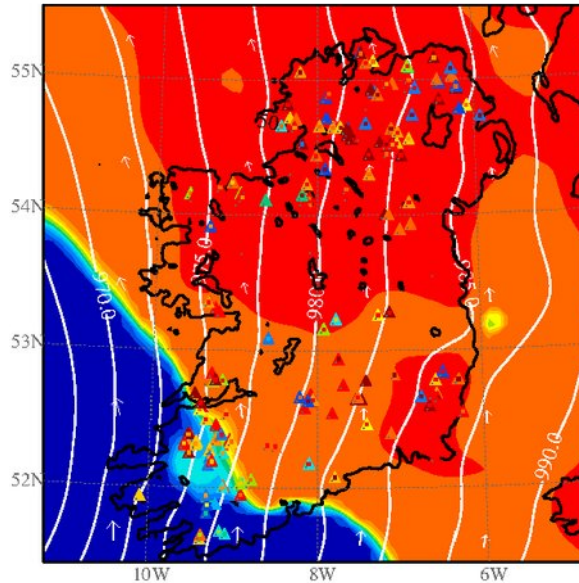


# Minute-scale Forecasting Workshop

*On the benefits and challenges of high-resolution ensemble forecasts for minute-scale ramping applications*



Risø Campus, Denmark 10-11 April 2024  
Dr. Corinna Möhrlen and Jess U. Jørgensen



# About WEPROG

Check out our free weather WebAPP @ <https://weather.weprog.com>

First commercial established Windpower Forecast Vendor in Europe in 2003!

WEPROG's name is an abbreviation of **W**eather & **E**nergy **PROG**nozes.

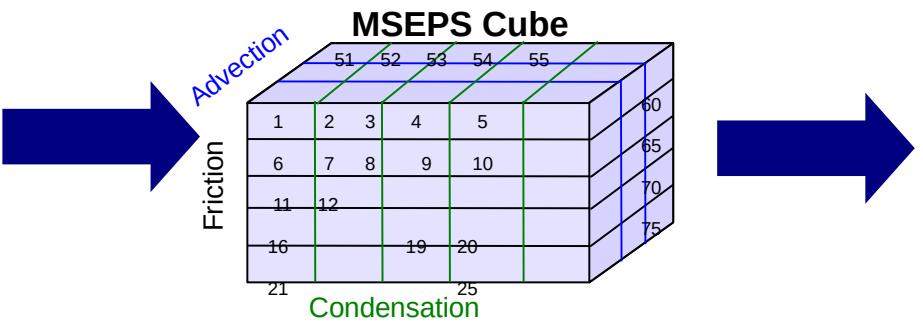
We provide **real-time ensemble forecasts** for a sustainable energy system

WEPROG operates a **continuous (24/7/365) forecast production.**

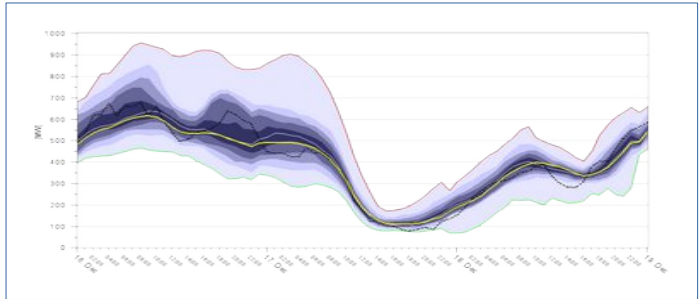
**Real-time products** are available in any region and all **continents.**



75 independent weather forecasts



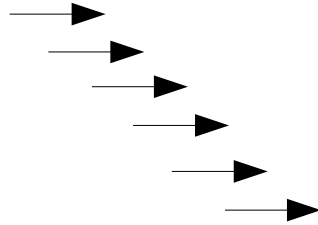
Probabilistic weather, wind & solar power forecasts



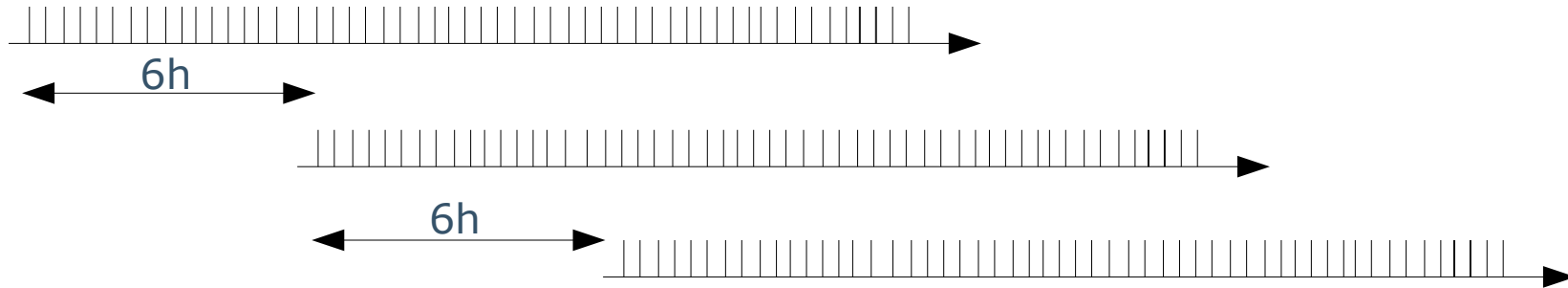


# Minute-Scale Forecasting Applications

**(1) Data-driven with high weight on persistence with a look ahead less than 2 hours with focus on precision frequently generated**



**(2) Weather forecast driven risk based applications with a look-ahead of 2-48 hours with write out on minute-scale**



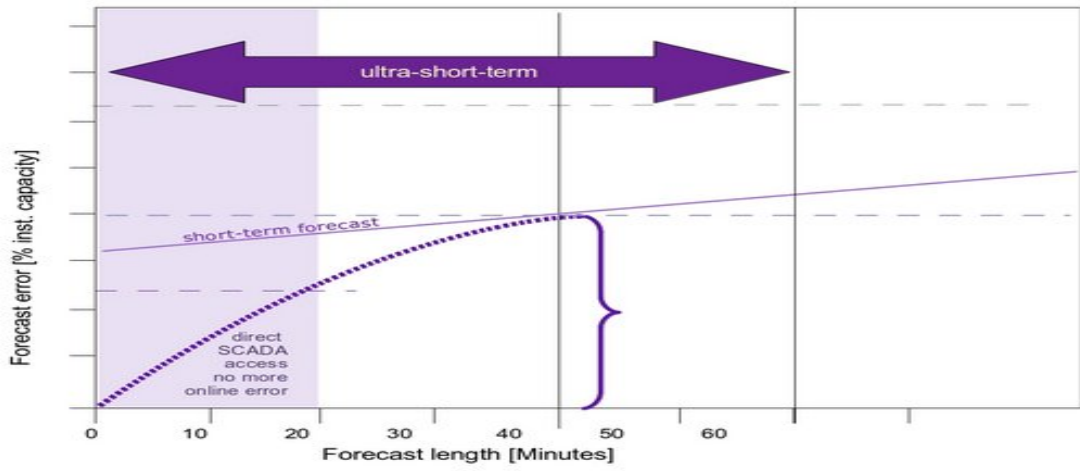
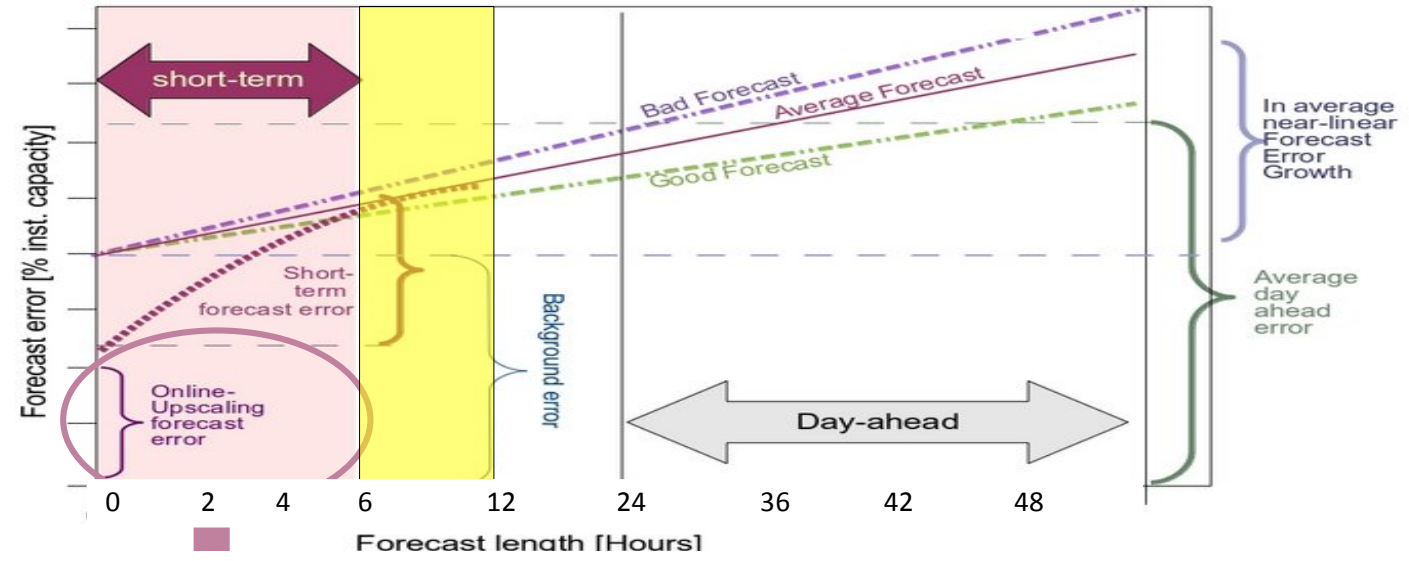


