

Using Remote Sensing to Reduce Energy Uncertainty

Key findings from Parsons Brinckerhoff's research paper Ben Inkster – Senior Wind Engineer, Parsons Brinckerhoff ^{16 April 2014}









- 1. Introduction to Parsons Brinckerhoff
- 2. Remote sensing what is it?
- 3. Energy uncertainty what are the causes?
- 4. The test case
- 5. Siting remote sensing devices where do I put these things?
- 6. Data duration how much data do I need?
- 7. Data rejection
- 8. Case study of combined uncertainty
- 9. Fiscal analysis what is it worth \$\$\$?



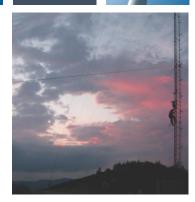
Introduction to Parsons Brinckerhoff

Parsons Brinckerhoff

- Global engineering consulting firm
- Wholly owned subsidiary of Balfour Beatty plc.
- We employ over 14,000 staff worldwide
- 1,500 staff and 13 regional offices in Australia/NZ

We have been advising on wind for over 20 years

- Pre-feasibility and feasibility
- Planning, environment and community consultation
- Detailed design
- Project and construction management
- Financial close out
- Operations and maintenance







Remote sensing – what is it?

SODAR (sonic detection and ranging)

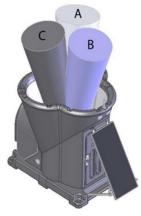




Image Source: Walls, E. 2010

Image Source: Fulcrum3D, 2013

LIDAR (light detection and ranging)







Image Source: CRES 2012

- Emits acoustic pulses
- Receives acoustic backscatter with Doppler shift
- Converts to signal to wind speed vectors at multiple heights
- Emits light signals
- Receives backscattered light with Doppler shift
- Converts to signal to wind speed vectors at multiple heights

Similar concepts, different signal medium, different sensitivities. Site validation is important, but not considered in this presentation

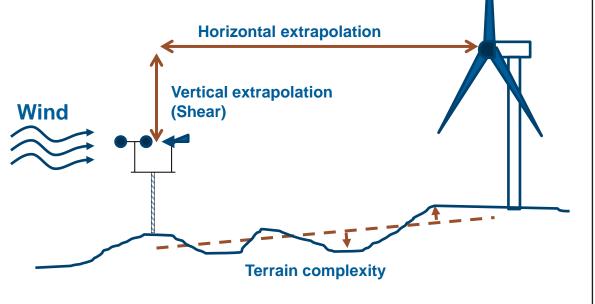


Energy uncertainty – what are the causes?

Wind resource and energy prediction is a complicated process. There are many steps and each adds to uncertainty.

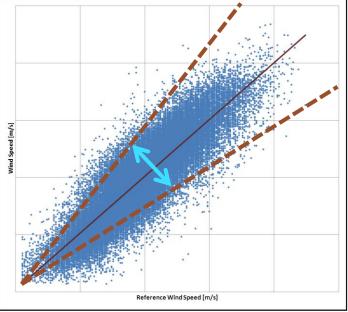
Two sources of uncertainty are considered today:

Wind Flow Modelling Uncertainty Can be affected by siting of remote sensing devices:



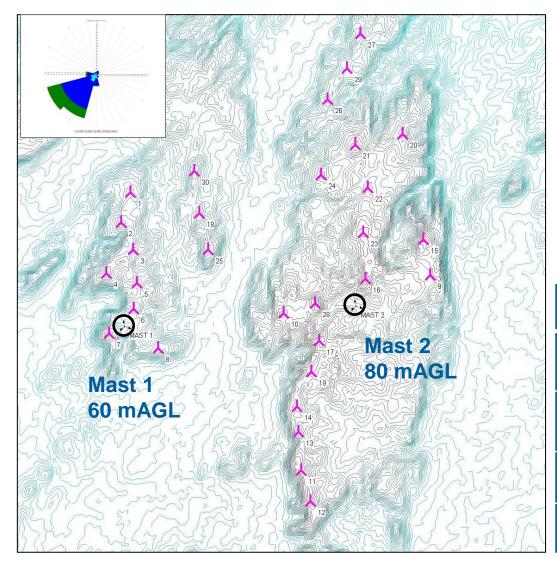
Measure, Correlate, Predict (MCP) Uncertainties

