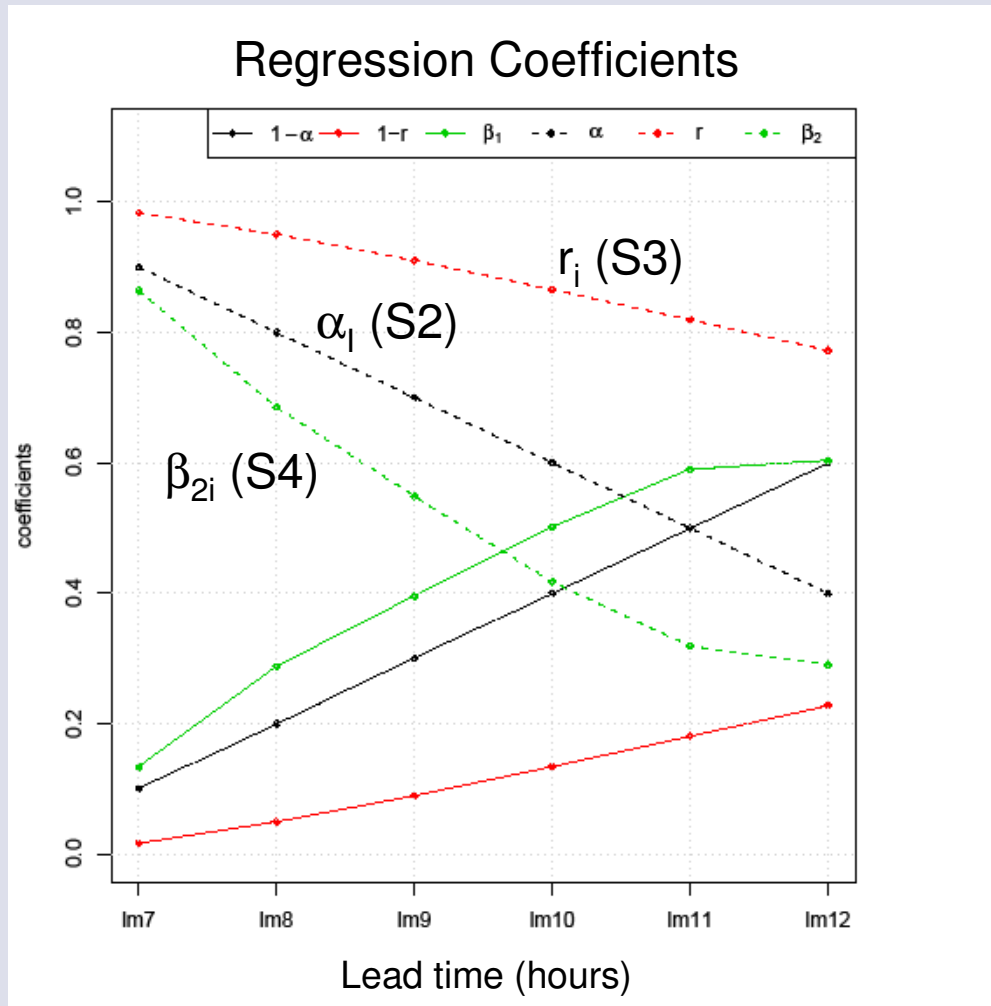
- **S3** : Most recent available forecast combined with observations at $t=0,6,12,18$ h, with weights from autocorrelation function r .

$$\hat{Y}_{t+i} = (1 - r_{i-5})F_{t+i} + r_{i-5}Y_{t+6},$$

- **S4** : Most recent available forecast combined with observations at $t=0,6,12,18$ h, with weights from LS regression

$$Y_{t+i} = \beta_{0,i} + \beta_{1,i}F_{t+i} + \beta_{2,i}Y_t$$

Short Term Forecasts



Coefficients in S2 (**black**) and S3 (**red**) forecast give more weight to the past observed values and less weight to NWP model than S4 weights (**green**).

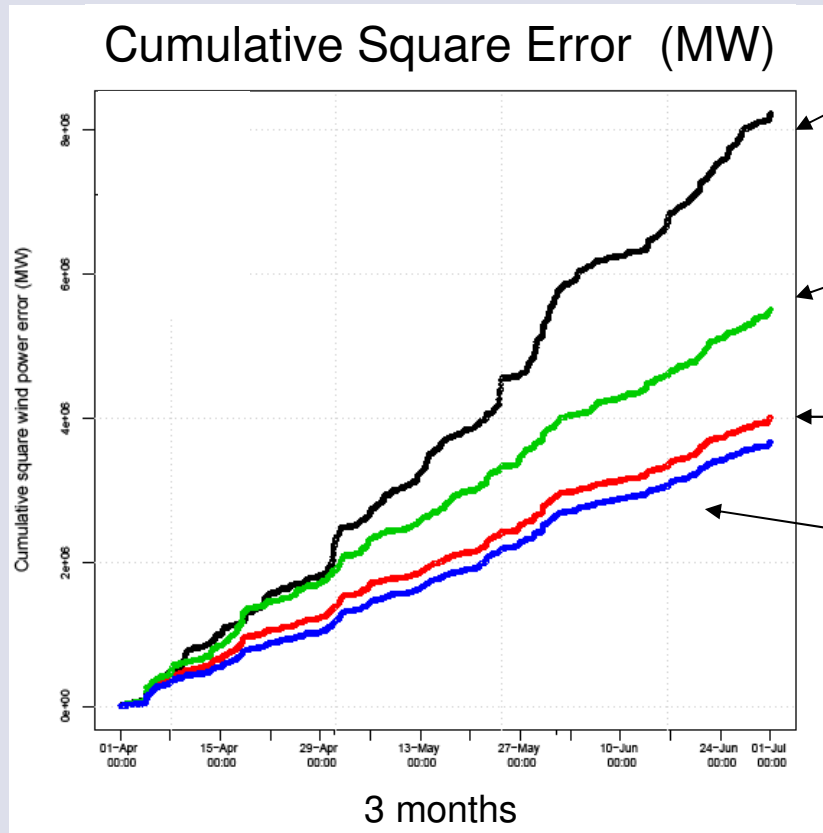
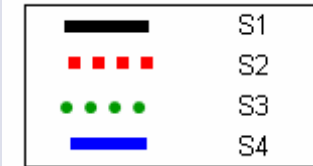
$$\mathbf{S2} \quad \hat{Y}_{t+i} = (1 - \alpha_i)F_{t+i} + \alpha_i Y_{t+6},$$

$$\mathbf{S3} \quad \hat{Y}_{t+i} = (1 - r_{i-5})F_{t+i} + r_{i-5}Y_{t+6},$$

$$\mathbf{S4} \quad Y_{t+i} = \beta_{0,i} + \beta_{1,i}F_{t+i} + \beta_{2,i}Y_t$$



