

Financing Community Wind

Wind Data and Due Diligence
What is the Project's Capacity Factor?

Community Wind Energy 2006
March 8, 2006

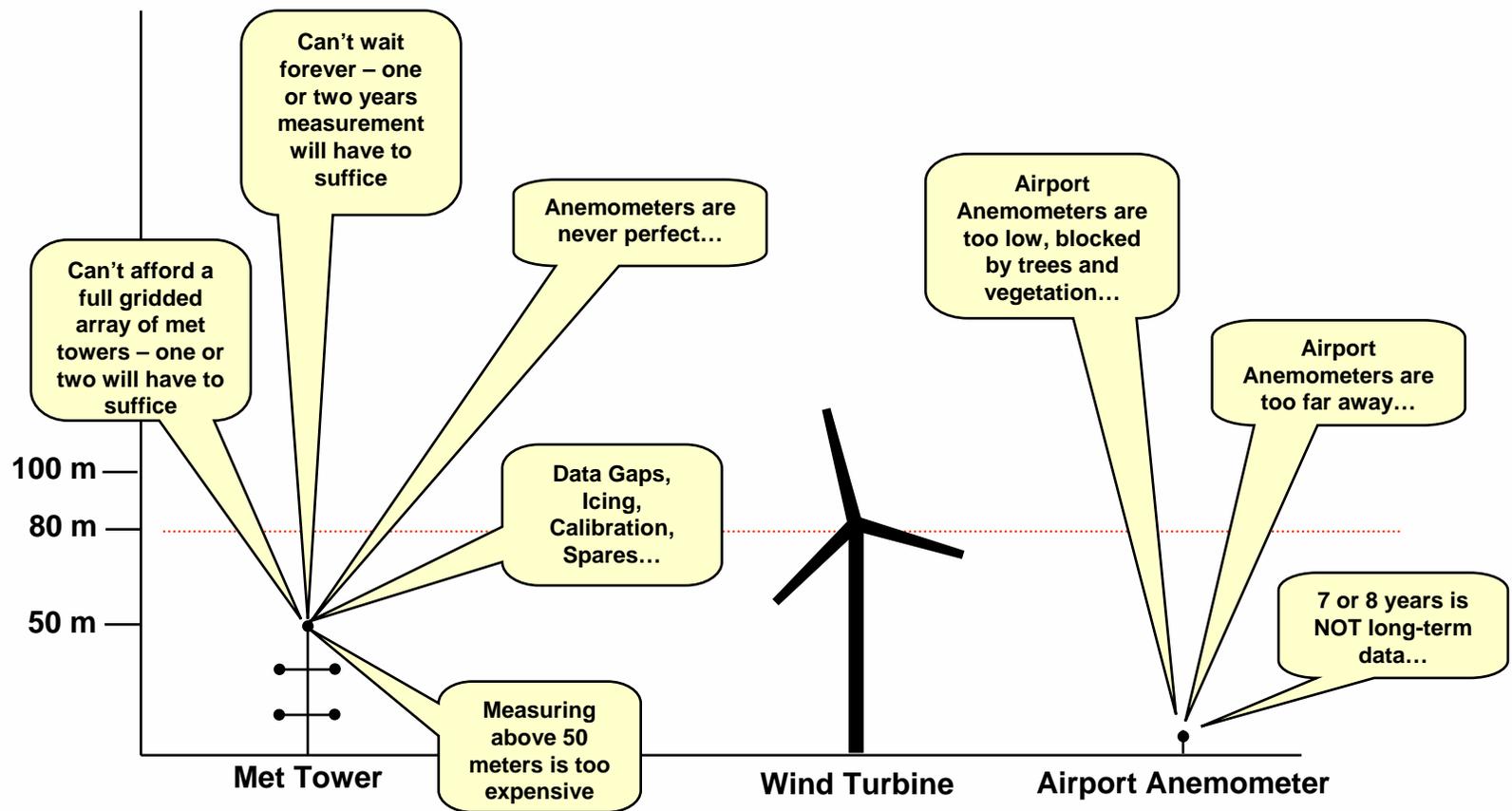
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The Need for Wind Assessment



- Location & terrain make big difference
- Power in the wind is proportional to the cube of wind speed, so great value in optimizing location, layout & height
- Many characteristics to consider
 - Shear (speed increase with height)
 - Diurnal & seasonal patterns
 - Long-term interannual variability
- Major planning and financing issue
 - A large investment with a 25-year timeline
 - Variability on many time scales
 - Implications for operations

Predicting Long-term Project Energy

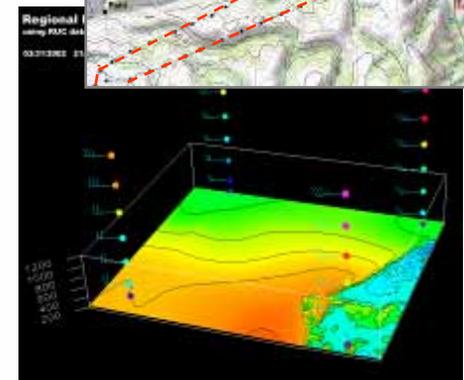
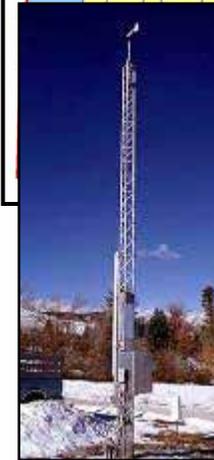


Challenges of distance, height, time & space

Integrated Wind Understanding

Taking advantage of all the available data:

- 1) Use best available “gridded” archives of real weather data from government agencies
 - *Actual recorded weather data, used to initialize weather forecast models*
- 2) Integration of tower data and other on-site measurement points
- 3) Add the best available high-resolution topography and land cover information
- 4) Properly apply meteorological models and wind field models integrating data over space and time
- 5) Analyze long-term variation and the financial impact on your specific situation
- 6) Use wind forecasting to minimize cost and operating impacts & maximize revenues

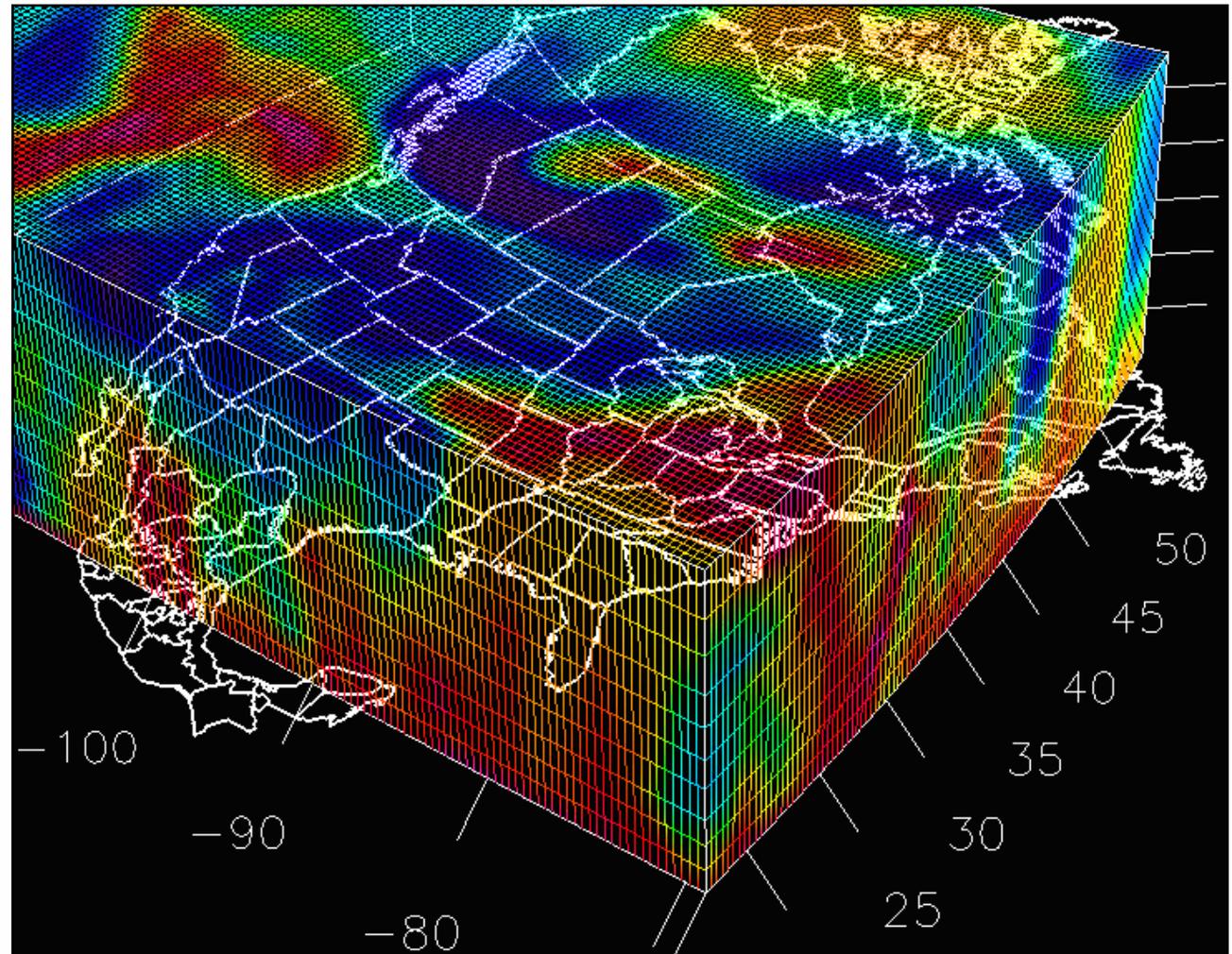


Gridded 3D Weather Data

Integrates all available data sources, from the surface to the upper atmosphere, into a unified and physically consistent state of all grid cells at a given point in time

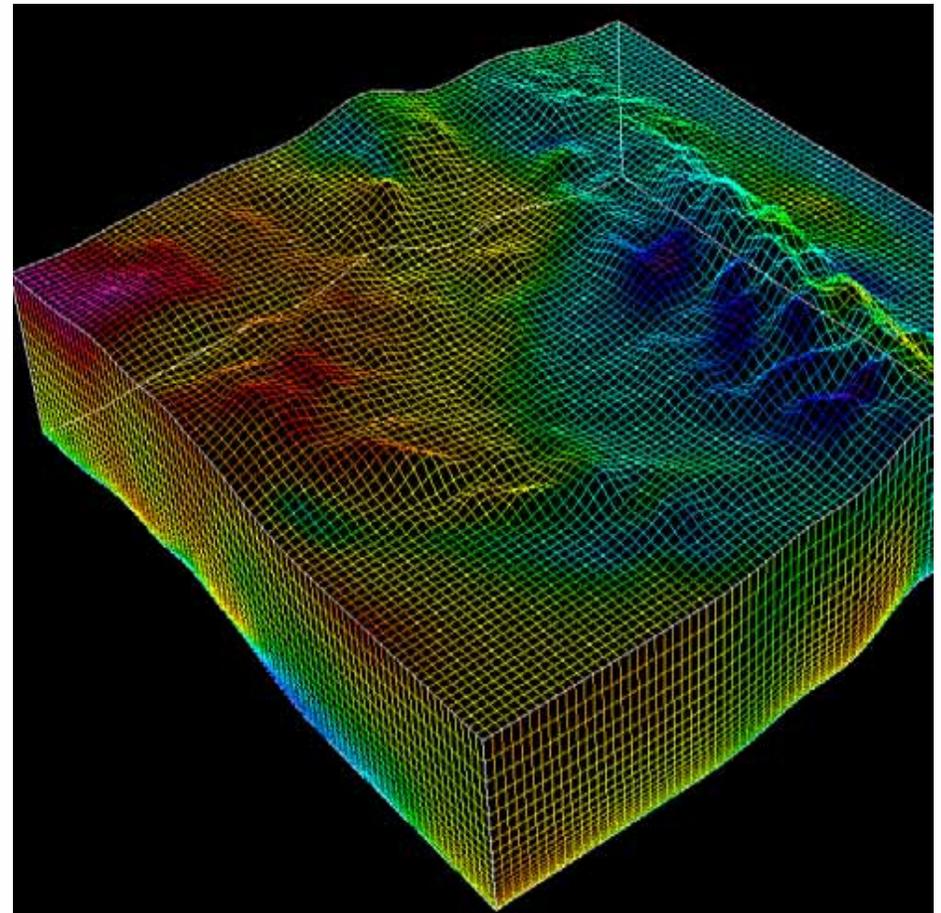
Over 160 weather variables collected from:

- Surface / METAR station data
- Oceanographic buoys
- Ship reports
- Aircraft (over 14,000 ACARS/day)
- NOAA 405 MHz profilers
- Boundary-layer (915 MHz) profilers
- Rawinsondes (balloon soundings)
- Reconnaissance dropwindsonde
- RASS virtual temperatures
- SSM/I precipitable water
- GPS total precipitable water
- GOES precipitable water
- GOES cloud-top pressure
- GOES high-density vis. cloud drift winds
- GOES IR cloud drift winds
- GOES cloud drift winds
- VAD winds from WSR-88D NEXRAD radars



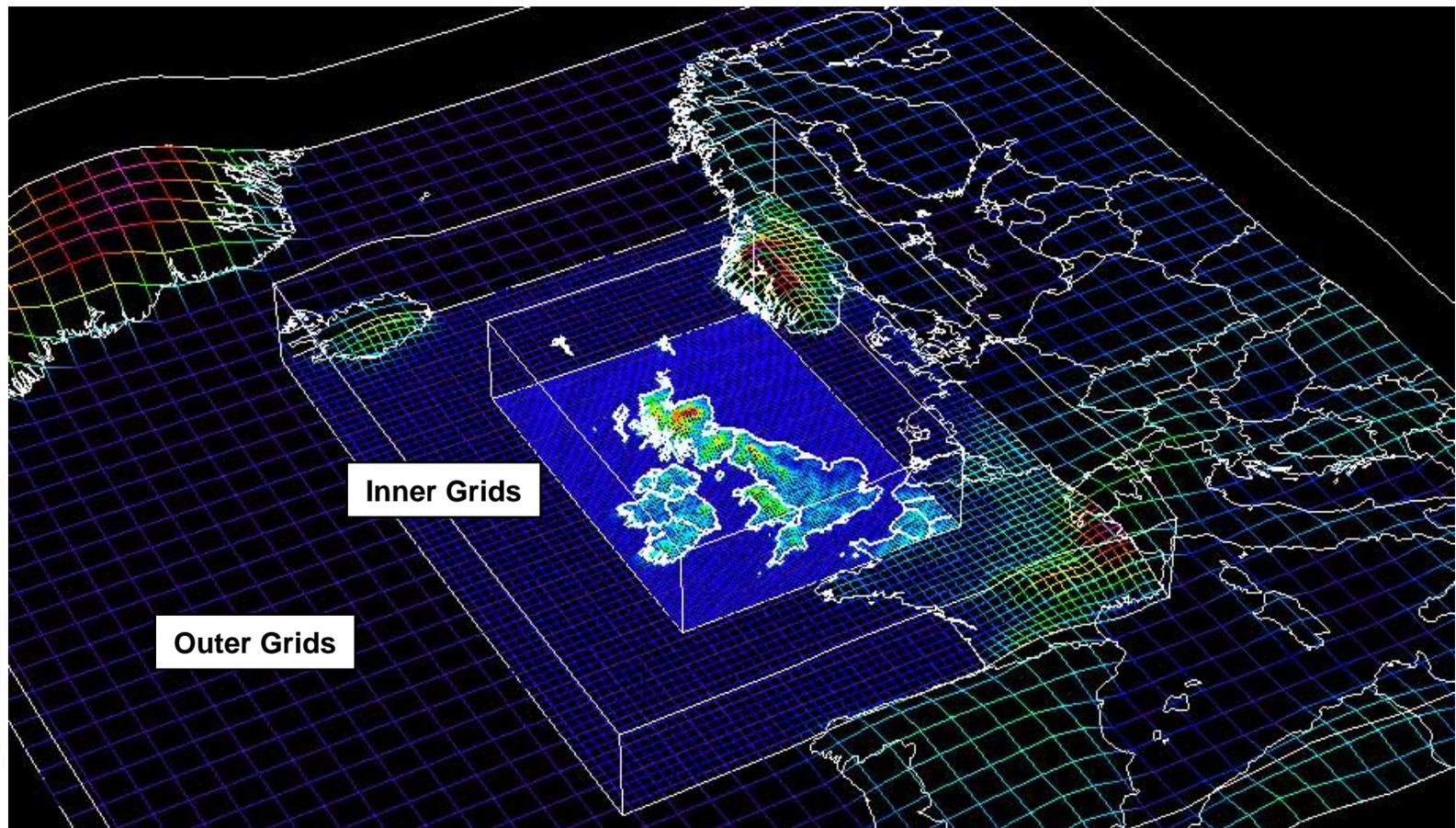
Meteorological Models

- **Numerical gridded representation of the laws of physics**
 - **Conservation relations**
 - Mass
 - Energy
 - Momentum
 - Water, etc.
 - **Physical processes**
 - Radiation
 - Turbulence
 - Soil/ocean interactions, etc.
 - **Use lots of fast computers**
 - Partial differential equations
 - Gridpoint difference values
 - Step all points through time using very small steps (a few seconds per step)