Can be affected by data duration and data rejection (among others)





Test case – a sample site to test uncertainty



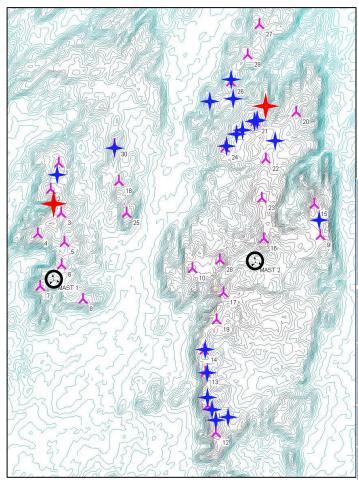
30 x 3 MW WTGs with a hub height of 80 m 2 wind monitoring masts: Mast 1 60 mAGL Mast 2 80 mAGL

Source of uncertainty	Energy uncertainty
Horizontal extrapolation using linear wind flow model	7.7%
Vertical extrapolation from measurement height to hub height	4.5%
Terrain complexity	0.7%
Combined wind flow model uncertainty	8.9%



Siting remote sensing devices – Test results

Two scenarios tested on the Sample Site, assuming <u>two SODARs</u> can be used on-site:



<u>Scenario 1 – Visual/Intuitive Selection:</u> Two SODAR locations chosen by 10 engineers (+) <u>Scenario 2 – Systematic Selection:</u> Two SODAR locations selected using PBs systematic uncertainty model (+)

Source of uncertainty	<u>Sample site</u> <u>default</u> energy uncertainty (no remote sensing)	default_energy Visual intuitive uncertainty (no energy			
Horizontal extrapolation using linear wind flow model	7.7%	4.7%	4.6%		
Vertical extrapolation from measurement height to hub height	4.5%	3.6%	1.2%		
Terrain complexity	0.7%	0.7%	0.4%		
Combined wind flow model uncertainty	8.9%	6.0%	4.8%		



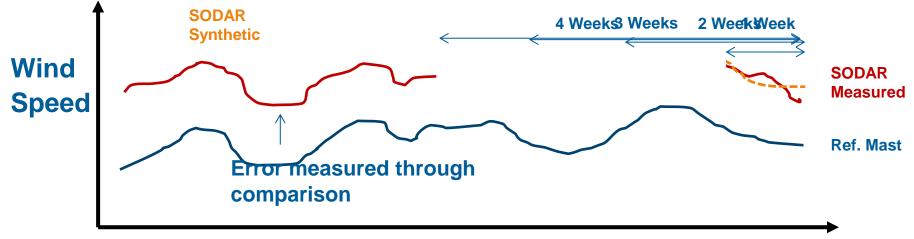
Data duration – test method

MCP was undertaken using SODAR data and reference mast data for the Sample Site

Using 9 shortened data durations (e.g 1 week, 2 weeks...16 weeks)

MCP is used to predict SODAR data for one year (synthetic)

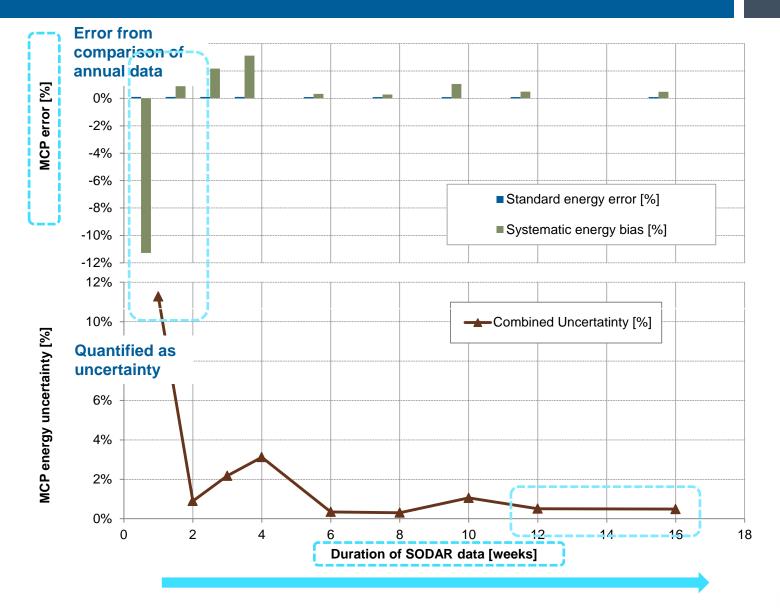
Predicted data is compared to measured data for one year