Short Term Forecasts



S1 (only NWP forecast)

S3 (weights from ACF)

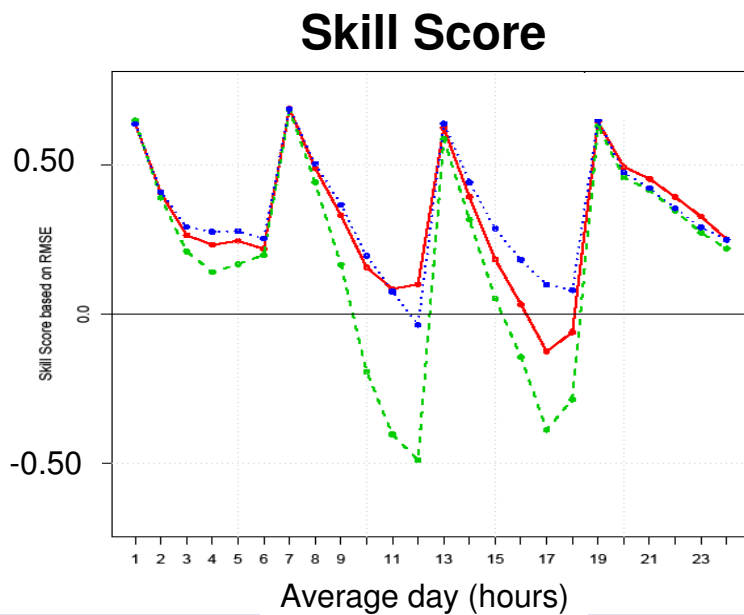
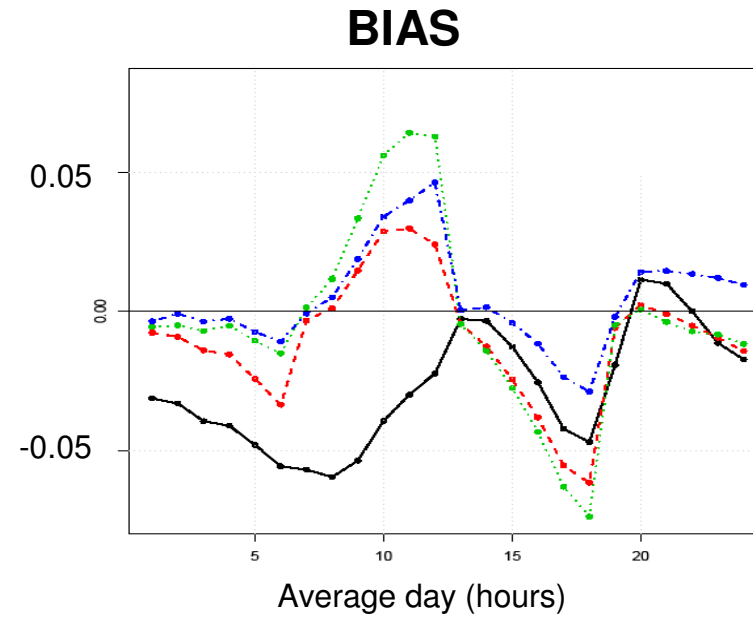
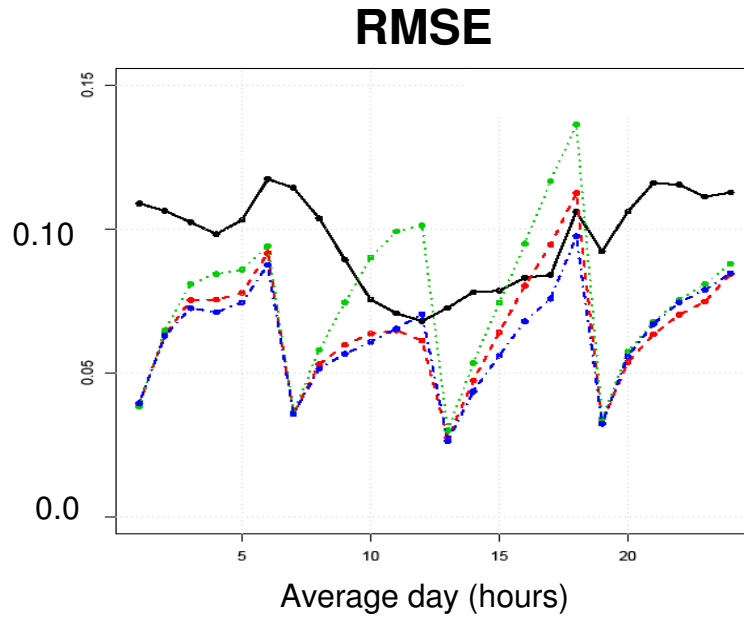
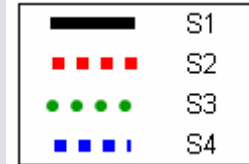
S2 (linear weights)

S4 (LS regression weights)



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Short Term Forecasts

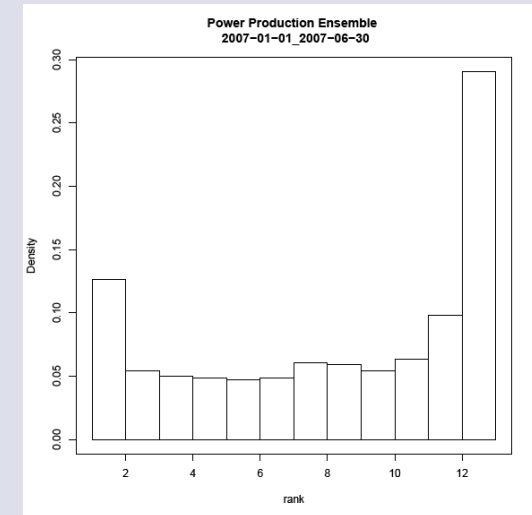


	S1	S2	S3	S4
RMSE	9.8%	6.8%	8.0%	6.5%
SS	-	30%	18%	33%



Time Lagged Ensemble

- There are 11 available forecasts for each hours and each site.
- **Can we improve the most recent forecast (S1)?**
 - Mean Ensemble (equal weights)
 - Stepwise Multilinear Regression (step.reg)
 - Principal Components Regression (PCR)
 - Partial Least Squares Regression (PLSR)