# Move from Yesterday's time scales to Today's requirements

TIME LENGTH of HOURS

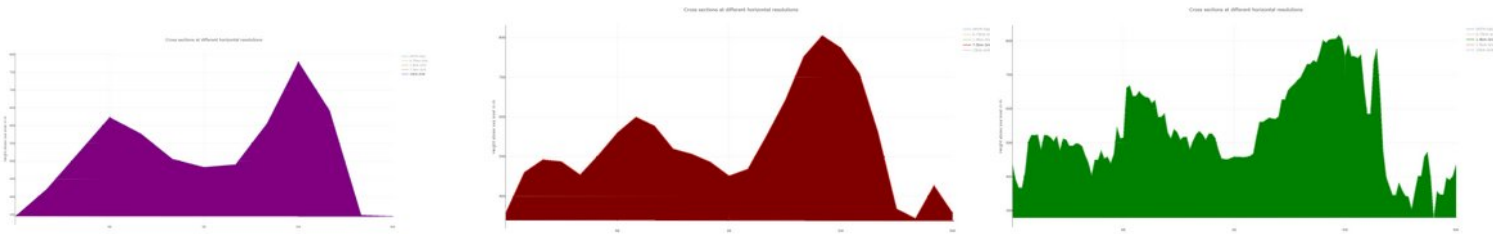


TIME LENGTH of MINUTES

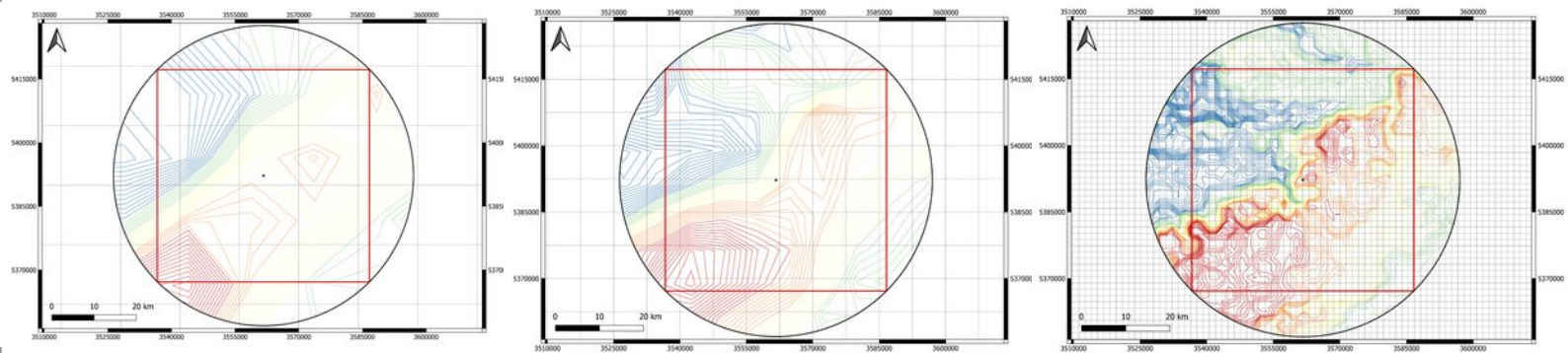


# Visualisations of different horizontal resolution in climate data for the numerical meso scale models

Profile of different model resolutions of the area around the WINSENT test field



Horizontal profile in different resolutions of the area around the WINSENT test field

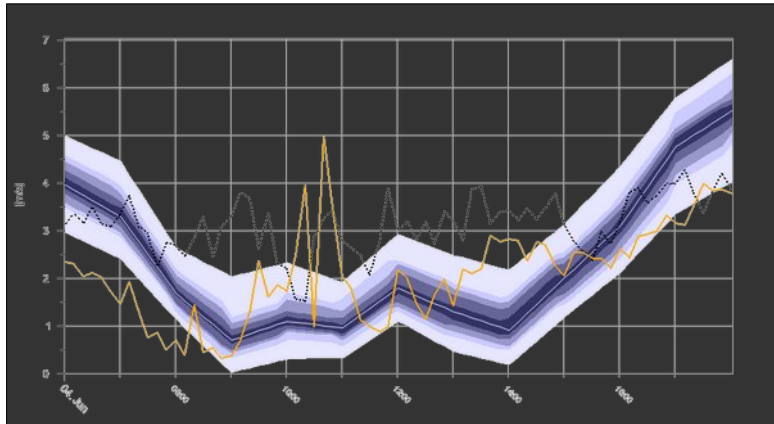
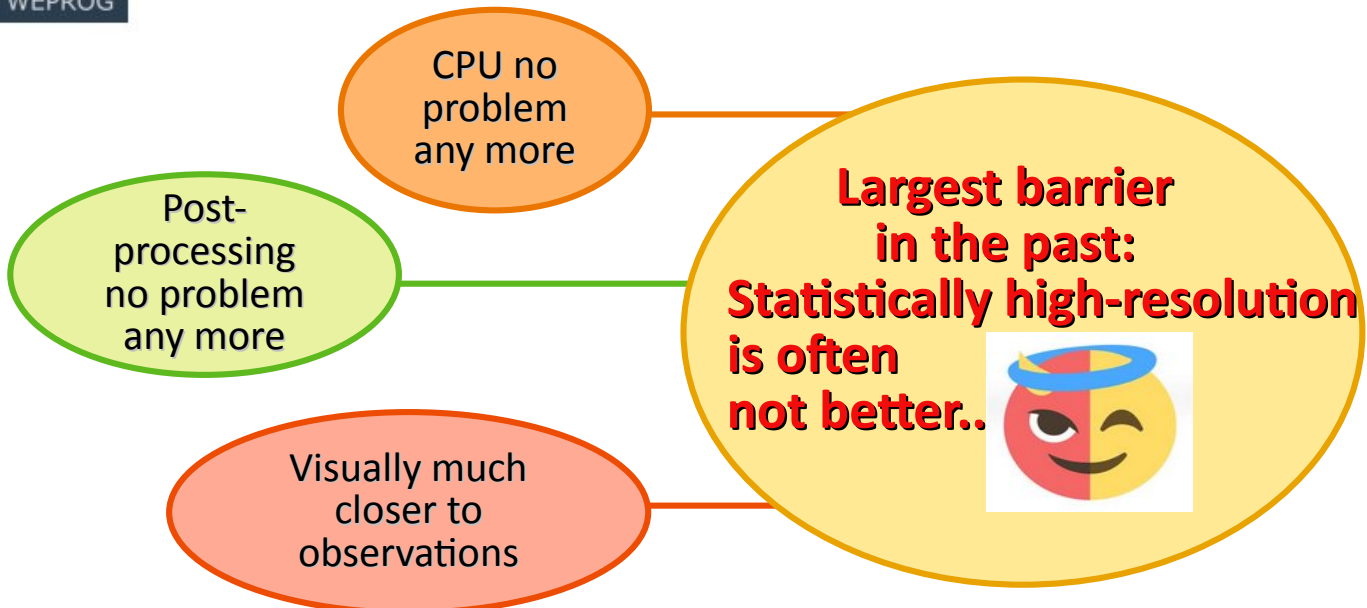


**15km**  
**(operational)**

**7.5km**  
**(operational)**

**2.5km**  
**(research)**

# Minute Scale Forecasting with NWP Ensemble



**HR worse or NO significance!**

Example: Wind Speed at 60m  
DTU Testsite Østerild

Forecast	Bias	MAE	RMSE
LR best member	0.08	1.21	1.53
HR best member	0.11	1.19	1.52

Powerforecasts ca. 20 windfarms in Ireland  
CRPS score: overall performance of prob. Forecast

Forecast	CRPS [MW]	CRPS [% inst. cap]
HR	10.5	5.8
LR	10.9	6.0
Reference	20.6	11.5

BRIER score example: accuracy of a probabilistic event forecast

Forecast	20MW 1hour	30MW 3 hours	40MW 3 hours	60MW 3 hours
HR	0.0501	0.089	0.0513	0.021
LR	0.0459	0.084	0.0464	0.018
$\Delta(HR - LR)$	0.0043	0.0053	0.0049	0.0028

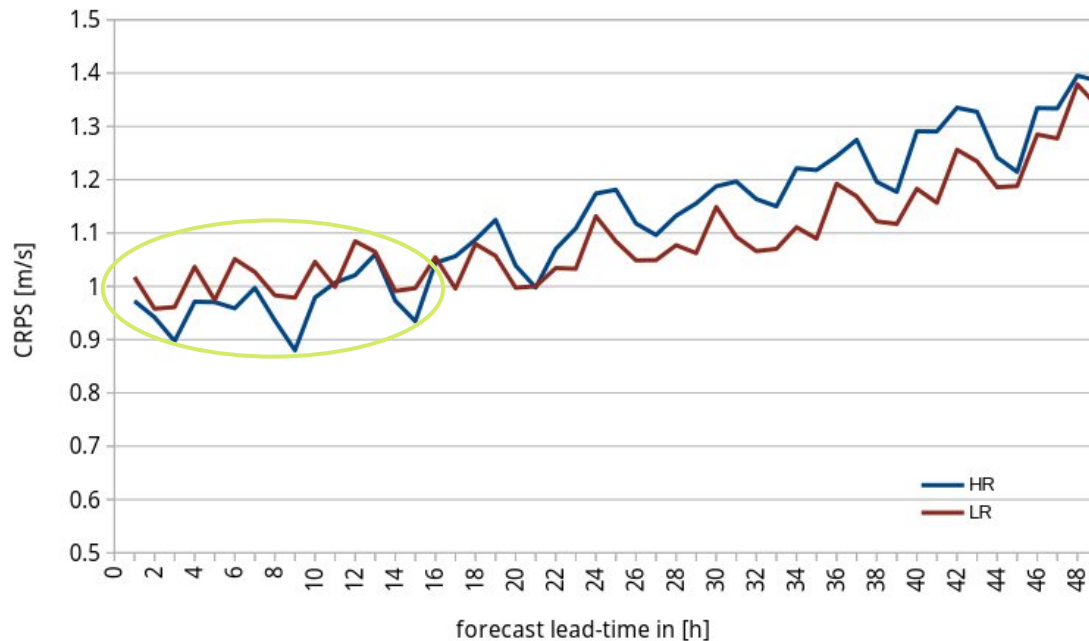




# Example Evaluation of Wind Speed at a Danish coastal site

## Assessment of a high-resolution versus low resolution ensemble system

Forecast Type	CRPS	Improvement to Reference [%]
Reference	1.6635	
<b>Lead-time 6-11h</b>		
HR	1.140	-31.5
LR	1.159	-30.3
<b>Lead-time 0-48h</b>		
HR	1.1236	-32.5
LR	1.0925	-34.3