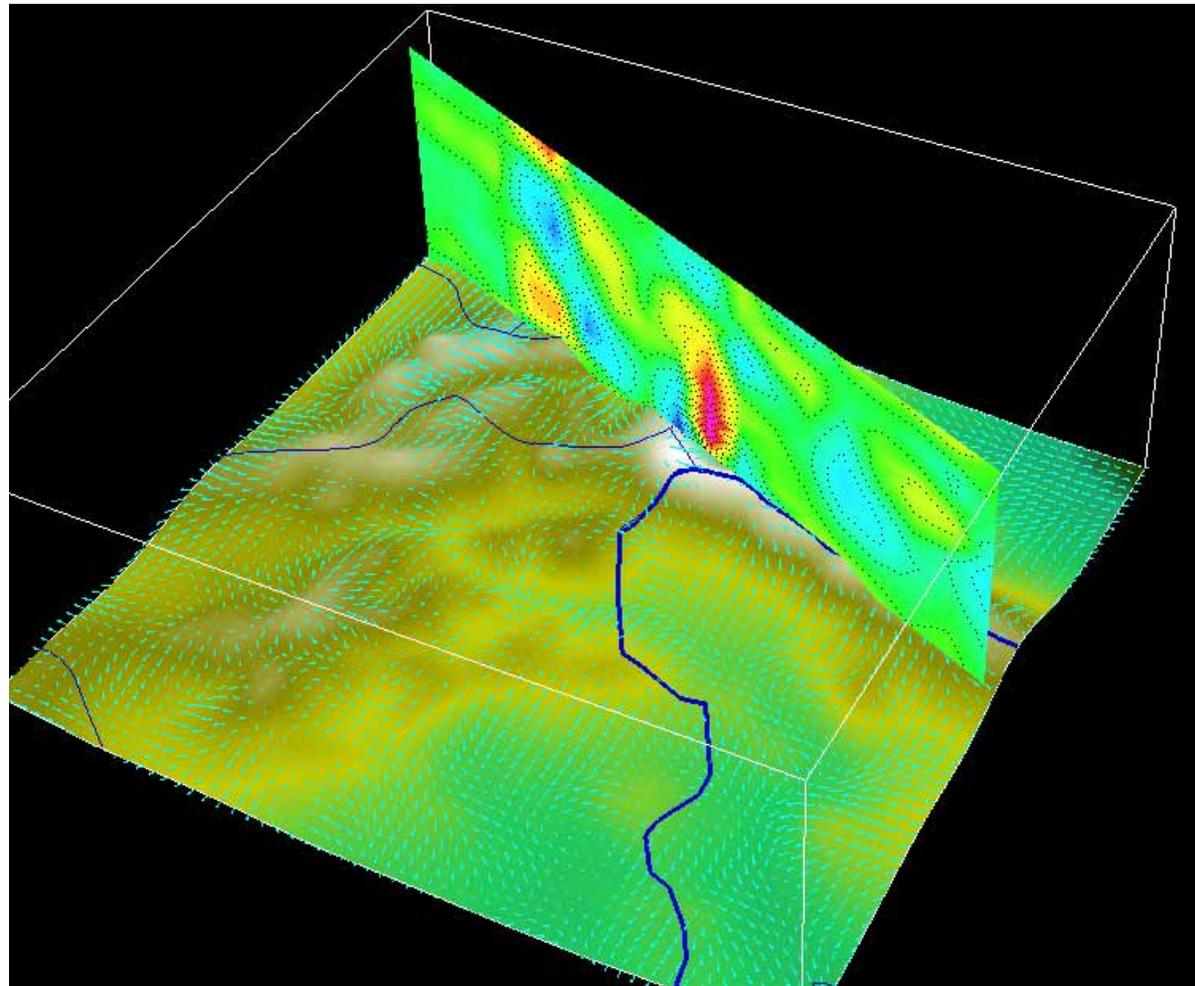


Nesting Modeling Techniques

Modeling “fills the gaps” in both space & time

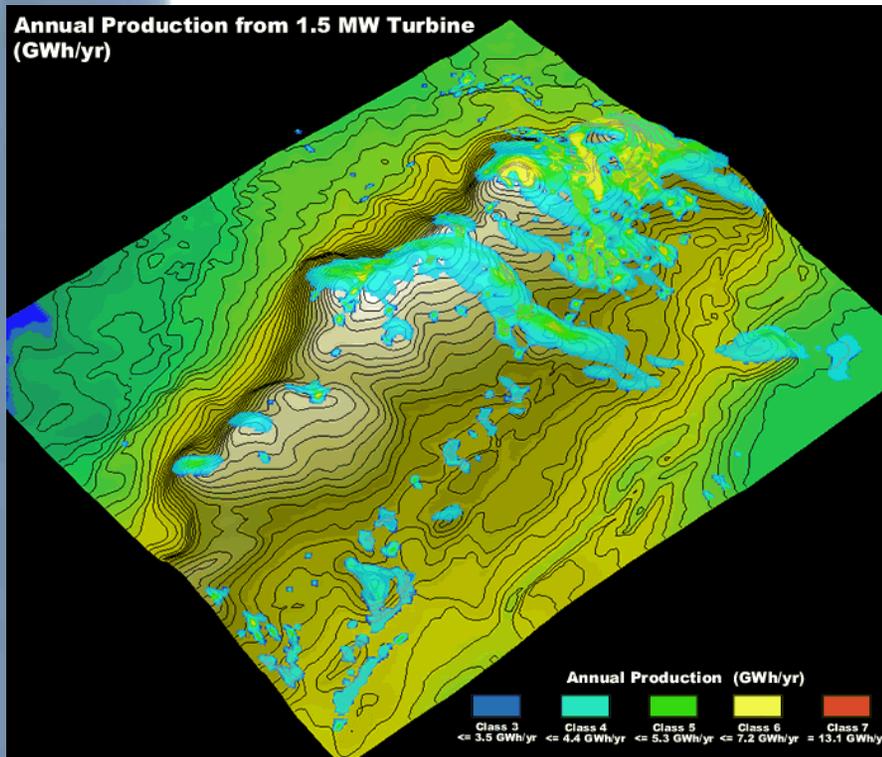


Understanding of the Entire 3D Space

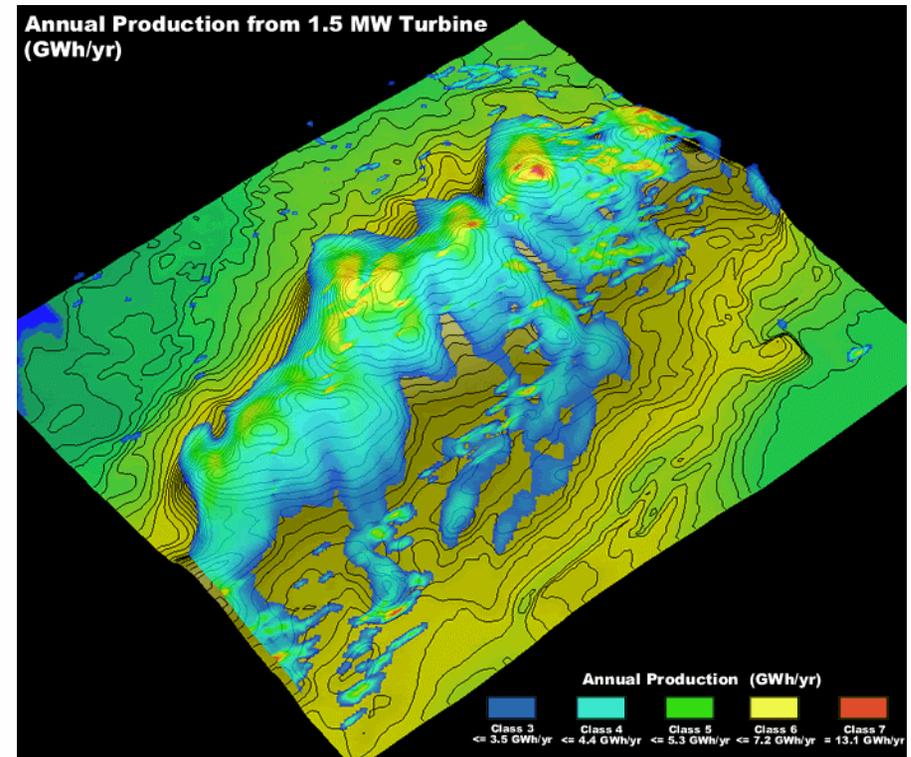


Gross Annual Production

Production calculation in GWh per year - multiple heights & various turbines
Using hour-by-hour wind/density/shear values & all data sources