**Members are not
dispersive and
tend to
underestimate
observations**



Day 0 – available forecasts

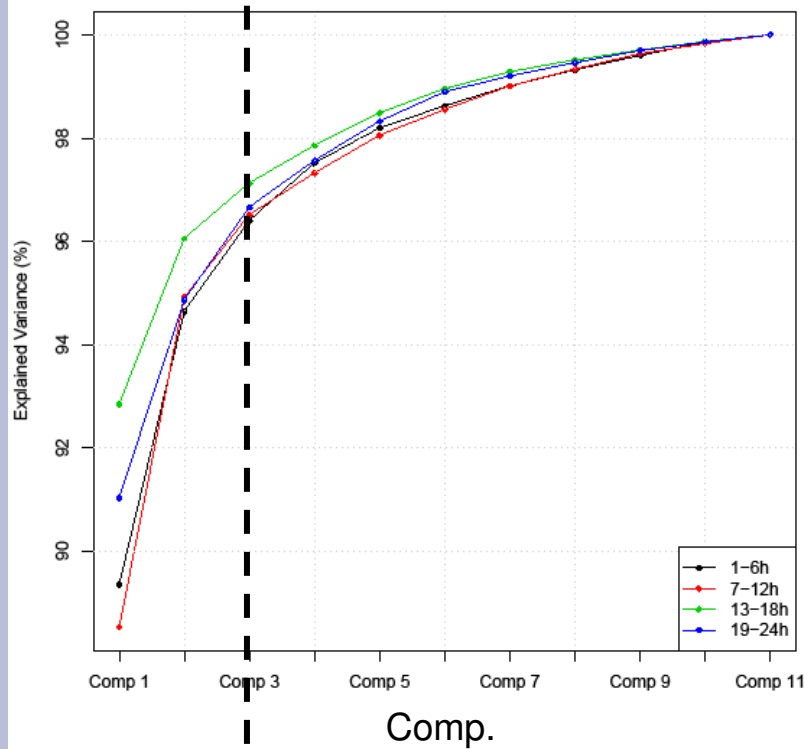
Day 0																								
Ensemble 00 (11 members)							Ensemble 06 (11 members)						Ensemble 12 (11 members)						Ensemble 18 (11 members)					
1	2	3	4	5	6		7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
06d2	06d2	06d2	06d2	06d2	06d2																			
67	68	69	70	71	72																			
12d2	12d2	12d2	12d2	12d2	12d2		12d2	12d2	12d2	12d2	12d2	12d2												
61	62	63	64	65	66		67	68	69	70	71	72												
18d2	18d2	18d2	18d2	18d2	18d2		18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2						
55	56	57	58	59	60		61	62	63	64	65	66	67	68	69	70	71	72						
00d2	00d2	00d2	00d2	00d2	00d2		00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2
49	50	51	52	53	54		55	56	57	58	59	60	61	62	63	64	65	66	67	68	69	70	71	72
06d1	06d1	06d1	06d1	06d1	06d1		06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2	06d2
43	44	45	46	47	48		49	50	51	52	53	54	55	56	57	58	59	60	61	62	63	64	65	66
12d1	12d1	12d1	12d1	12d1	12d1		12d1	12d1	12d1	12d1	12d1	12d1	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2
37	38	39	40	41	42		43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59	60
18d1	18d1	18d1	18d1	18d1	18d1		18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d1	18d2	18d2	18d2	18d2	18d2	18d2
31	32	33	34	35	36		37	38	39	40	41	42	43	44	45	46	47	48	49	50	51	52	53	54
00d1	00d1	00d1	00d1	00d1	00d1		00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1	00d1
25	26	27	28	29	30		31	32	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47	48
06d0	06d0	06d0	06d0	06d0	06d0		06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1	06d1
19	20	21	22	23	24		25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	41	42
12d0	12d0	12d0	12d0	12d0	12d0		12d0	12d0	12d0	12d0	12d0	12d0	12d1	12d1	12d1	12d1	12d1	12d1	12d1	12d1	12d1	12d1	12d1	12d1
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18d0	18d0	18d0	18d0	18d0	18d0		18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d0	18d1	18d1	18d1	18d1	18d1	18d1
7	8	9	10	11	12		13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30
00d0	00d0	00d0	00d0	00d0	00d0		00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0	00d0
4	5	6	7	8	9		10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	
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							4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	
													12d0	12d0	12d0	12d0	12d0	12d0	12d0	12d0	12d0	12d0	12d0	12d0
													4	5	6	7	8	9	10	11	12	13	14	
																		18d0	18d0	18d0	18d0	18d0	18d0	
																		4	5	6	7	8	9	