Result from Test 1:

High-resolution setup has only value in the first 12 hours

Conclusion:

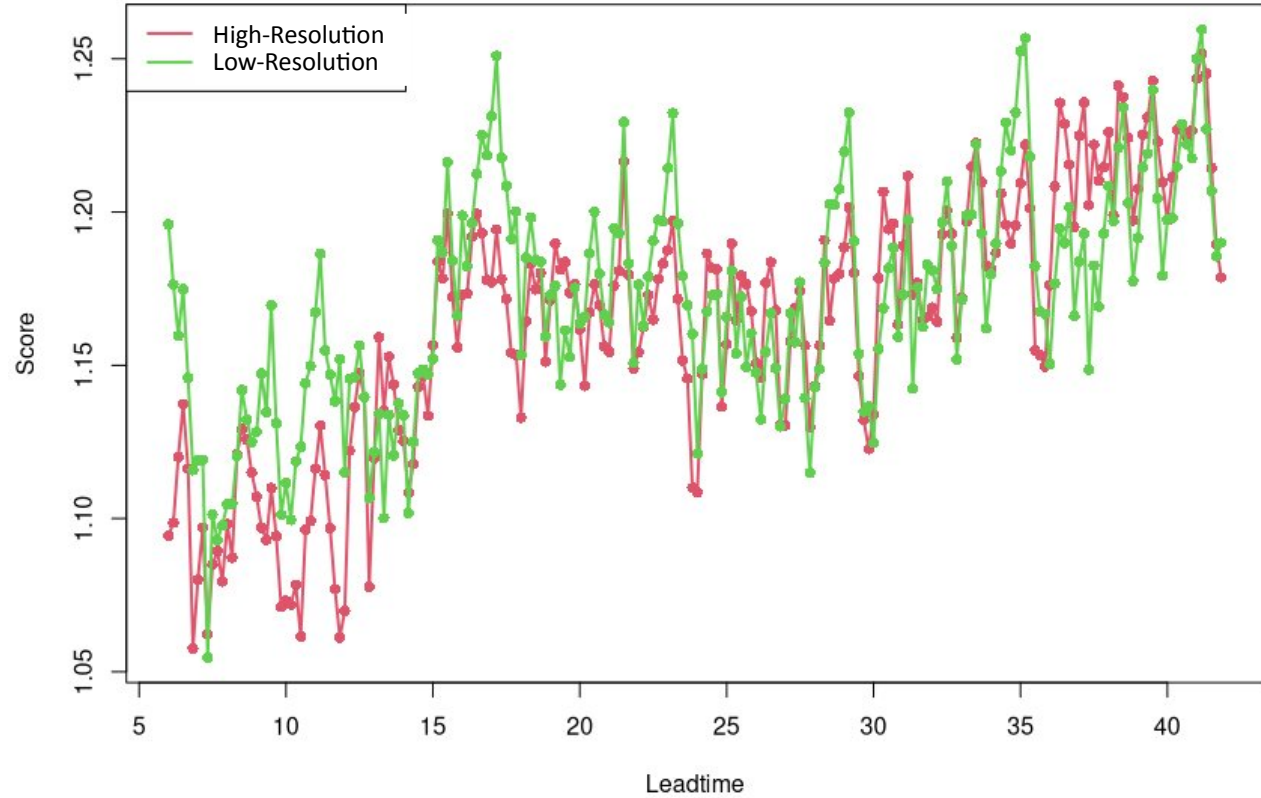
High-resolution setup can be complementary in the intra-day...



# Example Verification at WINSSENT Testsite

## CRPS: Wind Speed at 100m

CRPS by leadtime



**High-Resolution worse or  
NO significance!**

Forecast	CRPS
High-Resolution (f1)	1.163
Low-Resolution (f2)	1.171

Location: WINSSENT Test site, Germany  
Data period: Jan. - Nov. 2021  
High-resolution ensemble: 5km  
Low-resolution ensemble: 15km



## Fundamental verification challenge

Back to the roots... NWP model development has for decades been based on

***If you compare two model setups and they give equal objective verification scores, then take the one you believe is the better one, more future compatible, less costly and easier to maintain!!!***

The dilemma mostly disfavours changes..



**A new objective approach was needed that identifies detailed forecast skills for each model system of our target parameter.**

We designed 24 tests with each 10 event-definitions to ensure that the answer “yes” implies an “event of interest” and which is difficult, but still realistic to forecast.

**The correct answer is defined using the observation time series.**

**The forecast skill is quantified at different time-scale from the ratio of correct answers.**



# Observational Data for the Verification

Site: Stötten ([WINSENT](#) Testsite),

Southwest Germany

Period 1: Jan. - November 2021

1 Metmast with 4 measurement heights to 100m

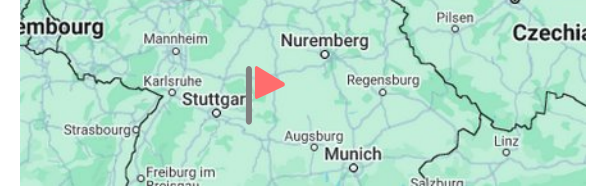
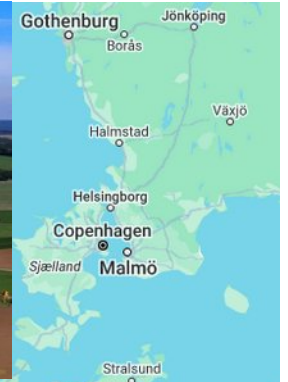
Period 2: October 2023 – January 2024

Campaign in collaboration with ZSW and AQ Systems

1 AQ510 SODAR measuring up to 300m

1 Windscanner LIDAR measuring up to 1000m

2 Metmasts up to 100m





# About the MSEPS Forecast data

**Verification period:** 145 intersecting days in 2021

**Data Sampling and forecast horizon:**

10-minute resolution from native model hour time resolution (<5min)

4 forecasts 48H per day 00,06,12,18 UTC

Only the 6-12 hours lead time is considered

**LowResolution Setup – LR -**

75 members and 32 levels – size 300 x 300

**High-Resolution Setup – HR -**

8 members and 60 levels - nested Domain in LR with 100 times more grid points

Acceleration at local scale can be ca. 200 times faster in HR than LR



# Test methodologies to compare HR and LR

## TEST 1: Skill measured using a deterministic answer per forecast

Symbol  $\Sigma$  indicates a double sum  $\Sigma_{time} \Sigma_{fc}$

**a:** Use all data

$$\text{Skill}_{T1a} = \frac{\Sigma(\text{True}_{fc} \text{True}_{obs} + \text{False}_{fc} \text{False}_{obs})}{\Sigma(\text{True}_{obs} + \text{False}_{obs})}$$

**b:** Count only during events

$$\text{Skill}_{T1b} = \frac{\Sigma \text{True}_{fc}}{\Sigma \text{True}_{obs}}$$

## TEST 2: Skill measured using probability for each evaluation

Use only events

$$\text{Skill}_{T2} = \frac{\Sigma_{time} \frac{100}{\#members} \Sigma_{fc} \text{True}_{fc}}{\Sigma_{time} \text{True}_{obs}} \geq \text{LTPCD} ? 1 : 0$$

**Lower Threshold Probability  
Criteria of Detection (LTPCD)**



## Test 2: Probabilistic answer using Ensembles

We determine **Lower Threshold Probability Criteria of Detection (LTPCD)** for each model setup targeted the best detection.

The optimal **LTPCD** value is in the end a user choice determined by the cost ratio between success, miss and false alarms.

In this verification we select 3 different sample values to illustrate how the **LTPCD** value influence the result.

### **LTPCD** for the HR setup

at least 1 member detects => 12.5% probability

at least 2 members detect => 25% probability

at least 4 members detect => 50% probability

### **LTPCD** for LR is set lower to compensate for reduced model resolution

at least 3 members detect => 4.4% probability

at least 5 members detect => 6.6% probability

at least 10 members detect => 13.3% probability



# Event-type definition at WINSENT Testsite for 100m wind speed

- 1 • Does the wind speed **increase** at least 2.5m/s ?
- 2 • Does the wind speed **increase** at least 2.0m/s ?
- 3 • Does the wind speed **increase** at least 1.5m/s ?
- 4 • Does the wind speed **increase** at least 1.0m/s ?
- 5 • Does the wind speed **increase** at least 0.5m/s ?
- 6 • Does the wind speed **drop** at least 0.5m/s ?
- 7 • Does the wind speed **drop** at least 1.0m/s ?
- 8 • Does the wind speed **drop** at least 1.5m/s ?
- 9 • Does the wind speed **drop** at least 2.0m/s ?
- 10 • Does the wind speed **drop** at least 2.5m/s ?

*Event-detection* was repeated every 10 min for 24 *DurationWindow* tests ranging from 10 min - 4h

→ YES answers from the observation data are considered events  
→ NO answer is non-events.

