
Hydro Scheduling Powered by Derivatives

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Summary

- An empirical analysis of how commodity storage is operated for 13 hydropower producers
- Testing different hypotheses on inventory and operational policies
- Our results indicate:
 - A simple regression model can explain a significant part of the variation in the scheduling policies
 - Electricity forward prices are used in the optimization of hydro scheduling
 - Real option theory applies: The higher the price volatility, the lower the production

Outline

- Related literature
 - Nordic electricity market
 - Hydropower scheduling
 - Empirical analysis
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Related literature

- Theory of storage: Telser (1958), Williams & Wright (1991), Deaton & Laroque (1996)
 - Hydropower scheduling
 - Many OR and engineering papers on methods, including stochastic programming: Wallace & Fleten (SP handbook, 2003)
 - Some econ papers, e.g., Førsund (2007)
 - Only few empirical studies. For instance, Tipping (2006) and Nasakkala & Keppo (2007)
 - Related OR papers: Ding, Dong & Kouvelis (OR 2007), Caldentey & Haugh (MOR 2006), Birge (2006)
 - Imply that financial information should be used
 - Empirical studies on nonfinancial firms: Guay & Kothari (JFE 2003), Bartram, Brown & Fehle (2006)
 - Nonfinancial firms don't trade much derivatives
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Nordic electricity market

- All the time supply equals demand
 - National grid companies manage short term imbalances
- Spot market
 - Daily submission of supply and demand bids for the next 12-36 hours
- Forwards and futures
 - Traded on Nord Pool (exchange) and OTC/bilaterally

Electricity derivatives market

- Underlying asset

- Elspot system price which is the average price of physical electricity in the whole Nord Pool area over the next 12-36 hours and calculated assuming no transmission bottlenecks

- Futures

- Exchange-traded contract for delivery in a specified future time interval at an agreed price
- Financially settled mark-to-market, week and month maturity lengths

- Forwards

- Financially settled during maturity period, quarters and years maturity lengths, up to five years into the future
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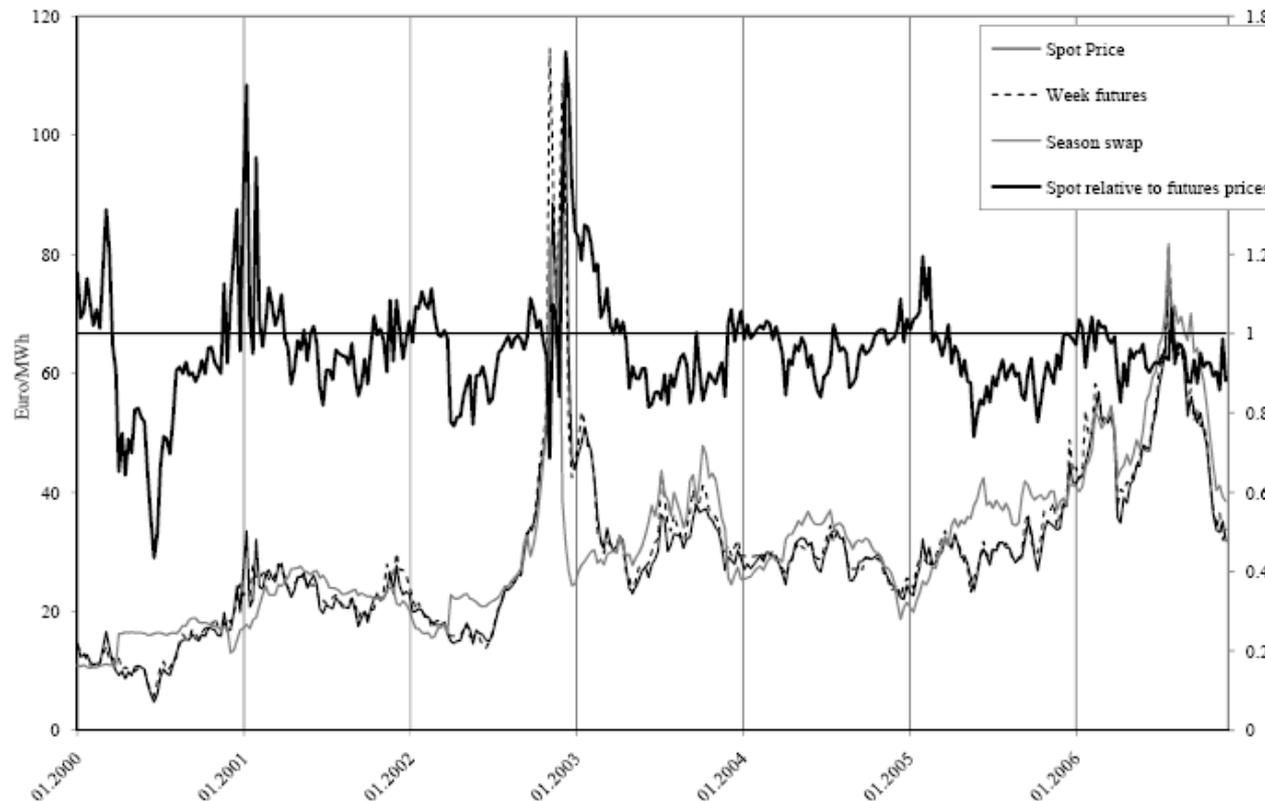
Nord Pool prices