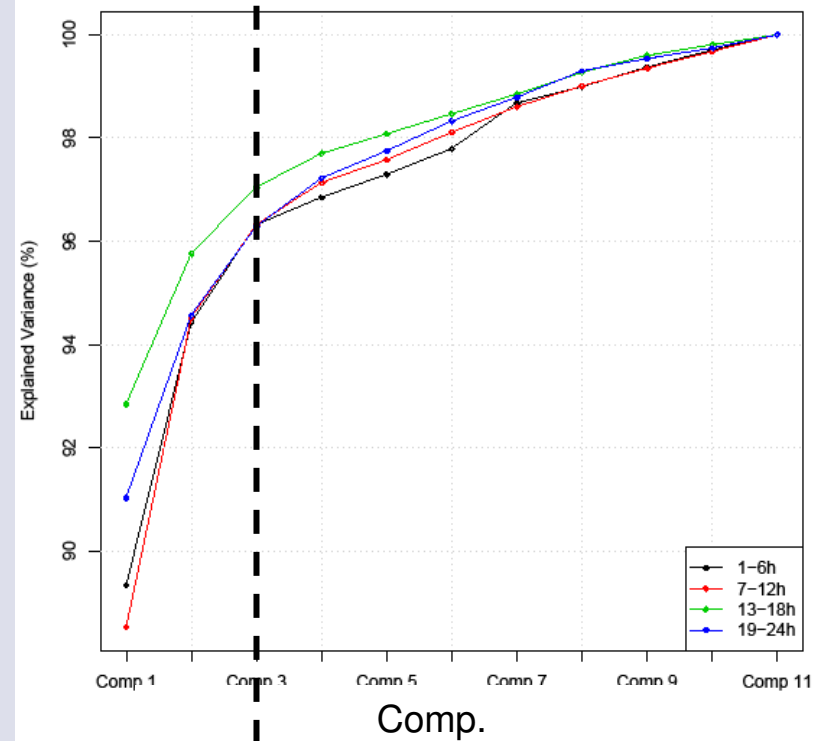
Components Regression



PCA Explained Variance



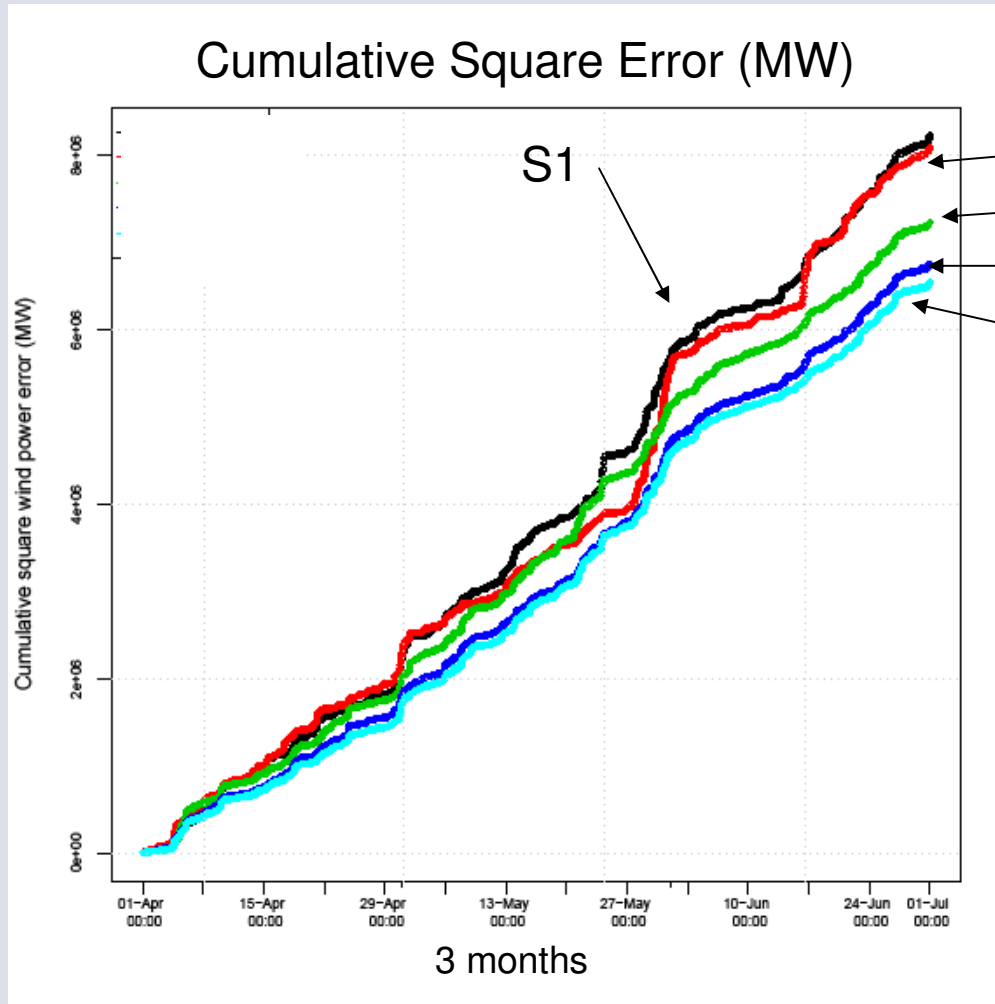
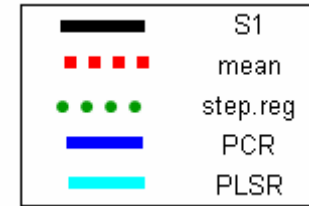
PLSR Explained Variance



3 components explain > 96% of variance



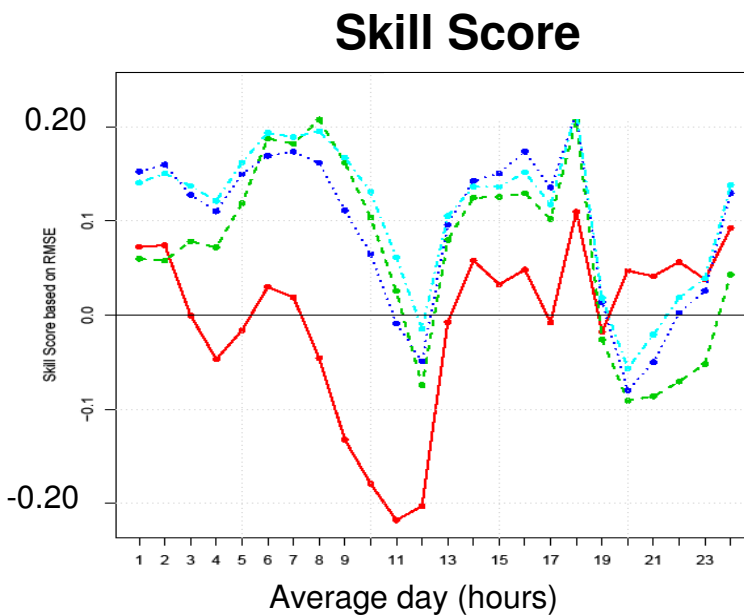
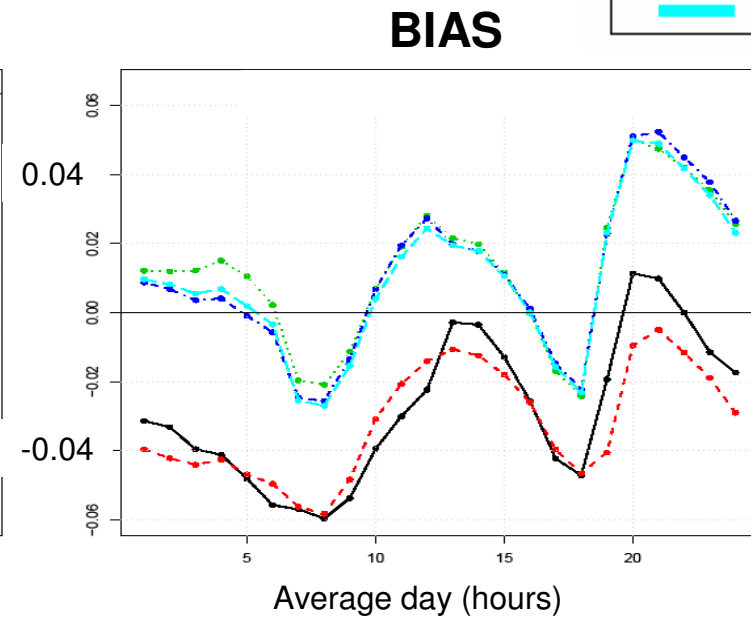
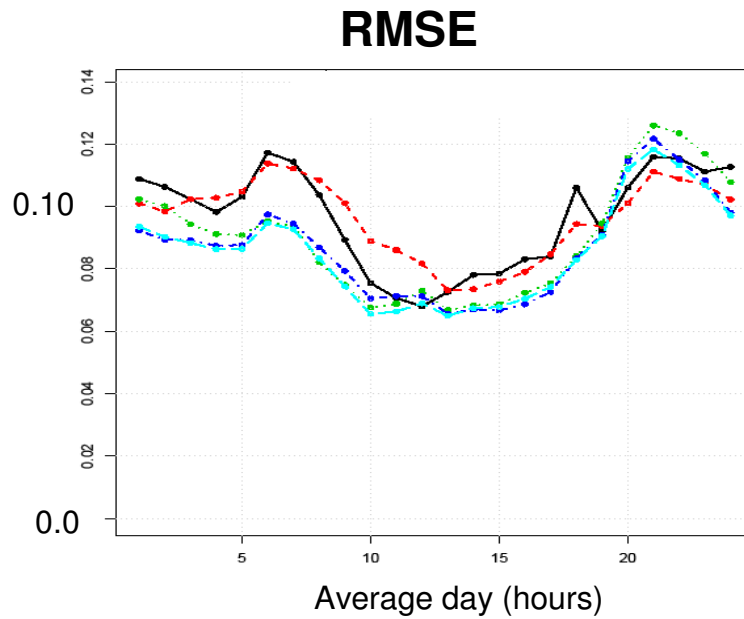
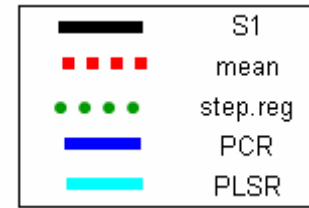
Time Lagged Ensemble