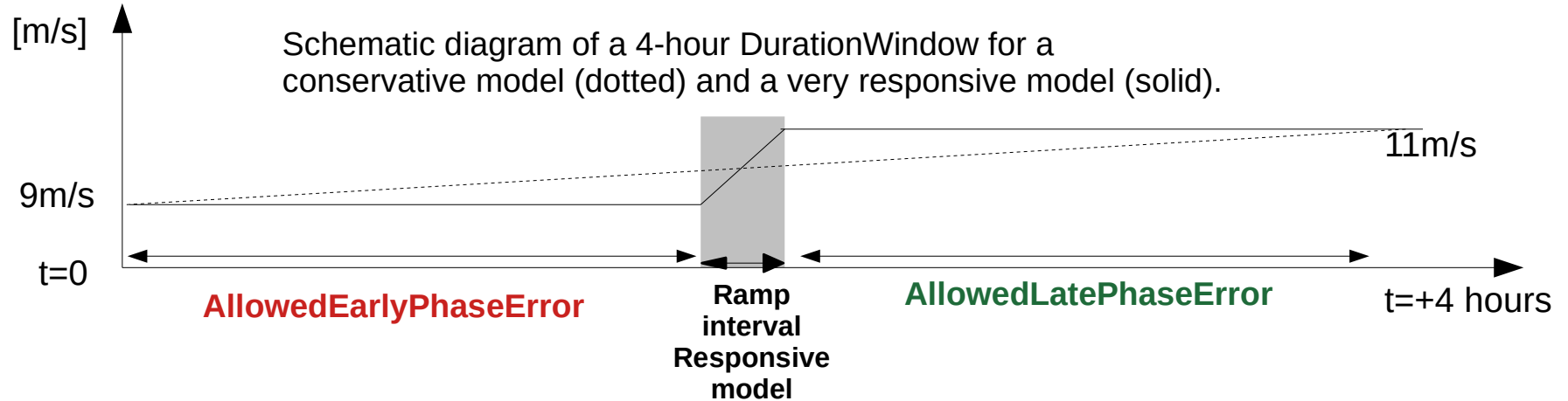
25.030 tests got a YES, which is approximately 1 every 20min.





# Duration Window and Ramp Interval concepts

We use 24 different DurationWindows ahead in time at every 10 minute interval. Within the DurationWindow, a change in wind speed can occur at a shorter interval (the RampInterval) together with an AllowedPhaseError.



The following relation applies

$$\text{DurationWindow} = \text{RampInterval} + \text{AllowedEarlyPhaseError} + \text{AllowedLatePhaseError}$$

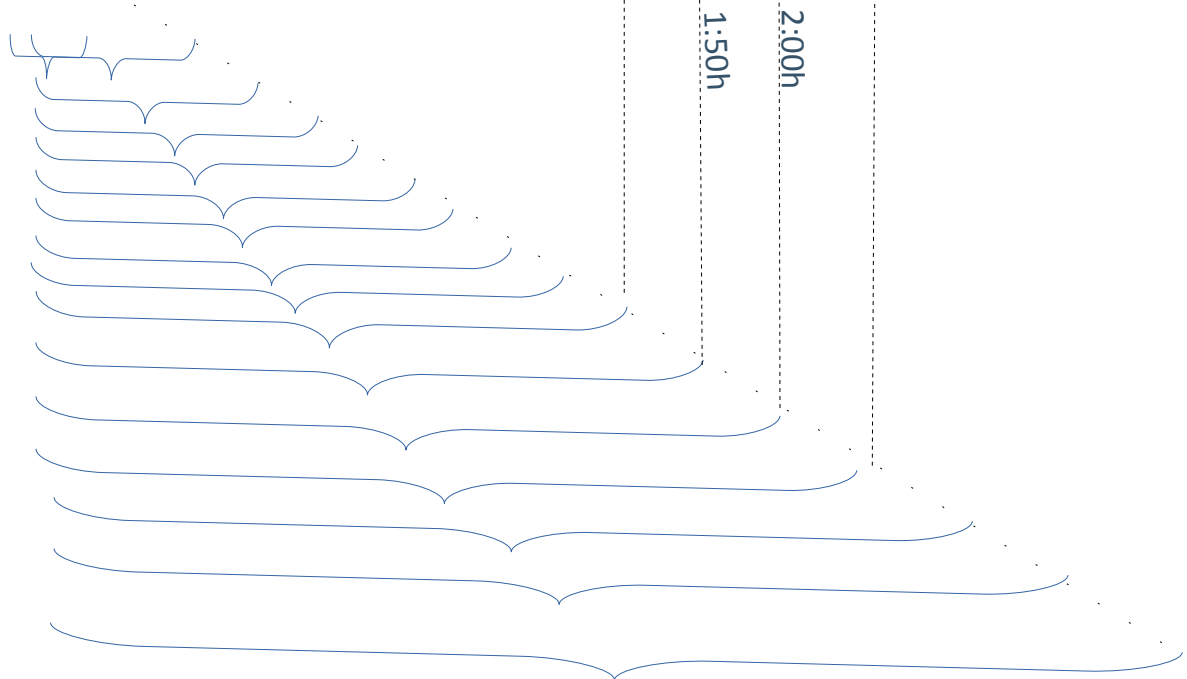
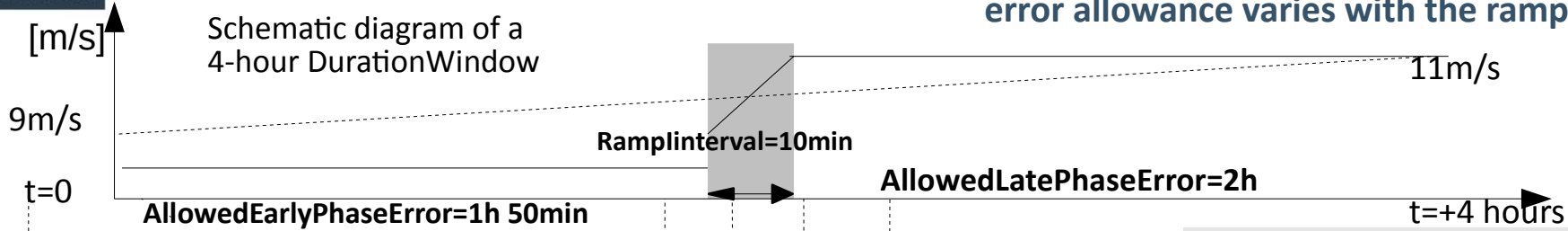
The RampInterval cannot exceed the DurationWindow, but it is allowed to be shorter. If the RampInterval is short, then a correspondingly longer phase error tolerance is allowed.

The YES/NO answer is determined from difference between start and end point of the DurationWindow.



# Ramp Interval definition

## Increase of 2m/s + explanation of how the phase error allowance varies with the ramp window



Q2) Does the wind speed increase at least 2.0m/s ?

A slowly reacting model (dotted ramp) will answer YES at DurationWindow=4h, but NO for all shorter DurationWindow.

The responsive model (solid ramp) will answer YES for DurationWindow >= 2h and NO for the shorter.

What is correct depends on the observed wind speed change.

At time 1:50h and DurationWindow=1 the responsive model will answer YES. At that point the model can be awarded only if it has no phase error



# Phase Error Allowance

The table below show hex numbers of how many 10min intervals a ramp may be shifted. Rows represent the DurationWindow scale and columns represent the RampInterval.

Time scale [10 min]		RampInterval																									
		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24		
DurationWindow	1	0																									
	2	0	0																								
	3	1	0	0																							
	4	1	1	0	0																						
	5	2	1	1	0	0																					
	6	2	2	1	1	0	0																				
	7	3	2	2	1	1	0	0																			
	8	3	3	2	2	1	1	0	0																		
	9	4	3	3	2	2	1	1	0	0																	
	10	4	4	3	3	2	2	1	1	0	0																
	11	5	4	4	3	3	2	2	1	1	0	0															
	12	5	5	4	4	3	3	2	2	1	1	0	0														
	13	6	5	5	4	4	3	3	2	2	1	1	0	0													
	14	6	6	5	5	4	4	3	3	2	2	1	1	0	0												
	15	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0											
	16	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0										
	17	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0									
	18	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0								
	19	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0							
	20	9	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0						
	21	a	9	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0					
	22	a	a	9	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0				
	23	b	a	a	9	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0			
	24	b	b	a	a	9	9	8	8	7	7	6	6	5	5	4	4	3	3	2	2	1	1	0	0		

AllowedPhaseError = (DurationWindow - RampInterval)/2

Dark Green cells are the most difficult to forecast

Strongest Phase error allowance

All numbers represent number of 10 minute intervals



# On robustness of the verification: number of Tests

Wind Speed Change	sign	Total Tests Count	Number of Events	Number of 10min scales
2.5	up	30829	763	1
2.0	up	30829	1163	4
1.5	up	30829	1803	14
1.0	up	30829	2953	56
0.5	up	30829	5077	262
0.5	down	30829	5788	262
1.0	down	30829	3409	68
1.5	down	30829	1984	19
2.0	down	30829	1258	8
2.5	down	30829	832	3
		308290	25030	697



# Summary of the steps required to find skill differences

## Preparation:

- Define 1-24 *Duration Windows* in 10-minute intervals (finest common granularity of obs/fc)
- Assemble 10-minute resolution of forecasts and observed values (hourly data is not enough!)
- Bias correct the forecasted wind speeds (same approach on each model system)
- Formulate ca. 10 questions on target variable changes from normal to extreme case
- Define **LTCPD threshold** for each ensemble member to be used in the probability calculation