Time (1 Year)



Data duration – test results







Remote sensing offers diagnostic metrics, such as:

- Signal to Noise Ratio
- Consistency

"Quality Factor" used at the Sample Site

High Quality Factor means high quality data

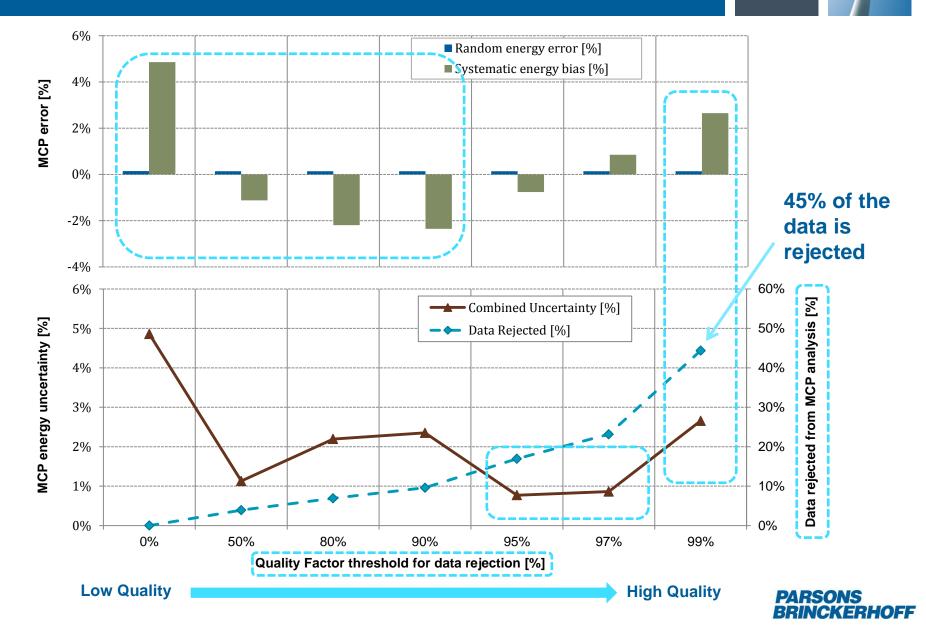


MCP was undertaken using SODAR data and reference mast data for the Sample Site

SODAR data was filtered using seven different Quality Factor thresholds: 0%, 50%, 80%, 90%, 95%, 97% and 99% MCP is used to predict SODAR data for one year (synthetic) Includes low Includes high Predicted data is compared to measured plata for one year data data



Data rejection – test results



Case study – three scenarios for comparison

Scenario 1:

Base case

- No Remote Sensing devices
- Only uses two masts

Scenario 2: Inconsiderate approach

- Two SODARs
- Sited through intuitive approach
- Data duration of 4 weeks
- Data rejected using a Quality Factor of 90%

<u>Scenario 3:</u> Considerate approach

- Two SODARs
- Sited through PBs
 Systematic model
- Data duration of 12 weeks
- Data rejected using a Quality Factor of 95%



Combined uncertainties for a 20 year period:

Source of uncertainty	Scenario 1 Base case	Scenario 2 Inconsiderate approach	Scenario 3 Considerate approach	
Horizontal extrapolation	7.7%	4.7%	4.6%	
Vertical shear extrapolation	4.5%	3.6%	1.2%	
Terrain variation	0.7%	0.7%	0.4%	
MCP data duration	-	1.6%	0.3%	
MCP data rejection	-	1.2%	0.0%	
Other assumed uncertainties	10.2%	10.2%	10.2%	
Combined uncertainty	13.6%	12.0%	11.2%	
			PARSONS	

BRINCKERHOFF

Case study – probabilities of exceedance

Assuming all three scenarios estimate:

P50 = 276 GWh and Capacity Factor = 35%

Probability of exceedance (for 20 year period)	Scenario 1 AEP [GWh]	Scenario 2 AEP [GWh]	Scenario 3 AEP [GWh]	Lenders will use a probability of exceedance to calculate debt size
P50	276.0	276.0	276.0	
P75	250.8	253.7	255.1	Let us assume
P90	228.1	233.7	236.2	they use P95
P95	214.5	221.7	225.0	-
P99	189.0	199.2	203.8	



Assuming: PPA = AUD\$100/MWh and a Lender uses the P95 to evaluate loan

Simplified generation revenue model for the project life:

Years of generation	Scenario 1 Revenue	Scenario 2 Revenue	Scenario 3 Revenue		
1 Year	\$21.5m	\$22.2m	\$22.5m		
10 Years	\$214.5m	\$221.7m	\$225.0m		
20 Years	\$429.0m	\$443.4m	\$449.9m		
Revenue estimate increase from Scenario 1	-	\$14.5m	\$20.9m		
Revenue estimate increase from Scenario 2	-	-	\$6.5m		

Generation Revenue = P95 x PPA x Number of Years







- Remote sensing can reduce uncertainty
- Applying knowledge further reduces uncertainty
 - A considered approach (such as Parsons Brinckerhoff's systematic model) can further reduce uncertainty
 - Data duration can effect uncertainty. A possible duration criteria has been presented
 - Data rejection can effect uncertainty
- A considered and knowledgeable approach can be valued in the millions of dollars \$\$
- Use Parsons Brinckerhoff, we'll save you millions!



Questions?



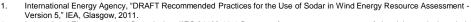
We need enough remote sensing data to reliably perform MCP. How much?

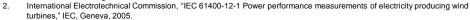
IEA:

"a longer period might be required or measurement periods in different seasons may to necessary, to achieve sufficient representation of varying conditions in the data..." ¹

IEC 61400-12-1 Annex C site calibration, per 10° sector:

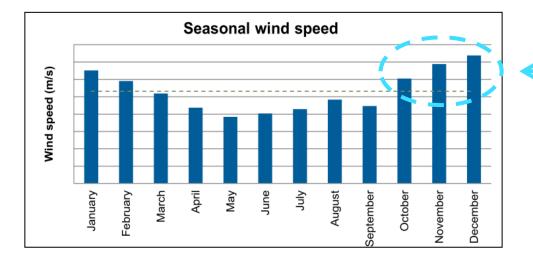
- 6 hours of data below 8 m/s per direction sector
- 6 hours of data above 8 m/s per direction
- 24 hours of data per direction sector ²





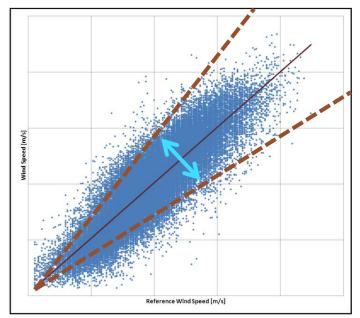


Wind speed and direction are seasonal



Data can be supplemented through Measure, Correlate, Predict (MCP)

Durations of data less than 1 year require supplementation.





MCP was undertaken using SODAR data and reference mast data for the Sample Site

Using 9 shortened data durations (e.g 1 week, 2 weeks...16 weeks)

Compared predicted data to one year of measured data

Shortened SODAR data duration [weeks of data]	1	2	3	4	6	8	10	12	16
Standard uncertainty [%]	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Systematic bias [%]	-11.3	0.9	2.2	3.1	0.3	0.3	1.0	0.5	0.5
Combined uncertainty [%]	11.3	0.9	2.2	3.1	0.3	0.3	1.1	0.5	0.5
Bins complaint with IEC 61400-12-1 Annex C for direction sectors 190-300 deg. [%]	25	47	53	64	75	75	81	83	86



MCP was undertaken using SODAR data and reference mast data for the Sample Site

12 weeks of SODAR data was filtered using differing Quality Factor thresholds (e.g 0, 50, 80, 90, 95, 97 and 99% QF)

Compared predicted data to one year of measured data

Quality Factor threshold [%]	0	50	80	90	95	97	99
Data rejected [%]	0	4	7	10	17	23	44
Standard uncertainty [%]	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Systematic bias [%]	4.9	-1.1	-2.2	-2.3	-0.8	0.9	2.7
Combined uncertainty [%]	4.9	1.1	2.2	2.4	0.8	0.9	2.7