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Time Lagged Ensemble



	RMSE	SS
S1	9.8%	-
mean	9.7%	1%
step.reg	9.2%	6%
PCR	8.9%	9%
PLSR	8.7%	11%



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Medium Range Forecasts

- Forecasts for the next 2 days
- Need to be ready at 6 a.m. (day 0)



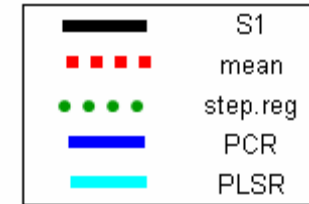
Day 1 – available forecasts

Day 1																							
Ensemble 00 (8 members)						Ensemble 06 (7 members)						Ensemble 12 (6 members)						Ensemble 18 (5 members)					
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
06d2	06d2	06d2	06d2	06d2	06d2																		
67	68	69	70	71	72																		
12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2	12d2												
61	62	63	64	65	66	67	68	69	70	71	72												
18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2	18d2						
55	56	57	58	59	60	61	62	63	64	65	66	67	68	69	70	71	72						
00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2	00d2
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43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59	60	61	62	63	64	65	66
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37	38	39	40	41	42	43	44	45	46	47	48	49	50	51	52	53	54	55	56	57	58	59	60
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31	32	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47	48	49	50	51	52	53	54
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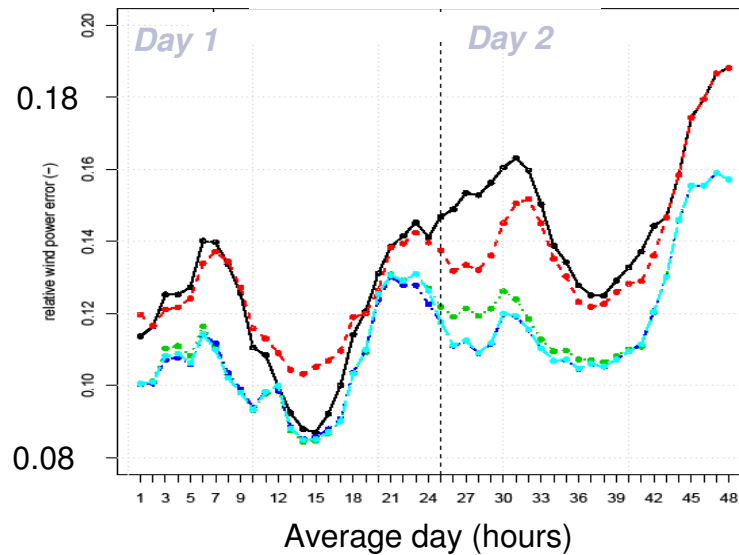


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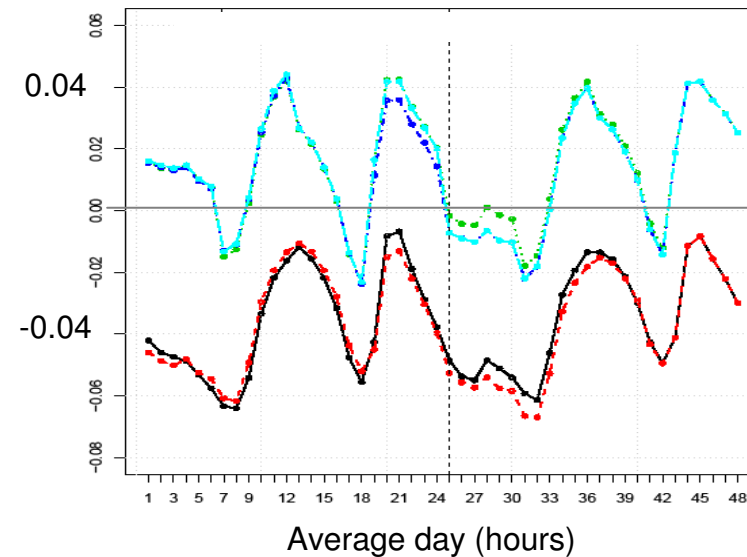
Medium Range Forecasts



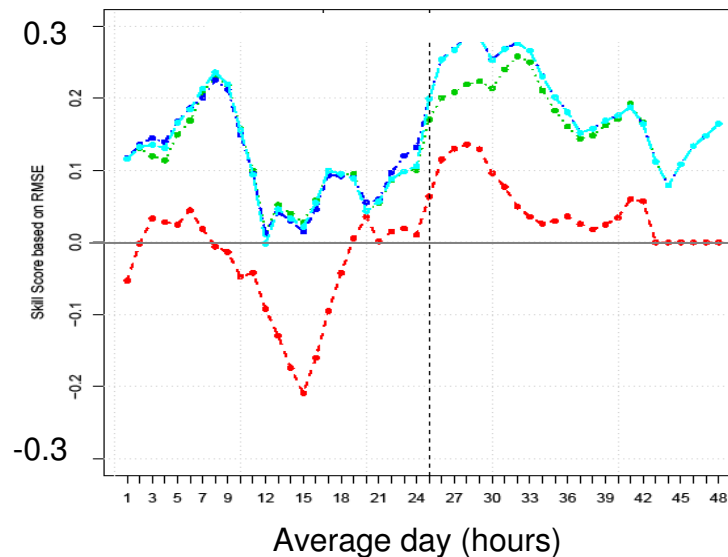
RMSE



BIAS



Skill Score



M1 = 00Z with HZ = 25 - 72 h

Day 1	M1	mean	step.reg	pcr3	plsr3
RMSE	12.0%	12.3%	10.7%	10.6%	10.6%
SS	-	-2%	11%	12%	12%