## Processing:

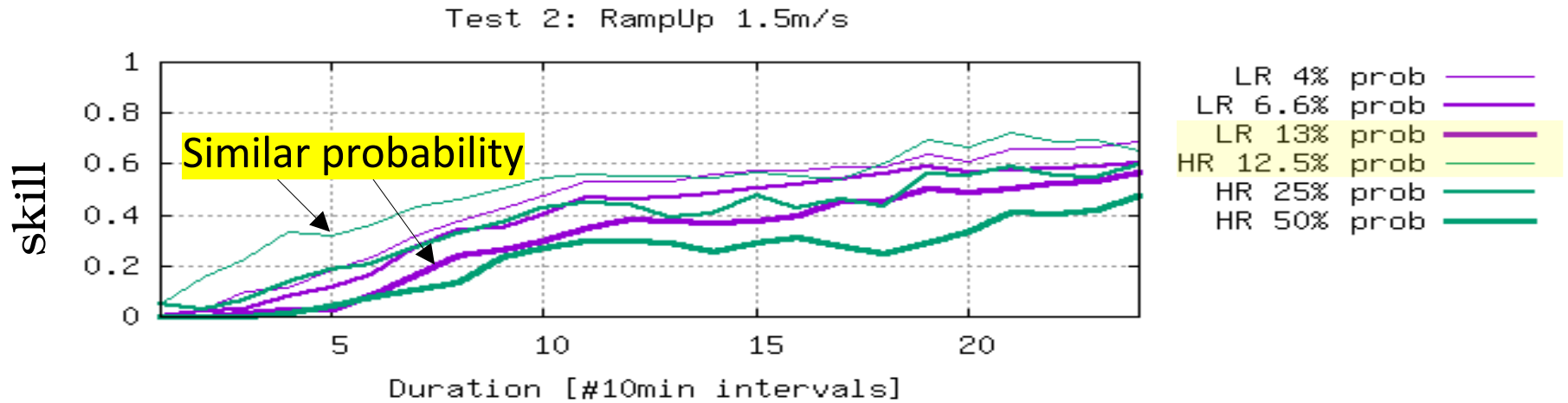
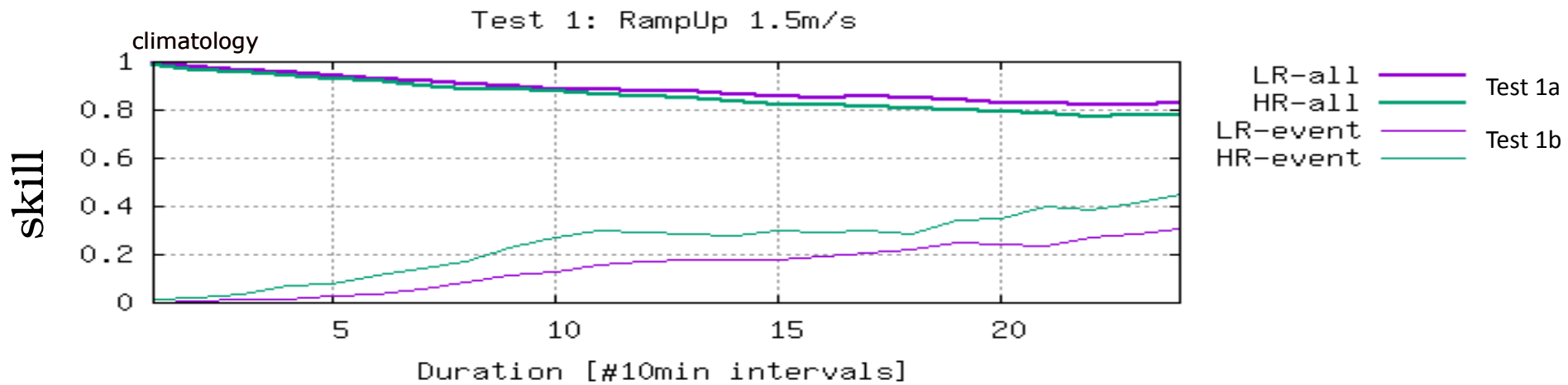
- Step in 10-minute resolution through the year and collect the 24 YES/NO values per model setup
- Calculate skill with 3 different formulas for each question
- Analyze results by question and assemble in table form

## Final Evaluation:

- Perform winner count by row and column in the result table



# Example Result plot for the 1.5m/s Ramp







# Summary of Results from Graphs on Test 1a, Test 1b and Test 2

Wind Speed Change	Up or Down	Test 1a Winner of all Data	Test 1b Winner of all Events	Test 2 Winner per DurationWindow interval		
				Short (1-6h)	Long (18-24h)	
2.5	up	HR	HR	HR	HR	0 - 4
2.0	up	HR	HR	HR	HR	0 - 4
1.5	up	LR	HR	HR	EVEN	1 - 2
1.0	up	LR	HR	HR	EVEN	1 - 2
0.5	up	LR	HR	HR	LR	2 - 2
0.5	down	HR	HR	HR	LR	1 - 3
1.0	down	HR	HR	HR	LR	1 - 3
1.5	down	HR	HR	HR	EVEN	0 - 3
2.0	down	HR	HR	HR	HR	0 - 4
2.5	down	HR	HR	HR	HR	0 - 4
		3 / 7	0 / 10	0 / 10	3 / 4	6 - 31



... the "typical" barrier for change ...



# Verification Results from LR and HR comparison

## -- summary --

Note: The slope of the curves differ between Test 1a and Test 1b, because both model setups forecast non-events with high skill and events with lower skill.

**Test 1a: Good performance in at +/-0.5m/s at all time scales can be achieved with conservative model setup.**

→ High skill at short time scales occur because the amount of events is low compared to non events.

→ At longer time-scale skill is actually lower.

**Test 1b: low skill at short time scales and higher skill at longer time scales,**

because the longer time scales are equivalent larger spatial scales

**Test 2: probabilistic method provides useful and realistic skill comparison – even for a small ensemble !**

→ Moderate ramps have higher skill and HR and LR perform equal

→ The hit rate decreases for fast ramps, but is significantly higher for HR than LR

→ Fast ramps have some skill with HR, no skill with LR

→ The 12.5% **LTPCD** condition of HR is the most reliable of all 6 **LTPCD** conditions

→ LR detection is generally lower except for +/- 0.5m/s, although smaller **LTPCD** values are used



# Result from the 4 comparison tests

HR setup has lower skill in modest ramp ups (0.5 and 1.0m/s) possibly due to false alarms. It is likely that HR generates meso-scale activity in this type of weather. If this type of activity does not happen, it could look like a weak point of HR.

LT shows higher skill for long time-scales and small ramps (0.5m/s) due to higher LTPCD value. If we compare **LTPCD=12.5%** for HR and **LTPCD=13%** for LR, then we find that HR and LR are even on the long time-scales.

The difficult part of Test 2 are the short time-scales without phase error allowance. Although HR is not great, it is outperforming LR on these time scales.



## Forecast Skill

The verification methodology provides a sharp test to quantify forecast skill at the important risk related short time-scales.

## Forecast Tuning

Ultimately this verification method can be used to further tune the forecasting toward predicting the need of the amount of primary reserve.



# Lessons Learned and Take-aways

event based probabilistic verification can provide a detailed time-scale dependent analysis of different model setups

event based probabilistic approach is capable to compare fair performance & conditions of better or worse skills

standard metrics are still relevant as independent tests and can reveal important information

NWP models/Ensembles can provide skillful minute-scale forecasts beyond ST horizon and where there are no observations



# Thank you for your attention !

## Questions ...



Funded by:



WIKKI project funded by  
Deutsche Bundesstiftung Umwelt (DBU) Germany  
Project no. 37549/01



**EARS4WindEnergy**

Ensemble-based Approach utilizing a Refined SODAR  
for Wind Energy Applications



EUREKA Eurostar III Call 3 project  
Project no. E2442 / Innovationsfonden 3109-00061B

Supported by:



WinForS Consortium and WINSSENT Testfield Coordinator:  
ZSW Baden-Wurttemberg, Germany

<https://www.windfors.de/en/projects/test-site/winsent/>

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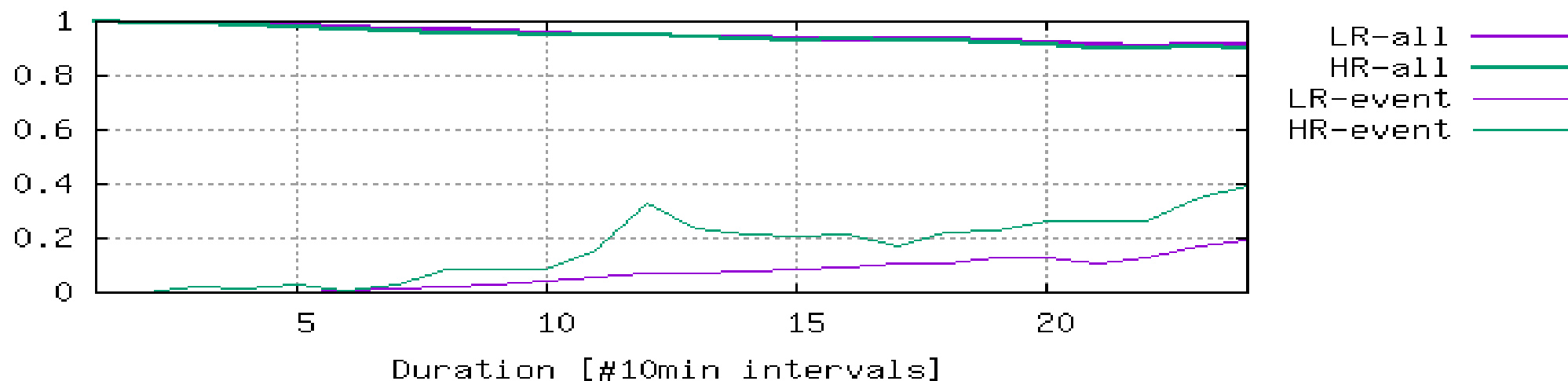
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[weather.weprog.com](http://weather.weprog.com)