- Descriptive statistics for spot prices, weekly futures, seasonal forwards, and spot price relative to the futures prices. All prices are in Euro/MWh. ADF is the Augmented-Dickey-Fuller stationary test statistic which has a critical value of -2.87 at a 5% significance level.
- An average of 0.96 indicates that forward prices above the spot price, i.e., risk premium

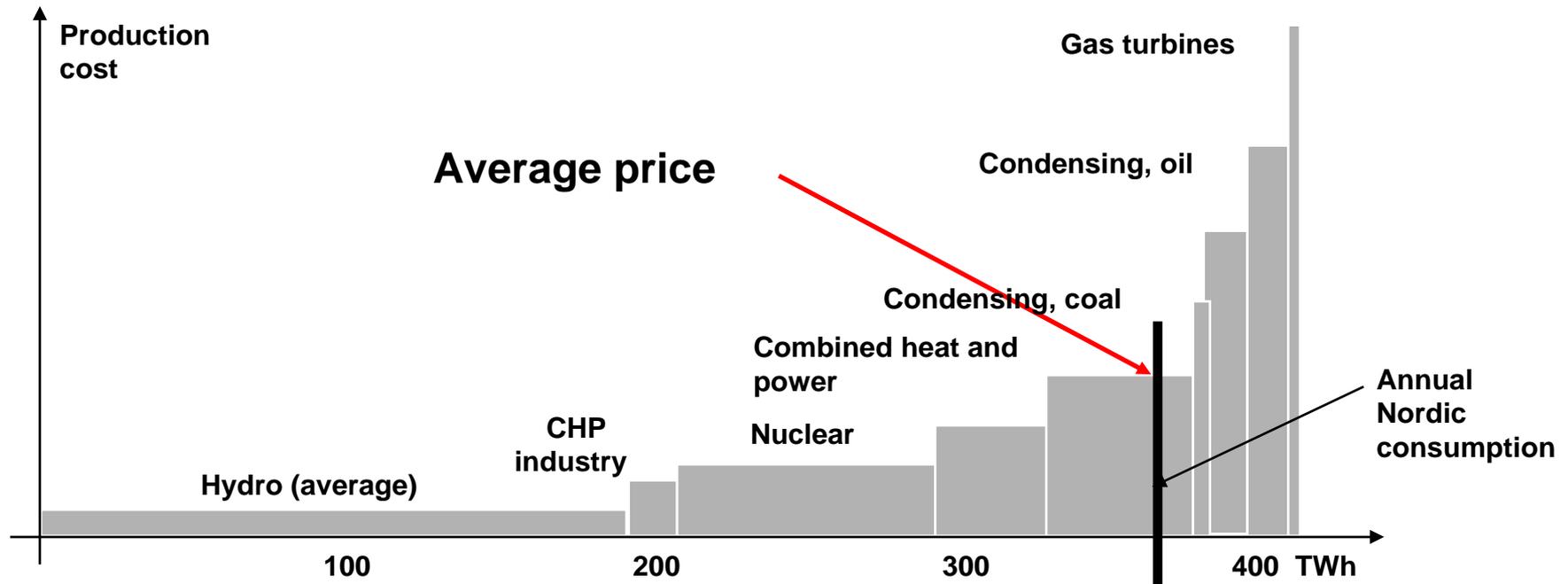
	<i>Mean</i>	<i>Min</i>	<i>Max</i>	<i>Std. dev</i>	<i>ADF</i>
Spot Price	29.63	4.78	103.65	14.01	-2.928
Weekly futures	30.44	5.70	114.56	14.89	-3.446
Seasonal forwards	31.16	10.48	83.25	13.56	-2.890
Spot relative to futures prices	0.958	0.435	1.71	0.136	