50m Height



80m Height

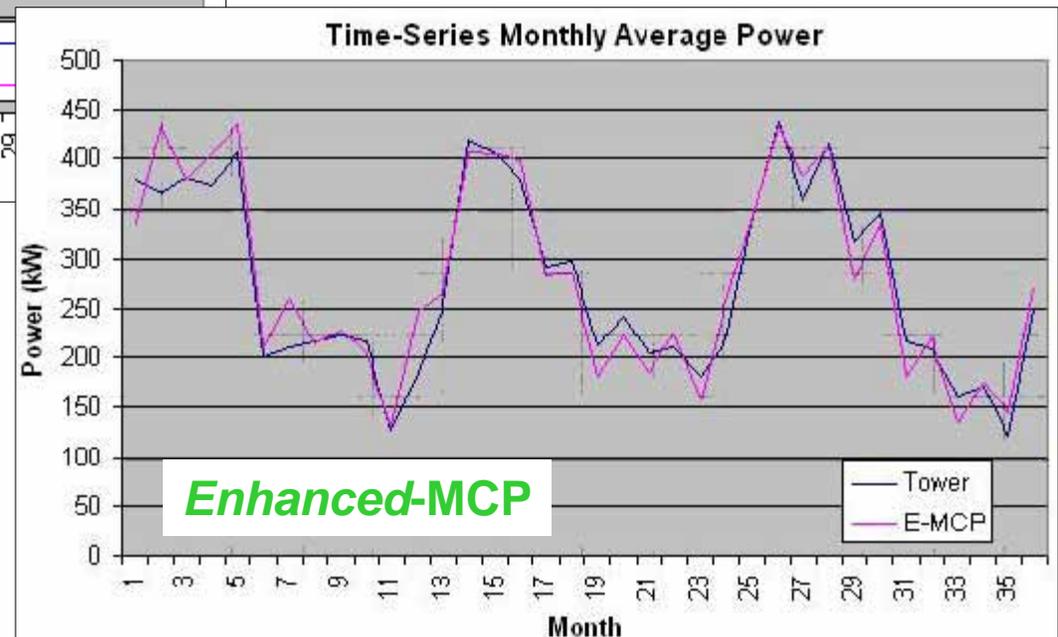
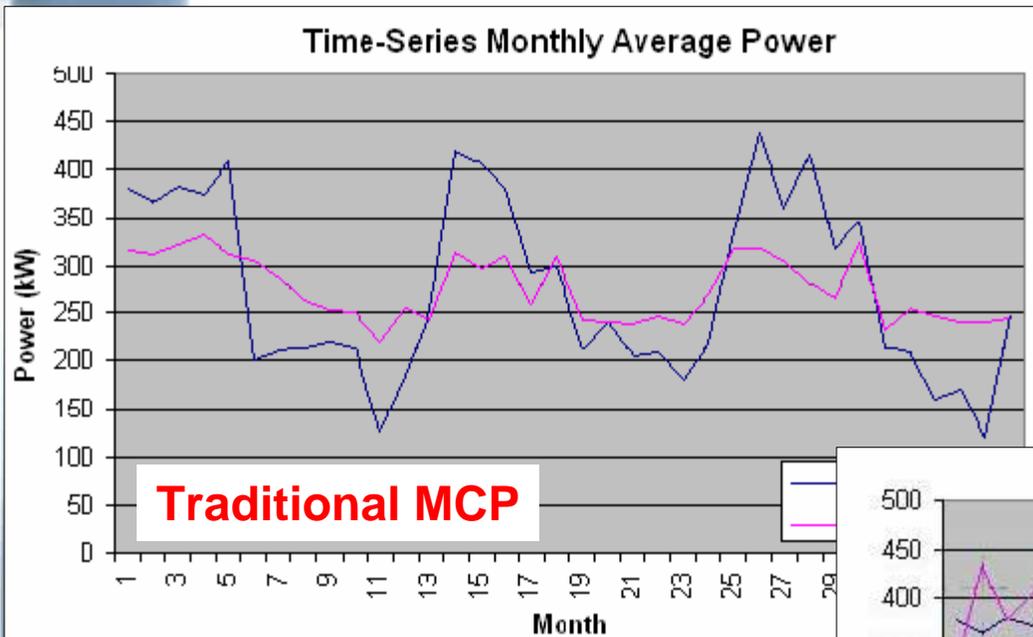
Long-term Variability