Day 2	M1	mean	step.reg	pcr1	plsr1
RMSE	15.2%	14.5%	12.5%	12.3%	12.3%
SS	-	5%	18%	19%	19%



Compare GFS Resolution

- Compare 00Z simulations
- $6 < \text{HZ} \leq 30$ hours (1st 6-hours are not forecast)
- BC updated in 3h intervals

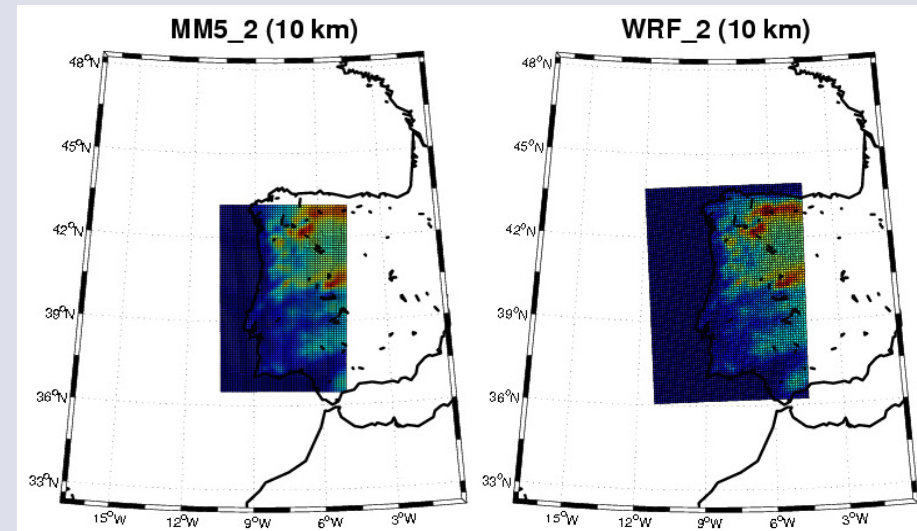
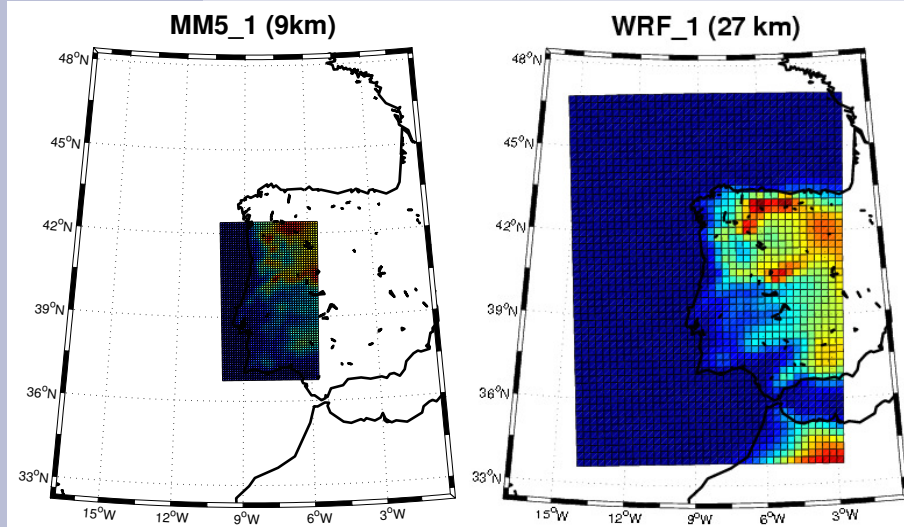
- MM5₁ & WRF₁ (GFS 1^o resolution)
- MM5₂ & WFR₂ (GFS 0.5^o resolution)



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GFS Resolution – Domains

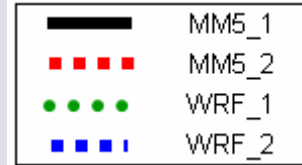
	D1	D2	D3
MM5_1	81 km (31x31)	27 km (43x31)	9 km (70x43)
WRF_1	81 km (50x40)	27 km (55x40)	-
MM5_2	-	27 km (55x40)	9 km (82x25)
WRF_2	-	30 km (81x81)	10 km (85x61)