Nord Pool prices, Cont'd

- Spot and selected futures and forward prices between February 2000 and December 2006.
- Timing matters!



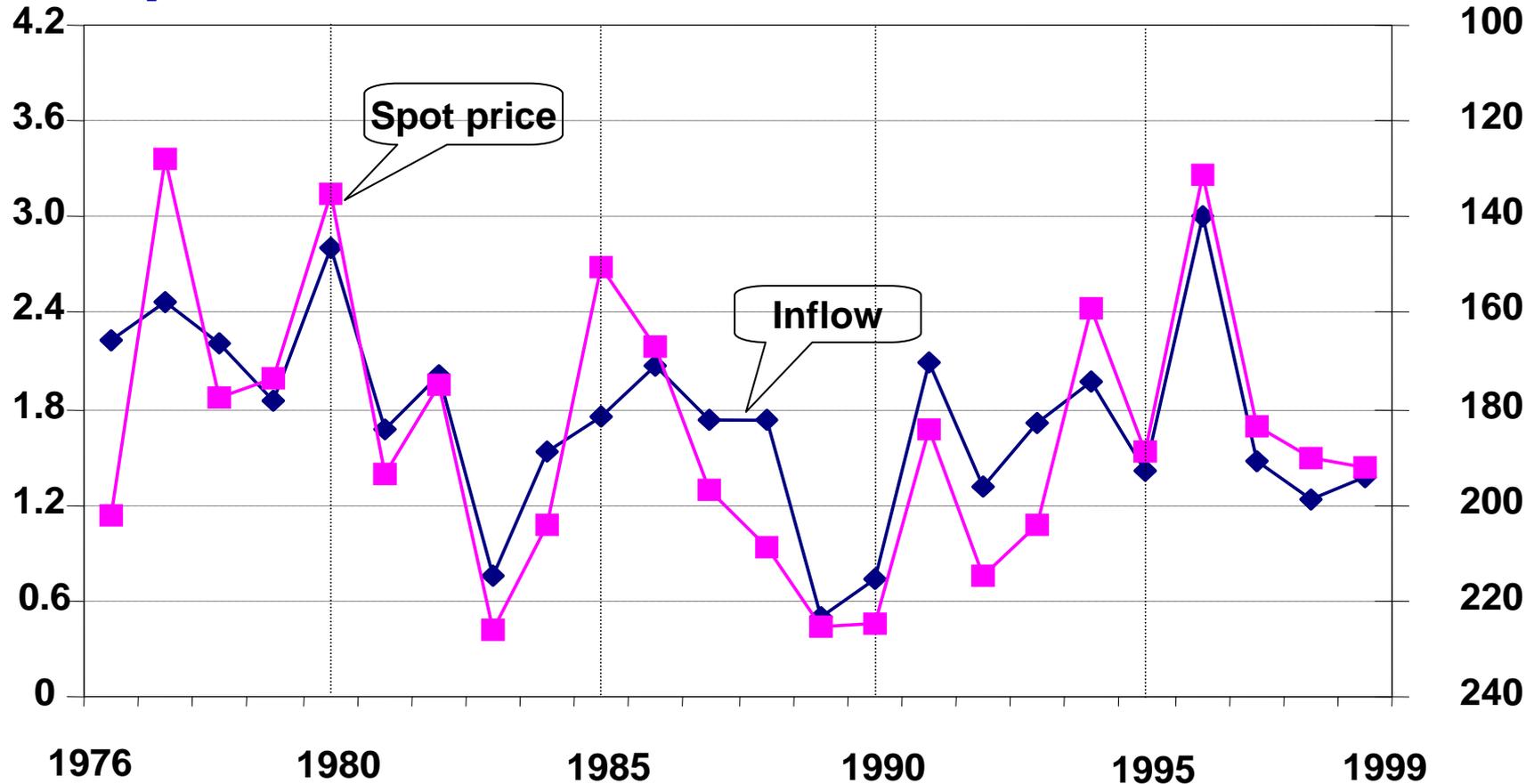
Supply curve



Key characteristic: Inflow uncertainty

Spot price *)
[€cent/kWh]

Inflow **)
[TWh]