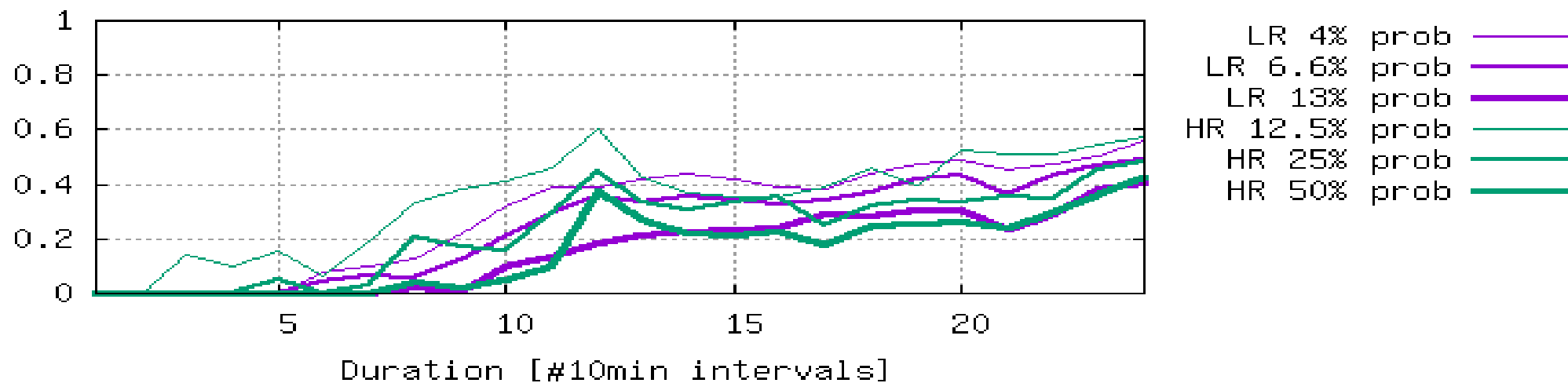
# Result plots



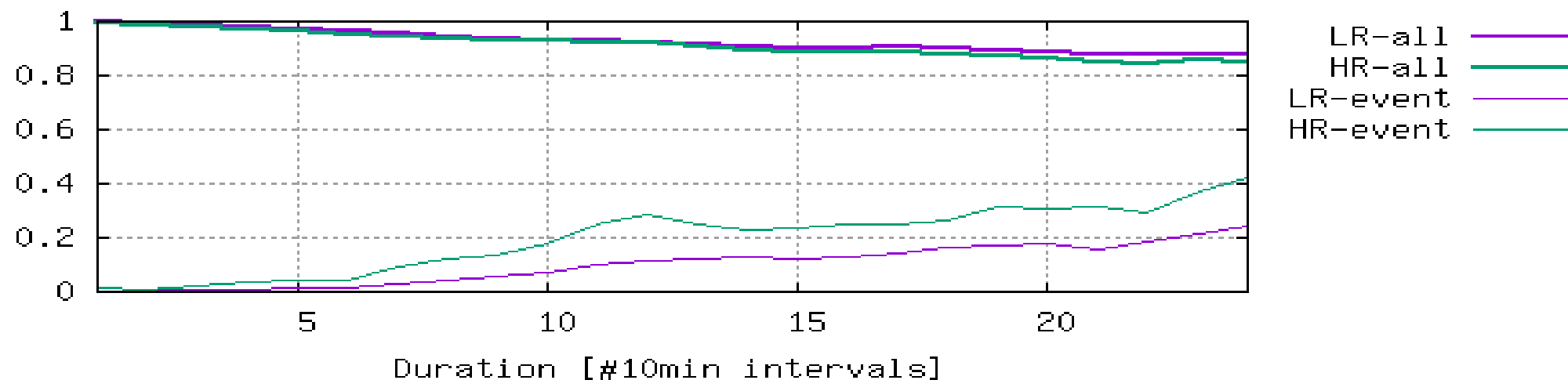
Test 1: RampUp 2.5m/s



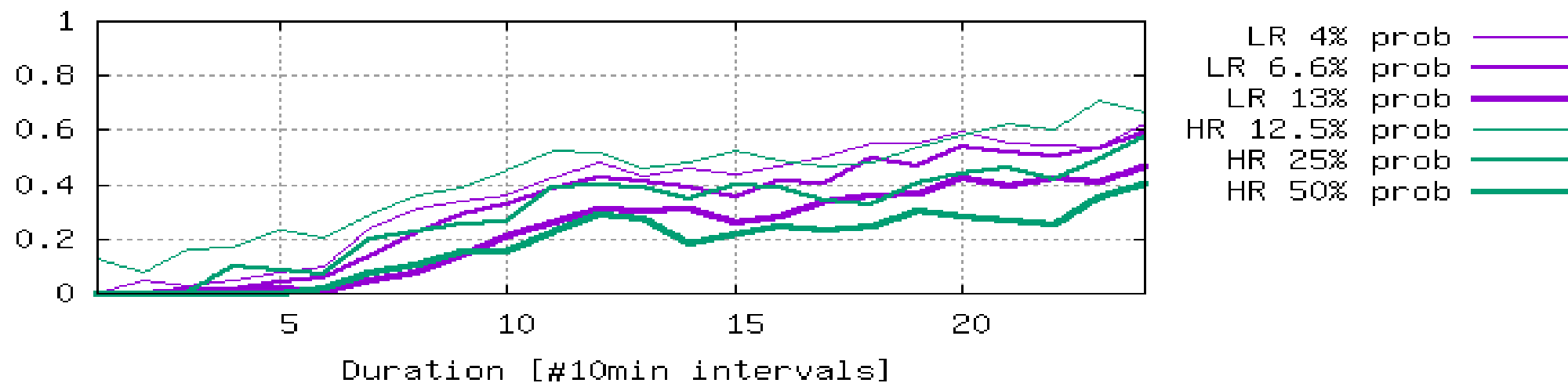
Test 2: RampUp 2.5m/s



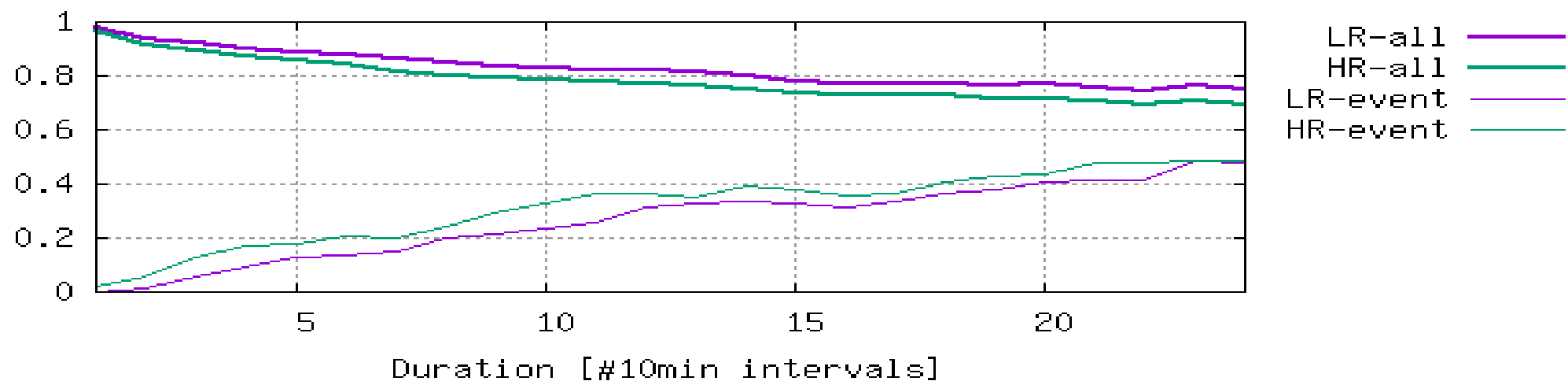
Test 1: RampUp 2.0m/s



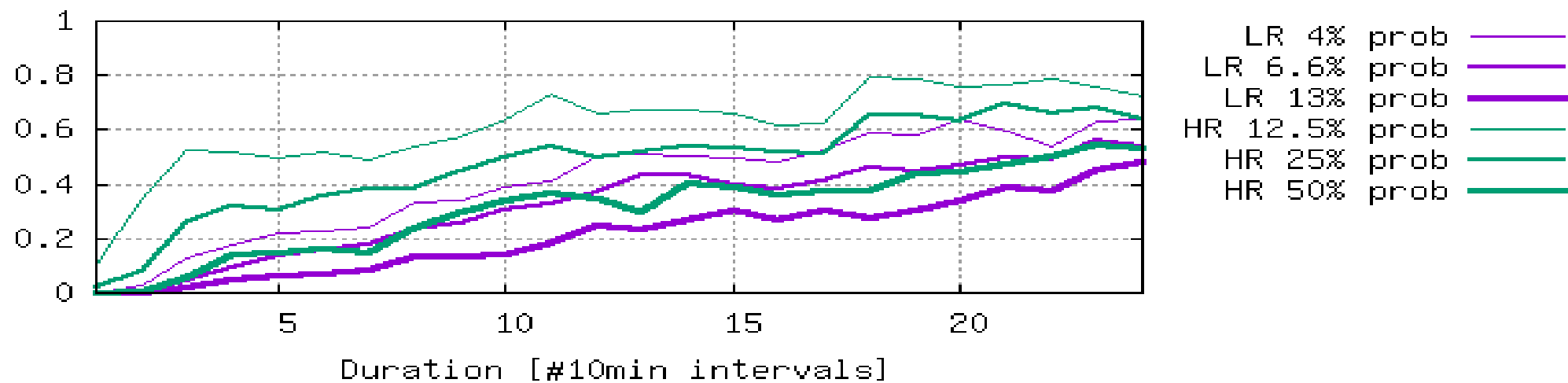
Test 2: RampUp 2.0m/s



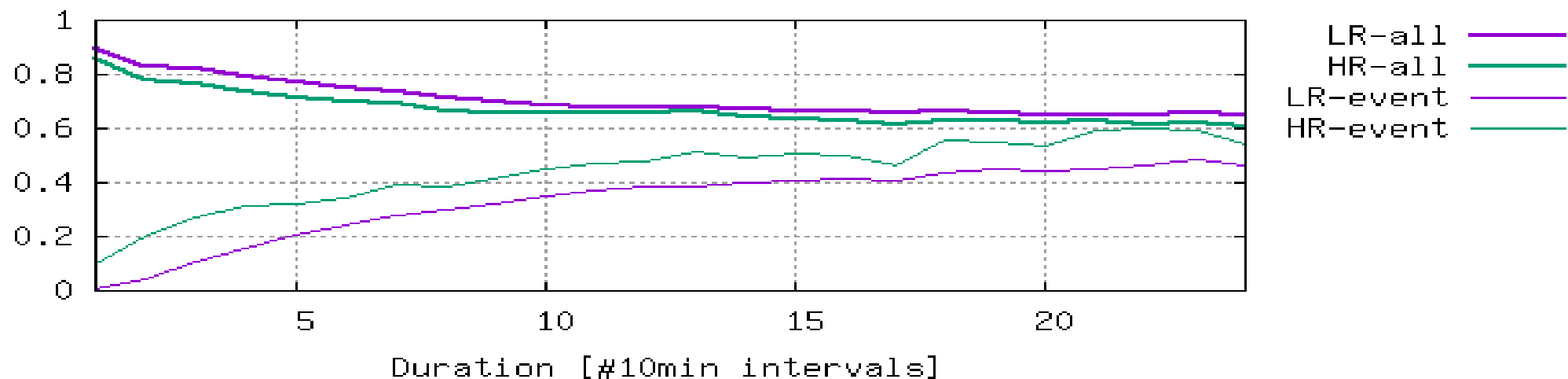
Test 1: RampUp 1.0m/s



