#### Workshop "Wind Power and Market design" -University Paris XI

# Portfolio analysis and wind power

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## Outline

### • Wind power variability and geographic diversity

- Time scale: Hourly variability, seasonality, yearly variability
- Unit commitment/balancing effects/Capacity credit

#### Applying Mean Variance Portfolio theory to wind power

- Physical electricity output vs. financial analysis
- Social planner (National or EU level) vs. investor project mix

#### • Data and preliminary results

– Case study for Austria, Germany, Spain and Denmark

#### • Next steps...

 Taking into account system (transmission constraints) and interactions between load and wind output

# Introduction: EU wind resource and investment planning



Sheltere	Sheltered terrain <sup>2</sup>		Open plain <sup>3</sup>		At a sea coast <sup>4</sup>		Open sea <sup>5</sup>		Hills and ridges <sup>6</sup>	
m s <sup>-1</sup>	$Wm^{-2}$	$m s^{-1}$	$Wm^{-2}$	$m s^{-1}$	$Wm^{-2}$	${ m ms^{-1}}$	$Wm^{-2}$	$m s^{-1}$	$Wm^{-2}$	
> 6.0	> 250	> 7.5	> 500	> 8.5	> 700	> 9.0	> 800	> 11.5	> 1800	
5.0-6.0	150-250	6.5-7.5	300-500	7.0-8.5	400-700	8.0-9.0	600-800	10.0-11.5	1200-1800	
4.5-5.0	100-150	5.5-6.5	200-300	6.0-7.0	250-400	7.0-8.0	400-600	8.5-10.0	700-1200	
3.5-4.5	50-100	4.5-5.5	100-200	5.0-6.0	150-250	5.5-7.0	200-400	7.0- 8.5	400- 700	
< 3.5	< 50	< 4.5	< 100	< 5.0	< 150	< 5.5	< 200	< 7.0	< 400	

- National focus on sites with best wind resource
- As penetration increases, investment planning needs to take into account:
  - wind variability and interaction with other technologies (Neuhoff et al., 2008)
  - Integration into power network
  - Transmission constraints, particularly at EU level
  - Correlation between sites to diversify risk => portfolio approach

## Wind power output correlation decreases with distance between sites



UK wind turbine output correlation decreases by about 0.1 every 100/200 km.

# Portfolio analysis of wind power development

- Wind power variability and geographic diversity
  - Diversify *physical output risk*
  - Diversify *financial risk exposure* more complex: related to local market (balancing mechanism) and regulatory (support scheme: feed in tariff or certificates) factors.
  - => Focus on physical output risk
- **Geographic dispersion at which level**? Portfolio approach can help to reduce variability and risk:
  - At the **project** level, within a company's assets portfolio
  - At the **country** level, within a national network system
  - At the **EU** level, countries have particular wind patterns

## Wind capacity additions in the EU



- **Static** perspective: what are the optimal country portfolios?
- **Dynamic** perspective: Are there appropriate incentives for wind power development across the EU countries?

# Portfolio analysis of wind power development (2)

#### • Diversification over which time scale?

- Hourly variability, seasonality, yearly variability
- Unit commitment/balancing effects/Capacity credit
- Support schemes introduce an additional dimension (Feed in tariffs vs. green certificates, etc.)

#### • Type of analysis is key to define time scale:

- 1. Investment planning from a "social" perspective: Optimal portfolios based on hourly consideration on national level have lower risk for unit commitment/balancing or capacity credit.
- $\Rightarrow$  detailed modeling of transmission capacity, market integration balancing and day ahead, etc.
- 2. Investment projects from an investor perspective: construct portfolios that minimize quantity risk / maximize return.
- $\Rightarrow$  Yearly analysis for long term contracts; monthly analysis for medium term contracts and hourly analysis for Day ahead/balancing trade.

## Quantifying the optimal degree of diversity

- The extent to which diversity is to be pursued depends on the balance between the *extra costs* and the degree of *risk reduction* achieved.
- Various methods have been developed to quantify and optimise the diversity of a portfolio of assets:

#### - Value at Risk

• The Value at Risk (VAR) calculates the maximum loss expected (or worst case scenario) on an investment, over a given time period and given a specified degree of confidence.

#### - Markowitz Mean Variance Portfolio theory

• The Mean-variance portfolio theory (MVP) defines efficient portfolios as the ones which have the smallest attainable portfolio risk for a given level of expected return (or the largest expected return for a given level of risk).