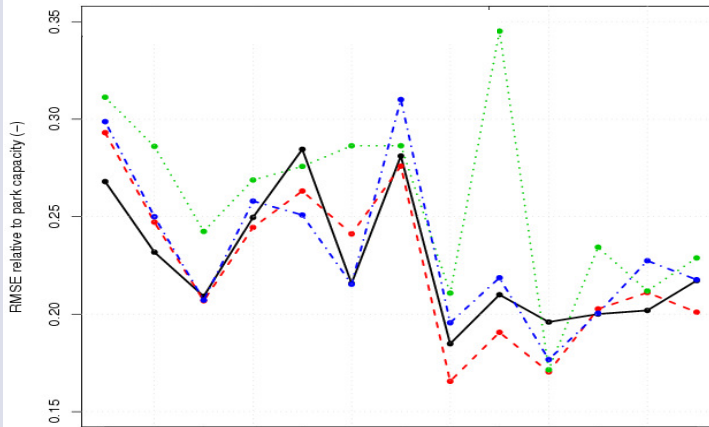


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GFS Resolution – Parks

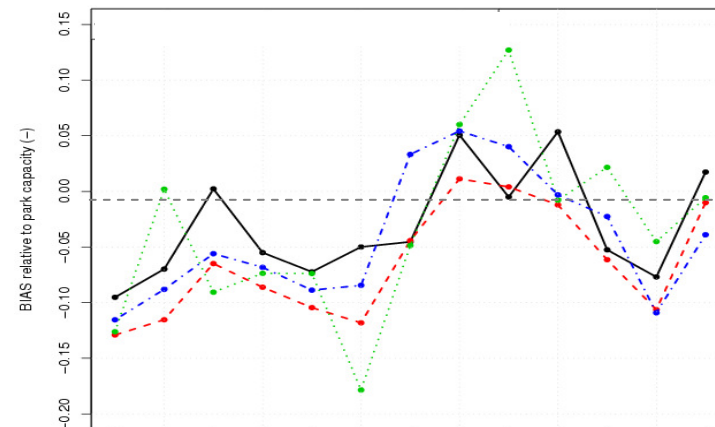


RMSE relative to each park



13 wind parks

BIAS relative to each park



13 wind parks

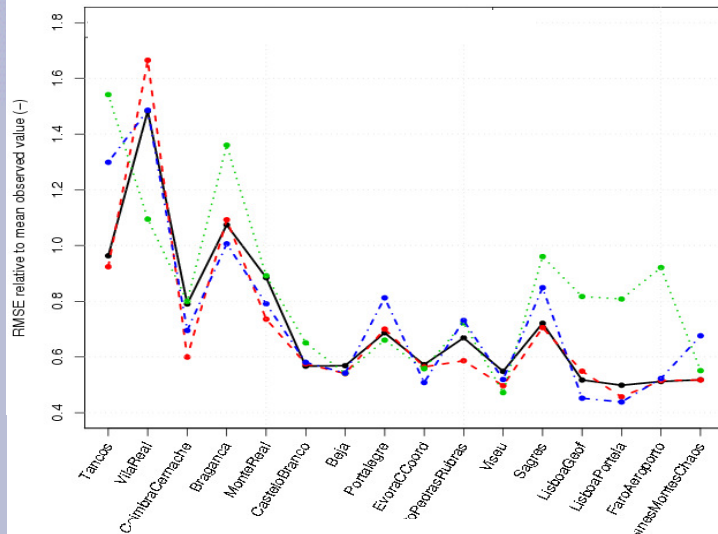
Parks are ordered by power capacity

	RMSE (-)	SS
MM5_1	2.41%	-
MM5_2	2.38%	2%
WRF_1	2.65%	-10%
WRF_2	2.49%	-3%

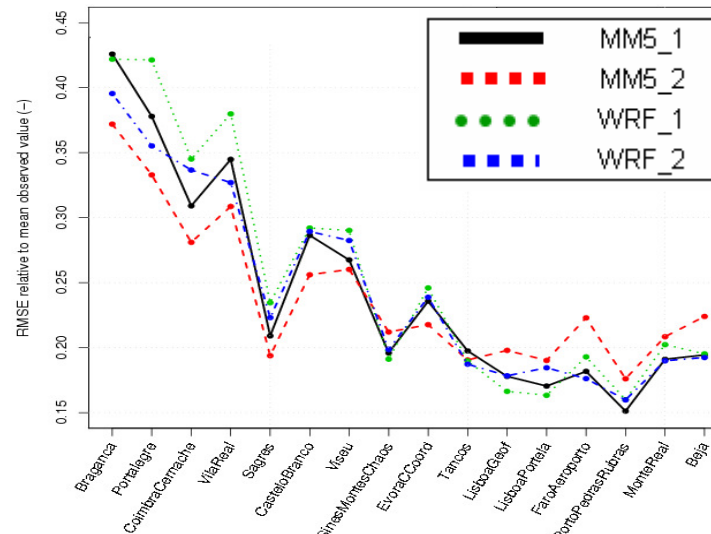


GFS Resolution – Stations (RMSE)

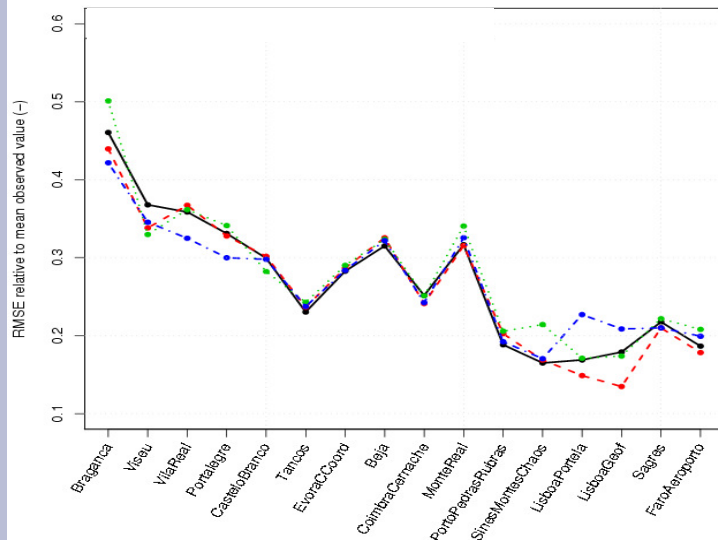
Wind speed at 10m



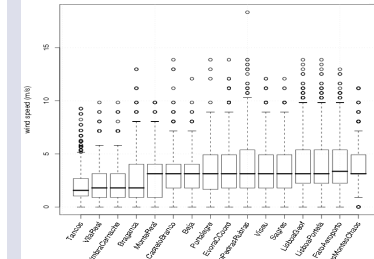
RH at 2m



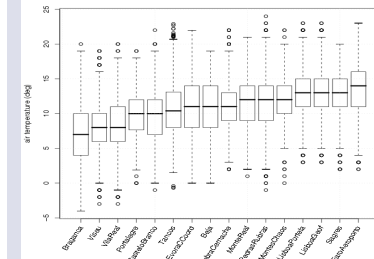
Air temperature at 2m



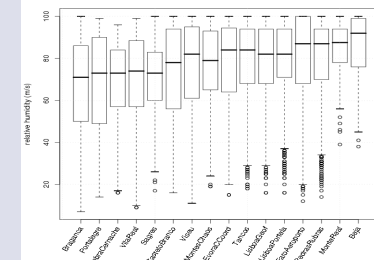
Wind speed at 10m



Air temperature 2m



Rel Hum. at 2m



Stations are ordered by mean observed value

RMSE	MM5_1	MM5_2	WRF_1	WRF_2
WS (m/s)	2.22	2.15	2.60	2.28
T (deg)	2.78	2.71	2.88	2.81
HR (%)	18.27	18.07	19.05	18.28

SS	MM5_1	MM5_2	WRF_1	WRF_2
WS (m/s)	-	3%	-17%	-3%
T (deg)	-	2%	-4%	-1%
HR (%)	-	1%	-4%	0%