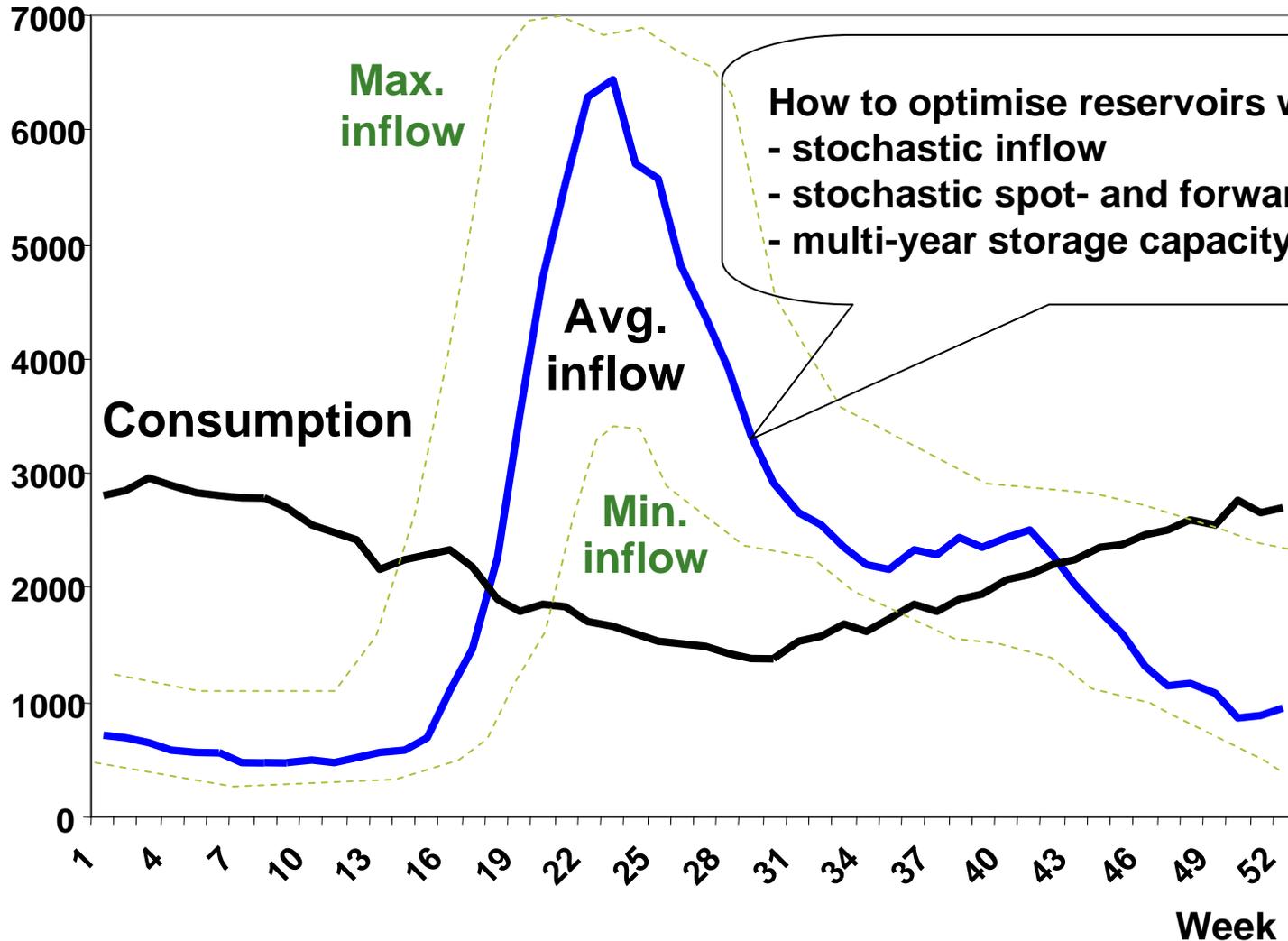


*) Average spot price in 1999-prices

**) Annual inflow