Normalized Speed vs. Year

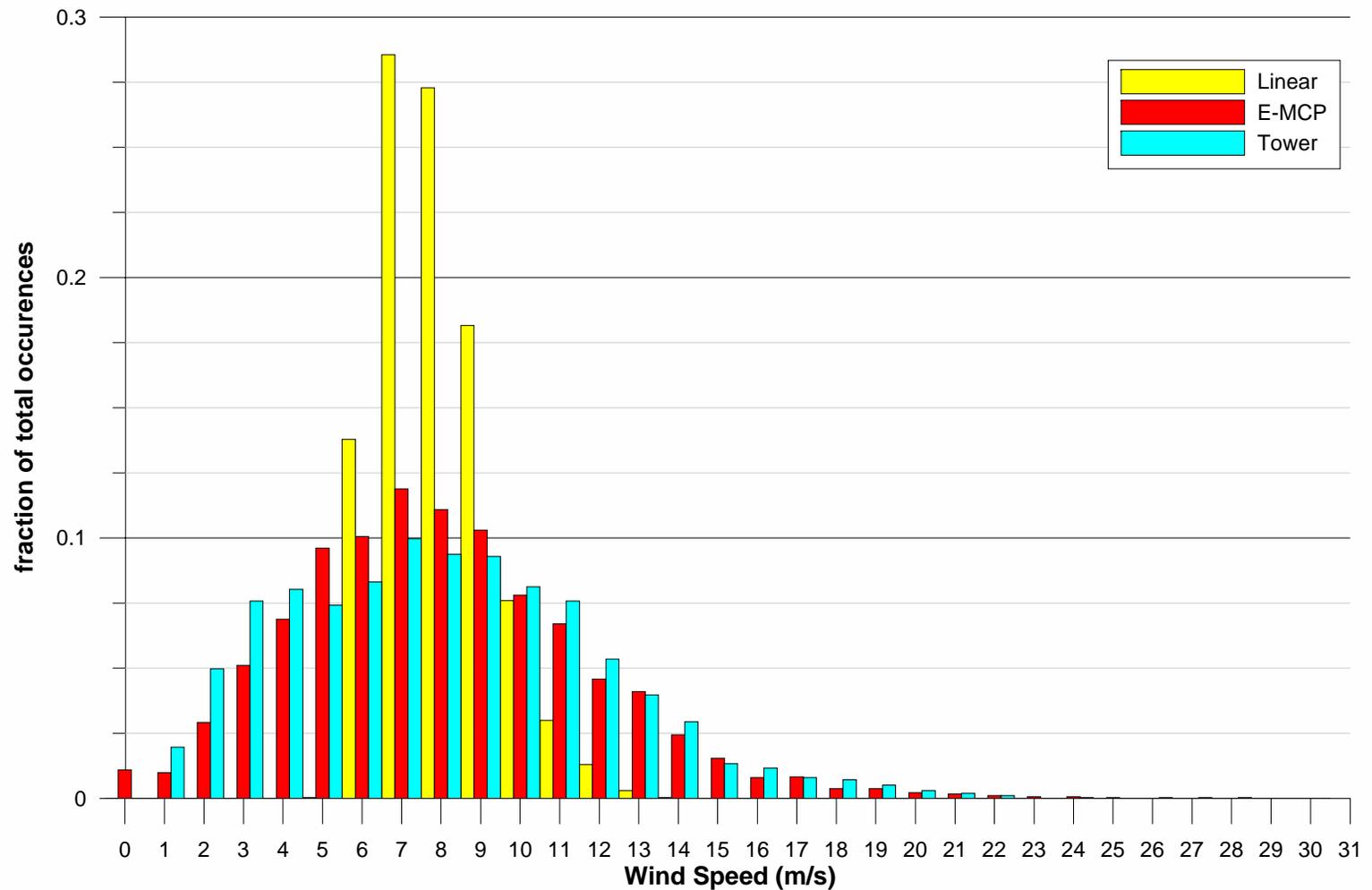


We must extend our understanding to 30 years or more

Examples of MCP Research – Time Series

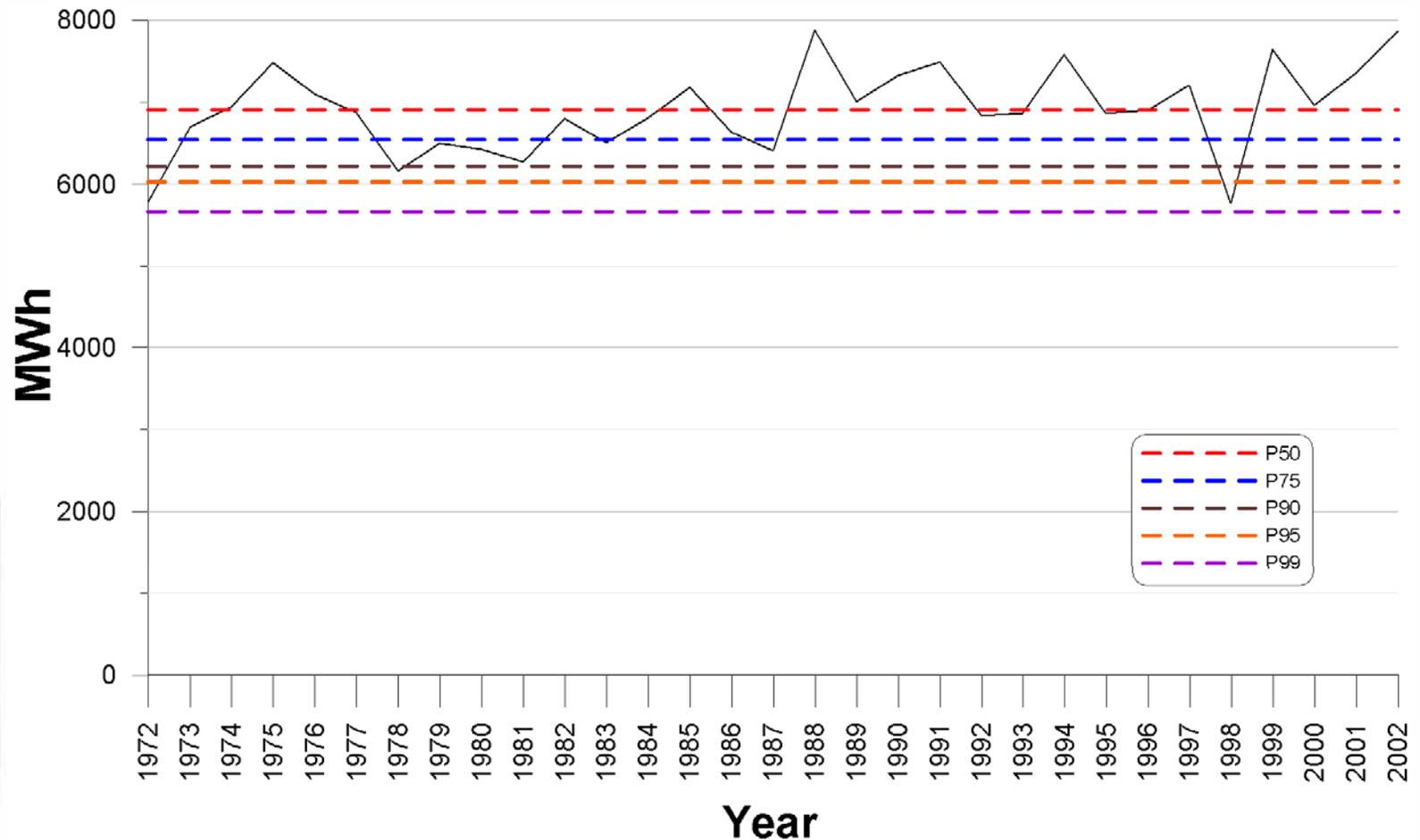


Examples of MCP Research – Distribution



Annual Historical Energy

(Wind Data 1972–2002)



Goal – Accurate Predictive Intervals

Speed	Gross Capacity Factor (Vestas V80 1.8 Class 1A)
7.78 m/s	41.47 %

Table 1. Forty-year mean quantities at 67 m AGL

Speed Standard Deviation (% of period mean)	Gross CF Standard Deviation (Vestas V80 1.8 Class 1A) (% of period mean)
3.95%	7.26%

Table 2. Forty-year standard deviations of annual-average quantities at 67 m AGL

Speed				Gross Capacity Factor (Vestas V80 1.8 Class 1A)			
P75	P90	P95	P99	P75	P90	P95	P99
7.57	7.37	7.26	7.02	39.40	37.50	36.34	34.08

Table 3. Predictive intervals for speed and gross capacity factor at 67 m AGL

Gross Energy to Net Energy

- 1) **Start with a gross energy value (P50, P90, etc.)**
 - Predictive intervals (P90, etc.) are statistically derived directly from the long-term variability of the project site at hub height
 - Predictive intervals are NOT confidence intervals
 - A long-term time series of energy data at hub height, generated with best available data & practices, is a very powerful tool... consider combinations of projects, etc.
- 2) **Subtract other net losses for a net energy value**
- 3) **What about uncertainty & confidence intervals?**