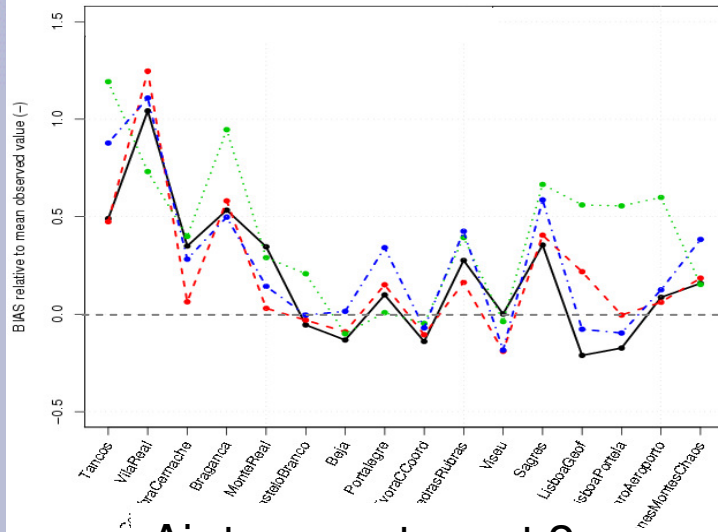
Data set from Wunderground for 1st trimester 2007



GFS Resolution – Stations (**BIAS**)

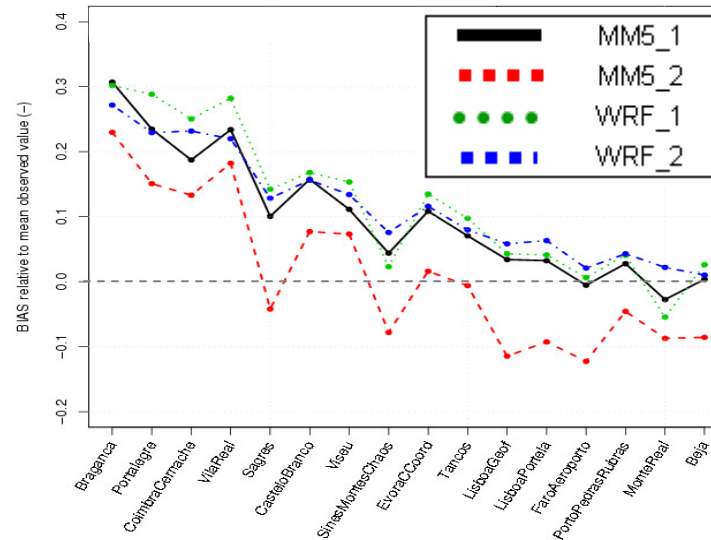
Wind speed at 10m

ordered by Mean Observed Value

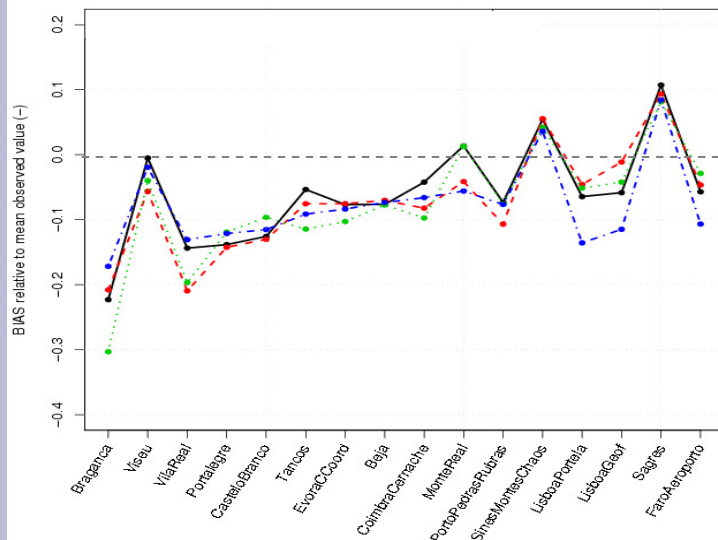


RH at 2m

ordered by Mean Observed Value



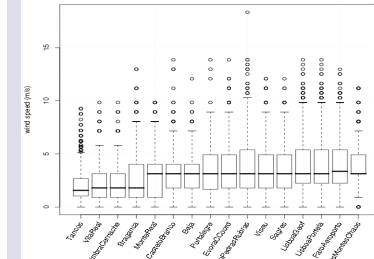
Air temperature at 2m



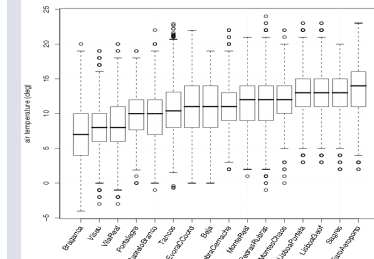
Stations are ordered by mean observed value

- MM5_2 does not underestimate RH as much as the other models
- Higher amplitude errors for high mean temperature sites, which are in this case the farthest away from the Sea.

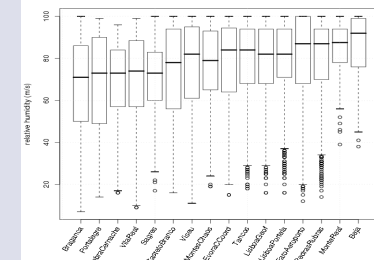
Wind speed at 10m



Air temperature 2m



Rel Hum. at 2m



Data set from Wunderground for 1st trimester 2007



Conclusions

	RMSE ref	RMSE best	SS
Short Term with data	10%	7%	33%
Short Term without data	10%	9%	11%
Medium Term Day 1	12%	11%	12%
Medium Term Day 2	15%	12%	19%

- Persistence is good to forecast wind power up to 6h (33%), but can go against diurnal cycle if not updated regularly.
- When there is no online data, improved forecast can be obtained with regressions from the time lagged ensemble (11%, 12% and 19% for days 0,1,2), but these regressions do not improve phase errors.
- The current configuration (MM5 with 1° GFS resolution) is:
 - Less skilled than MM5 0.5° GFS (~2%)
 - More skilled than WRF 0.5° GFS (~2%).



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Thank you!

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Error Decomposition

$$rmse^2 = bias^2 + sde^2 = bias^2 + sdbias^2 + disp^2$$

$$\varepsilon = x_{prd} - x_{obs}$$

Error

$$rmse = \sqrt{\varepsilon^2}$$

Root Mean Square Error

$$bias = \bar{\varepsilon}$$

Bias

$$sde = \sigma(\varepsilon)$$

Standard Deviation of Error

$$sdbias = \sigma(x_{prd}) - \sigma(x_{obs})$$

Variability Error

$$disp = \sqrt{2\sigma(x_{prd})\sigma(x_{obs})(1-r)}$$

Dispersion (phase error)

Lange M. (2005). *On the Uncertainty of Wind Power Predictions — Analysis of the Forecast Accuracy and Statistical Distribution of Errors*. Journal of Solar Energy Engineering. Vol. 127:177-184.