## The portfolio effect – The case of a two-asset portfolio

- For two assets (X1, X2) with respective returns (r1, r2) and standard deviation (σ<sub>1</sub>, σ<sub>2</sub>):
  - Portfolio return:
  - Portfolio variance:
- Efficiency frontier:



$$\sigma_{p} = \sqrt{X_{1}^{2}\sigma_{1}^{2} + X_{2}^{2}\sigma_{2}^{2} + 2X_{1}X_{2}\rho_{12}\sigma_{1}\sigma_{2}}$$





Some amount of diversification occurs whenever the returns of two (or more) securities are less than perfectly correlated (i.e.  $\rho < 1.0$ )

## **Portfolio theory efficient frontier**



- The efficient frontier for a portfolio of two risky assets.
- MVP theory does not prescribe a single optimal portfolio combination, but a **range of efficient choices**.
- Investors will choose a risk-return combination based on their **own preferences and risk aversion**.

## Literature review - Wind power and Portfolio Analysis

- Portfolio analysis that consider the effect of wind power in a conventional electricity generation portfolio (gas, coal, nuclear, etc.)
  - DeLaquil P. et al. (2005), McLoughlin and Bazilian (2006), Kienzle et al (2007), Awerbuch and Berger (2003), Twomey (2005), etc...

#### • Geographical or spatial effects of wind power

- Correlation analysis: Sinden (2007), Hirst (2002), Giebel (2000), etc.
- Porfolio analysis:
  - Drake and Hubacek (2007)
  - Kyle Datta E. and Hansen L. (2005)
  - Hansen L. (2005)
- Other effects (e.g. power network, demand)
  - Drake and Hubacek (2007) take into account transmission losses.
  - Sinden (2007) takes into account the correlation between wind power and demand

## Optimisation of EU power generation mix – Awerbuch and Yang (2005)



2020 EU Baseline Portfolio Optimization (CO2=E35/tonne) – Source: Awerbuch and Ynag (2005)

#### 2020 EU Baseline Portfolio Optimization indicates that renewables can reduce risk and cost

# Applying portfolio theory to geographical dispersion

- The key point: wind speed correlations between different wind farms
  - Focus can be on *physical output risk; or*
  - on investment project financial risk exposure
- Holding period return defined as in finance:
  - Physical output: (Pt Pt-1)/Pt
  - Financial return: project NPV, or variation of cash flows/generation cost

#### • Constructing the efficiency frontier:

- Data on average wind power generation, standard deviations and correlation coefficients
- Optimization model to compute minimum standard deviation (portfolio risk) that exists for any given rate of average power generation (portfolio return) that is input into the model

## The database

#### • Type of data:

- Real production data (Hirst 2002)
- Simulated data from wind speed data (Hansen 2005, Kyle Datta and Hansen 2005, Drake and Hubacek 2007, Sinden 2007)

#### • Data resolution:

- Hourly (Drake and Hubacek (2007) and Sinden (2007))
- $\Rightarrow$  How many years are necessary to have reliable data?
- $\Rightarrow$  How does « geographical aggregation » of data impact results?
- $\Rightarrow$  Which data resolution/filtering for what type of analysis?

#### • Our database: aggregated hourly wind production data:

- Spain (from 2002 to 2007)
- Germany by TSO zone (from 2006 to 2008)
- Austria (from 2006 to 2007)
- Denmark by zone (from 2000 to 2008)

#### • We are waiting for:

- French production data (from 2006 to 2007)
- Wind speed data for several European Countries

## **Preliminary results**

- Results based on hourly wind production data (2006 –2007) for Spain, Germany, Austria, and Denmark
- Outputs:
  - Wind Capacity Factor variability (Sinden 2007)
  - Interaction of wind production and demand (Sinden 2007)

#### - Portfolio Analysis

- Wind output
  - Hourly analysis
  - Monthly analysis
- Wind output and demand

## Intern-annual variability Average capacity factor



- Capacity factors computed dividing hourly wind power production by installed capacity
- Assumption that installed capacity changes linearly during the year

## Intern-annual variability Standard deviation of capacity factor



- Hourly standard deviation varies significantly year to year...
- Smaller countries have less dispersed wind farms => higher standard deviation



- Patterns significantly different across countries
- Spain has much less seasonal variability

## Wind Power and demand Monthly comparison



## Wind Power and demand Hourly comparison





## Portfolio analysis (à la Hansen 2005) Wind power only (Hourly analysis)

#### **Correlations analysis**

Correlation						
	Spain	Germany	Austria	Denmark		
Spain	1,0000000	0,0970128	0,1516223	0,0235352		
Germany		1,000000	0,2473075	0,7092889		
Austria			1,000000	0,1045161		
Denmark				1,000000		

- Social planner optimisation, assuming:
  - only four countries
  - no physical constraints (integrated markets)
  - no limit in wind potential
  - that capacity factors by country are geographically consistent