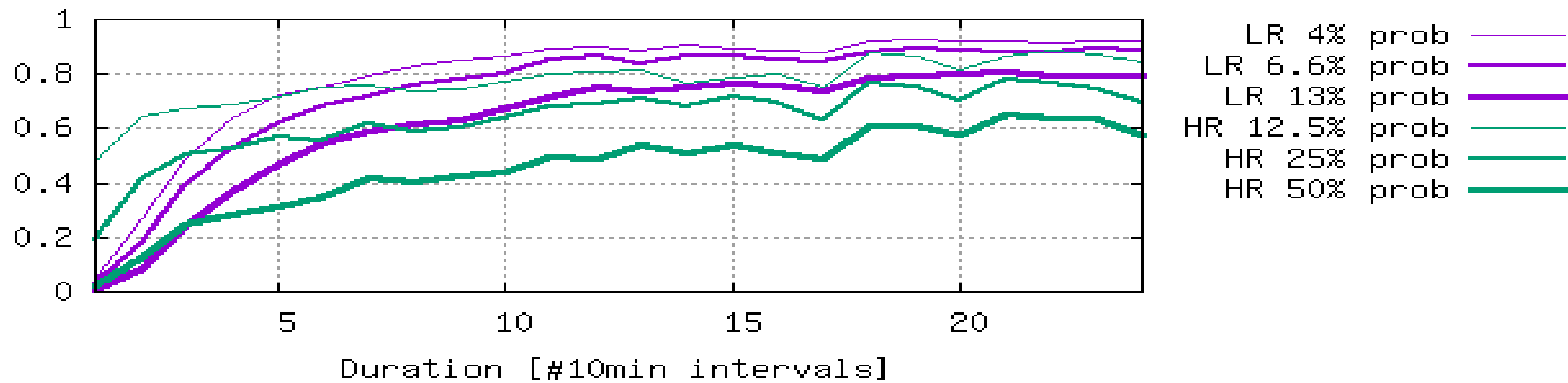
Test 2: RampUp 1.0m/s



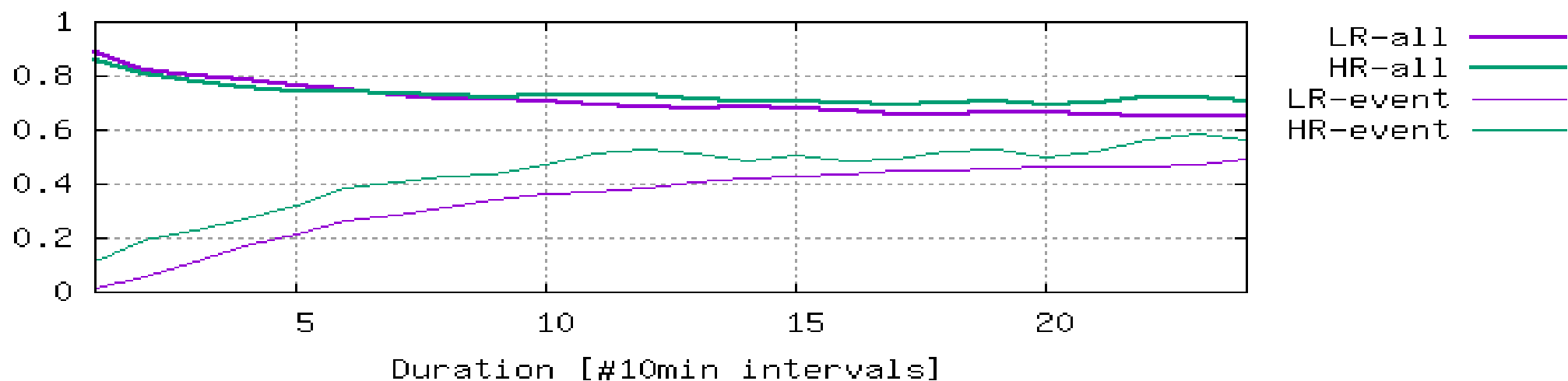
Test 1: RampUp 0.5m/s



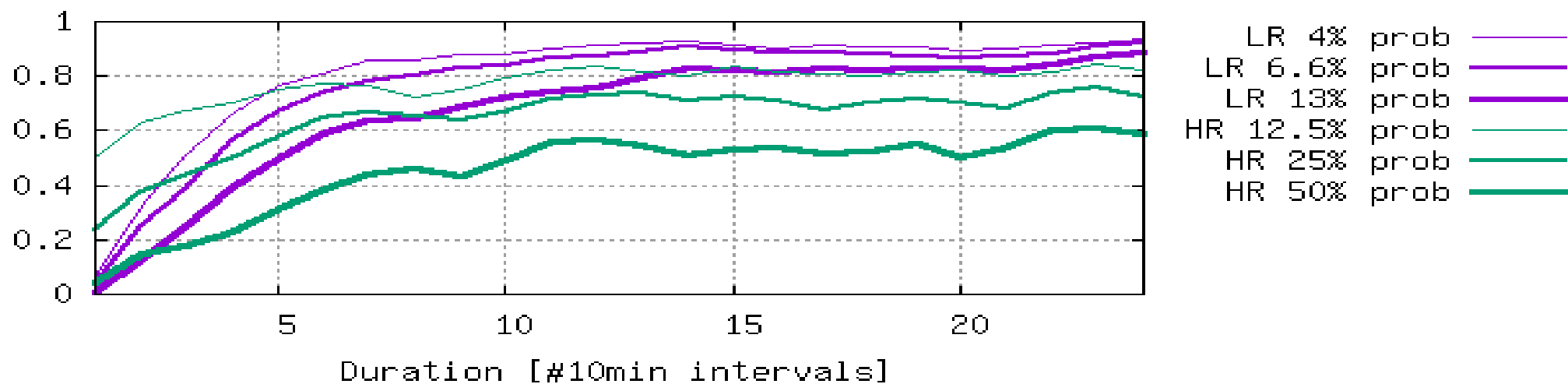
Test 2: RampUp 0.5m/s



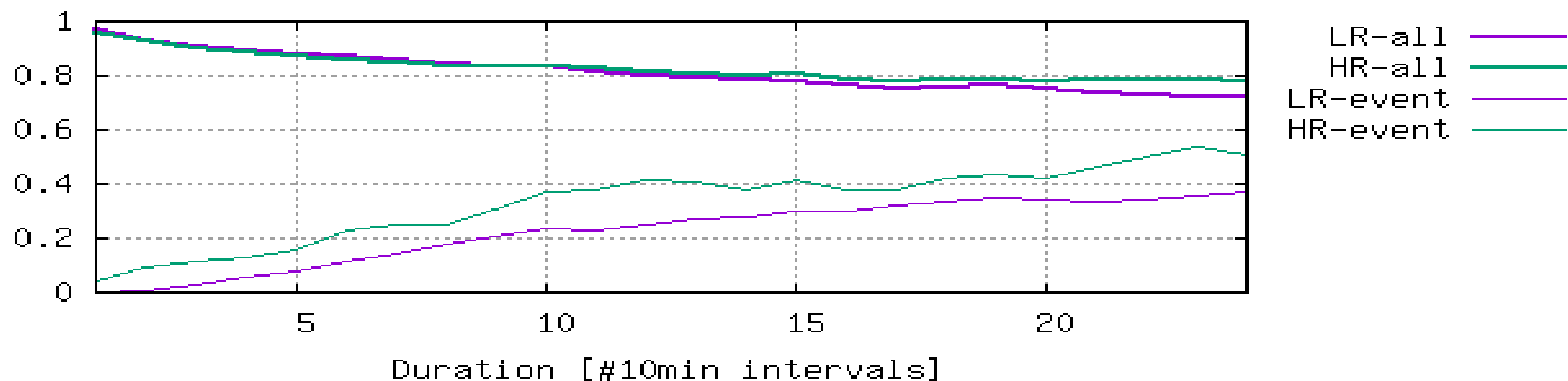
Test 1: RampDown 0.5m/s



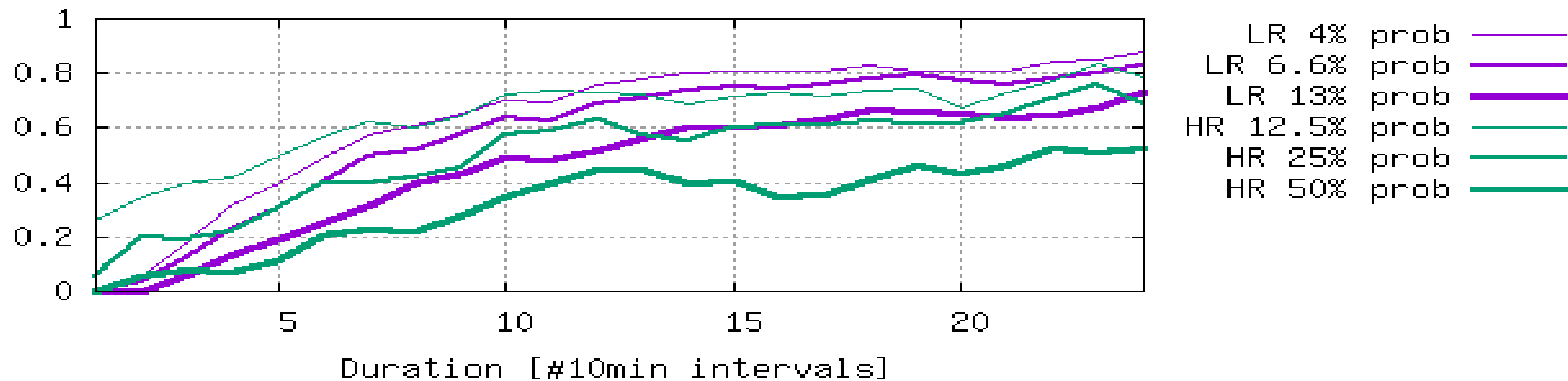
Test 2: RampDown 0.5m/s



Test 1: RampDown 1.0m/s