Net Project Losses

- **Other project losses, usually linear in nature and related to the project design and turbines:**
 - **Wake and array losses**.....3-7% typical, less for single row or small project
 - **Turbine availability**.....3% typical, negotiable in turbine agreement
 - **Turbine power curve**.....2% typical, negotiable in turbine agreement
 - **Electrical losses**.....2-3% typical, based on design & current levels
 - **Parasitic/icing losses**.....1-2% typical, site & turbine/technology dependent
- **Other items that apply to some projects:**
 - **Wind sector management**.....if needed on project, perhaps 1%
 - **Substation maintenance/downtime**.....on remote feeders, perhaps ½%
 - **High wind hysteresis**.....for high wind sites, frequent cutouts
- **Bottom line:**
 - **Total net losses of 9% – 15%**
 - **Turbine layout has the biggest impact on this value, and accuracy of wake/array losses depends on accuracy of long-term wind data**

Uncertainty & Confidence Intervals

- **“Uncertainty” is not a statistical term, and real samples for calculating the distributions are usually lacking, but there are other “uncertainties” that may impact financial risk**
 - **Example where we have some samples: anemometer uncertainty**
 - New, calibrated #40 anemometers have relative uncertainty in wind speed of ~1.6%
 - Concerns about time (e.g., bearing wear), installation (booms, vane alignment), data quality (missing data), calibrated vs uncalibrated anemometers, etc.
 - Errors are magnified if used for shear to hub height & power calculations
 - **Examples where site-specific data is limited:**
 - Depending on approach, could also adjust for “uncertainty” in the methods, data or model for power curve, wake losses, hub height values, long-term normalization, etc.
 - It is very difficult to calculate actual values, so rules of thumb are traditionally used
- **Investing in best practices will allow the bankers to use modest uncertainty assumptions and to give you financing at better values**

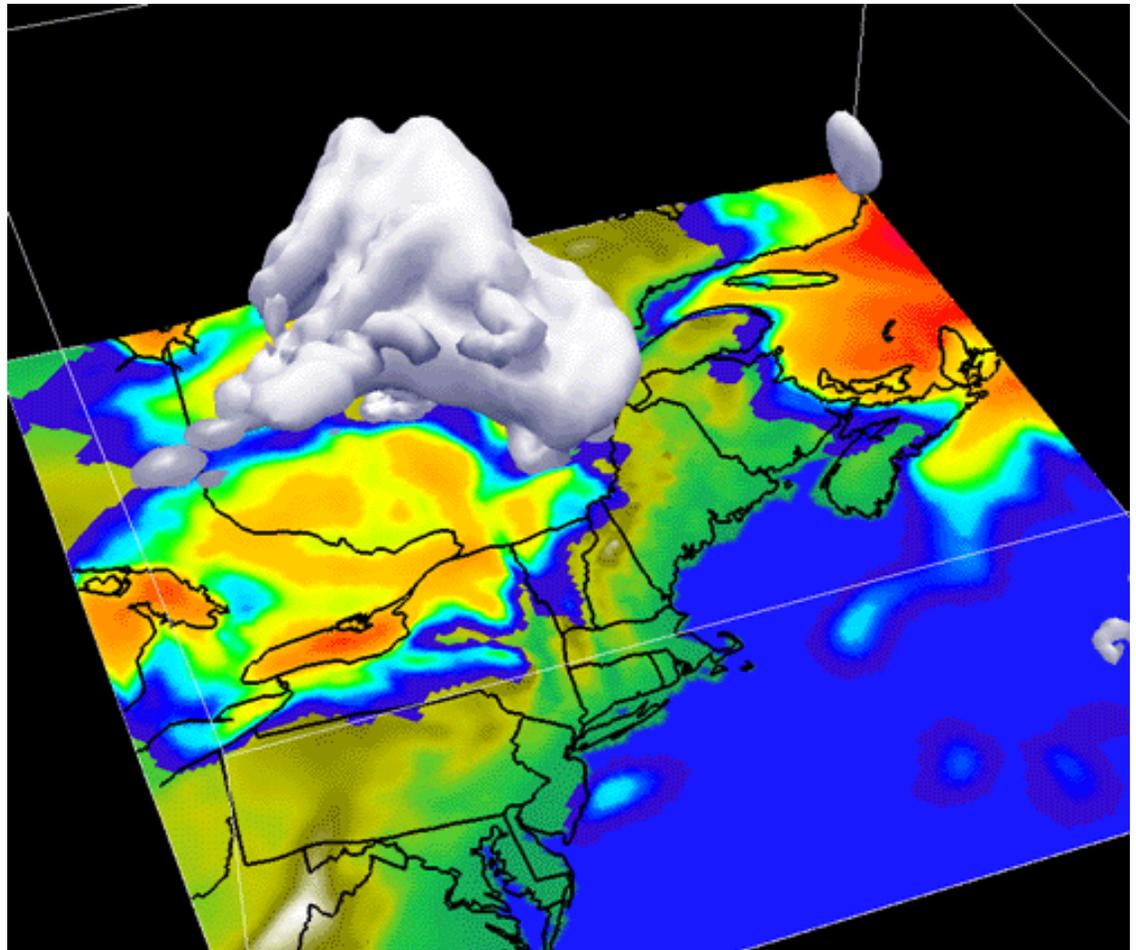
Financial Risk is the Bottom Line

- Improved understanding of wind resource and variability has great value
- Need integrating approaches that unify all data and reduce the risk inherent in single-point data sources
 - Meteorological Towers
 - Modeling
 - Statistics
- The goals:
 - Accurate results based on scientific and statistical methods
 - Consistent ways to view risk and compare projects
 - Take wind plants into the mainstream of power generation projects in terms of process, financing, operations and utility perception

WindLogics Inc.

Time series showing
forecast with wind speed
and cloud cover

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