#### **Optimal wind portfolios for Spain and Denmark** (weakest correlation = 0.0235) – Only Wind Power



### **Optimal portfolios for Spain-Germany-Austria-Denmark**

				Weight	
Mean	Risk	Weight Spain	Weight Germany	Austria	Weight Denmark
0,2233	0,107	53,99%	20,99%	14,87%	10,15%
0,2254	0,108	54,79%	16,11%	15,64%	13,46%
0,2274	0,108	55,58%	11,22%	16,42%	16,78%
0,2295	0,109	56,37%	6,34%	17,20%	20,09%
0,2316	0,11	57,16%	1,46%	17,98%	23,41%
0,2337	0,115	49,46%	0,00%	14,84%	35,71%
0,2358	0,131	38,14%	0,00%	10,04%	51,82%
0,2378	0,156	26,83%	0,00%	5,24%	67,94%
0,2399	0,185	15,51%	0,00%	0,44%	84,05%
0,242	0,218	0,00%	0,00%	0,00%	100,00%

## Efficient frontier for wind portfolios in Spain-Germany-Austria-Denmark



- Current portfolio is far from efficient frontier
- Things will improve in the future as weight of Germany decreases

# Wind Power and demand "Net demand" variability

#### • Correlations between

- Demands
- Wind power productions
- Net demands?

#### • Methodology:

- To construct data series for hourly wind capacity factor by country "i" (this capacity factor is the production for 1 MW installed in each country assuming that these aggregated capacity factors are representative for the whole country) --> WCFi
- To construct a data series for hourly total demand factor (this represents hourly demand in terms of global installed capacity or peak load) --> TDF
- To construct data series for hourly "needed generation capacity" as NCi = WCFi – TDF. Then, computing efficient frontier for NCi and determine "efficient portfolio"

# Wind Power and demand « Net demand » variability



- This "net demand approach" is not so realistic (because we are not considering wind potential, transmission constraints and lack of market integration).
- It could become more interesting if we can include these issues.

# **Summary of results**



 Correlation between each country wind capacity factor Spain Germany Austria Denmark and total demand: 0,04770768 0,13084032 0,07255191 0,15446145

# Conclusions

#### • As wind power penetration increases:

- Focus shifts from best sites towards optimization of utilities / countries portfolios
- Correlation between wind sites is key, but also correlation with load and other power production technologies
- Portfolio theory is a powerful tool to optimize wind portfolios at different geographical levels...
  - Risk-reward tradeoffs of utilities investments
  - Social planners for deployment support policy
- ...But realistic analysis requires to take into account dispatching and transmission constraints...

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### Thank you for your attention!

### **Comments much welcome!**

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# **Next steps**

- Data
  - Type of data (we will have more real wind power generation)
  - Scaling data?
  - What to do with data for different years ?
    - Construct probability distribution and simulate wind power generation?
  - Is it possible to combine different type of data (actual generation and simulated generation with wind speed)?
    - Under which assumptions
  - Costs & support scheme Data  $\rightarrow$  \$ porfolio analysis
- Computations to do
  - Scenarios

# With average monthly data – wind only



- Same mean but different variance changes optimal portfolios
- More comment?

# With average monthly data – wind only



- Same mean but different variance changes optimal portfolios
- More comment?

Mean	Mean Risk Weight		Weight Germany	Weight Austria	Weight Denmark	
0,2284	0,054	63,34%	2,96%	33,00%	0,71%	
0,23	0,054	63,82%	0,00%	31,44%	4,75%	
0,2315	0,055	65,53%	0,00%	17,62%	16,85%	
0,233	0,057	67,24%	0,00%	3,80%	28,96%	
0,2345	0,061	59,13%	0,00%	0,00%	40,87%	
0,236	0,067	47,31%	0,00%	0,00%	52,69%	
0,2376	0,074	35,48%	0,00%	0,00%	64,52%	
0,2391	0,082	23,65%	0,00%	0,00%	76,35%	
0,2406	0,091	11,83%	0,00%	0,00%	88,17%	
0,2421	0,1	0,00%	0,00%	0,00%	100,00%	











# **Summary of results**

	Spain	Germany	Austria	Denmark
Min Return/Risk (only wind)	53,99%	20,99%	14,87%	10,15%
Medium Return&Risk (only wind)	57,16%	1,46%	17,98%	23,41%
Max Return/Risk (only wind)	0,00%	0,00%	0,00%	100,00%
Min Return/Risk (wind&demand)	49,04%	19,30%	15,43%	16,23%
Medium Return&Risk (wind&demand)	51,97%	1,19%	18,31%	28,52%
Max Return/Risk (wind&demand)	0,00%	0,00%	0,00%	100,00%

• Comment here



## Efficient frontier for wind portfolios in Spain-Germany-Austria-Denmark