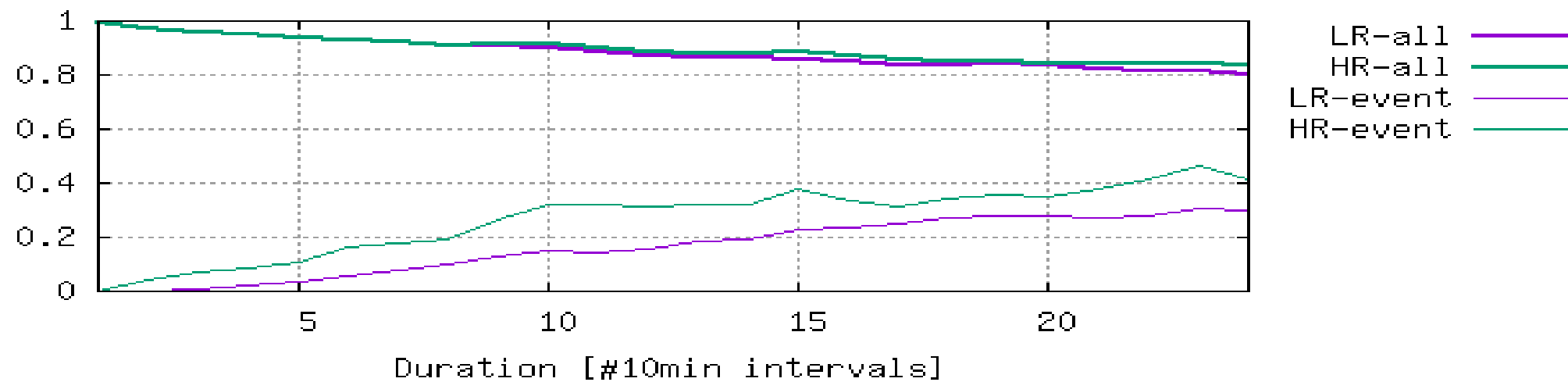


Test 2: RampDown 1.0m/s

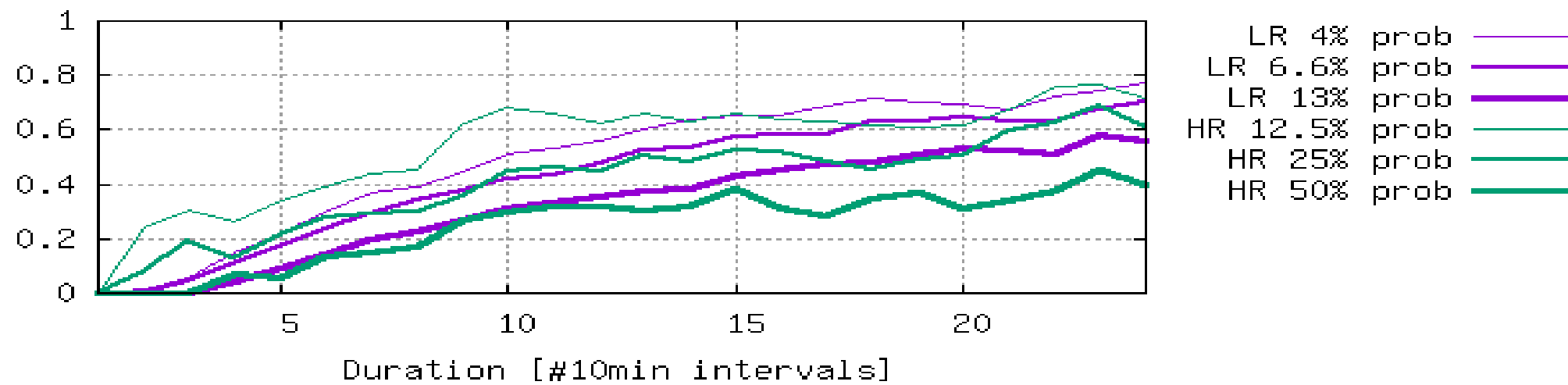




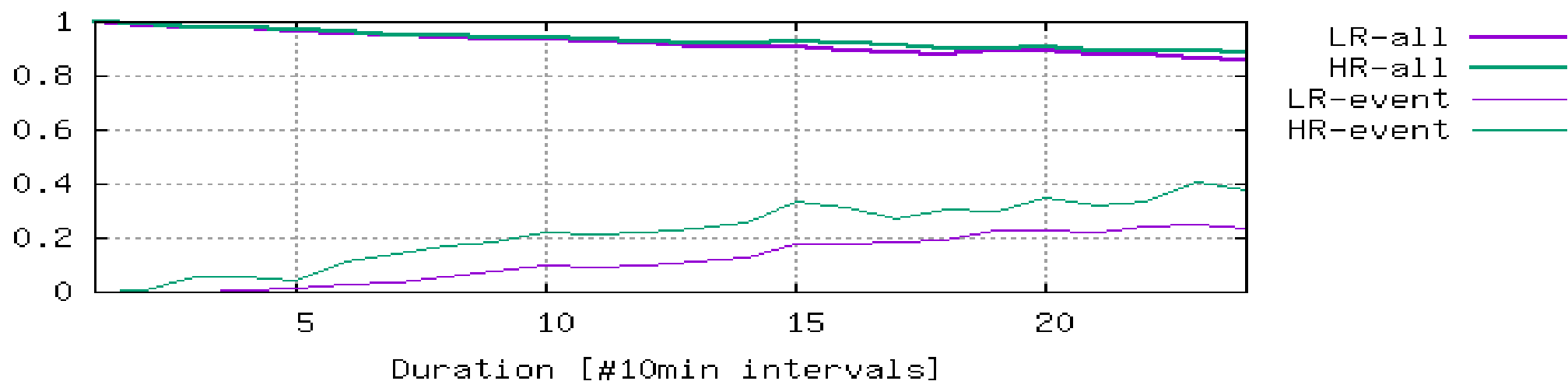
Test 1: RampDown 1.5m/s



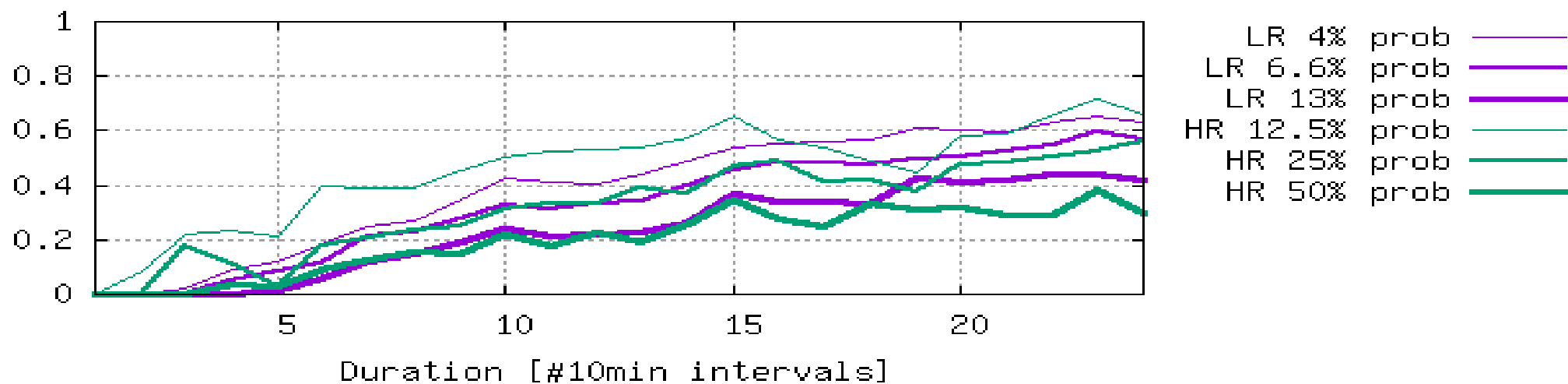
Test 2: RampDown 1.5m/s



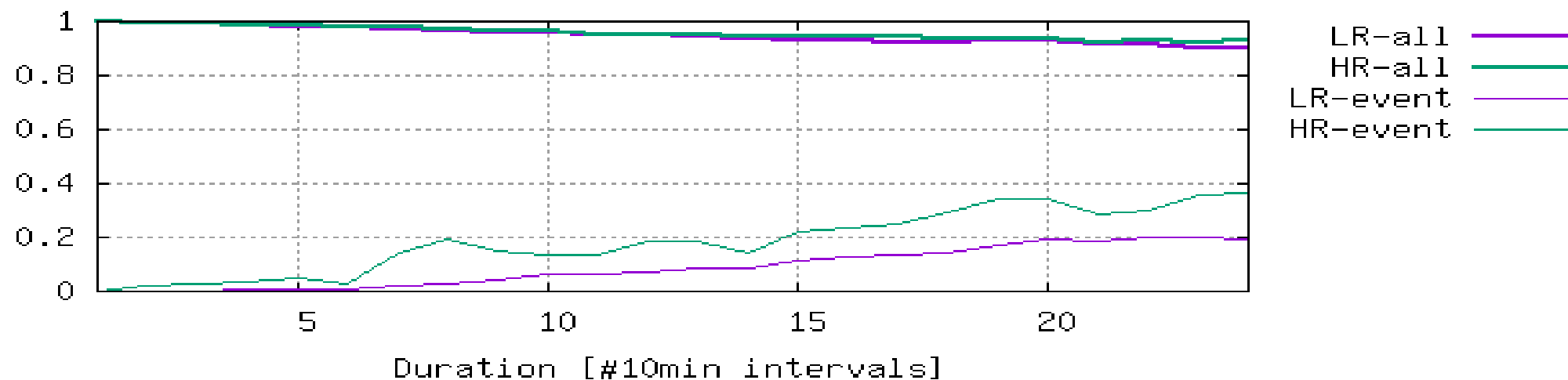
Test 1: RampDown 2.0m/s



Test 2: RampDown 2.0m/s



Test 1: RampDown 2.5m/s



Test 2: RampDown 2.5m/s