Norway and Sweden

Inflow and hydro scheduling

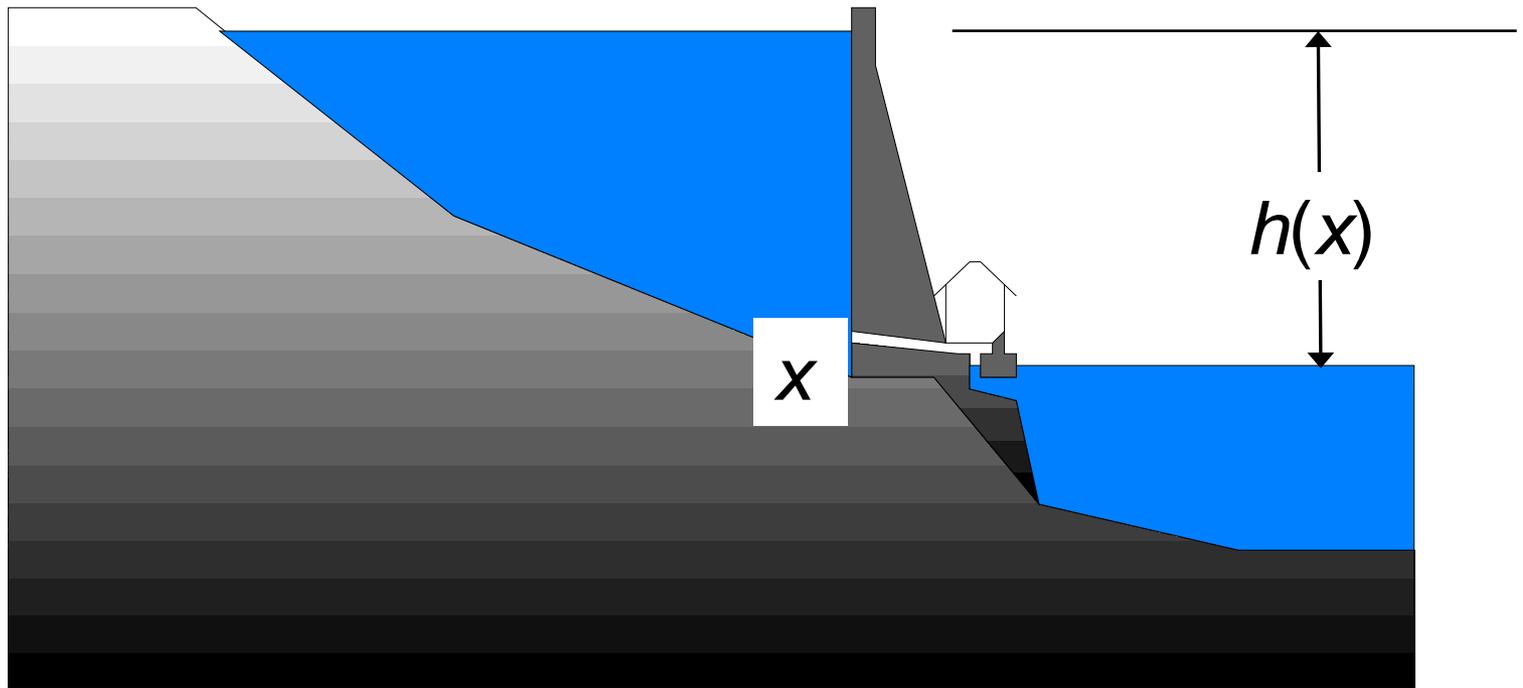
GWh/week



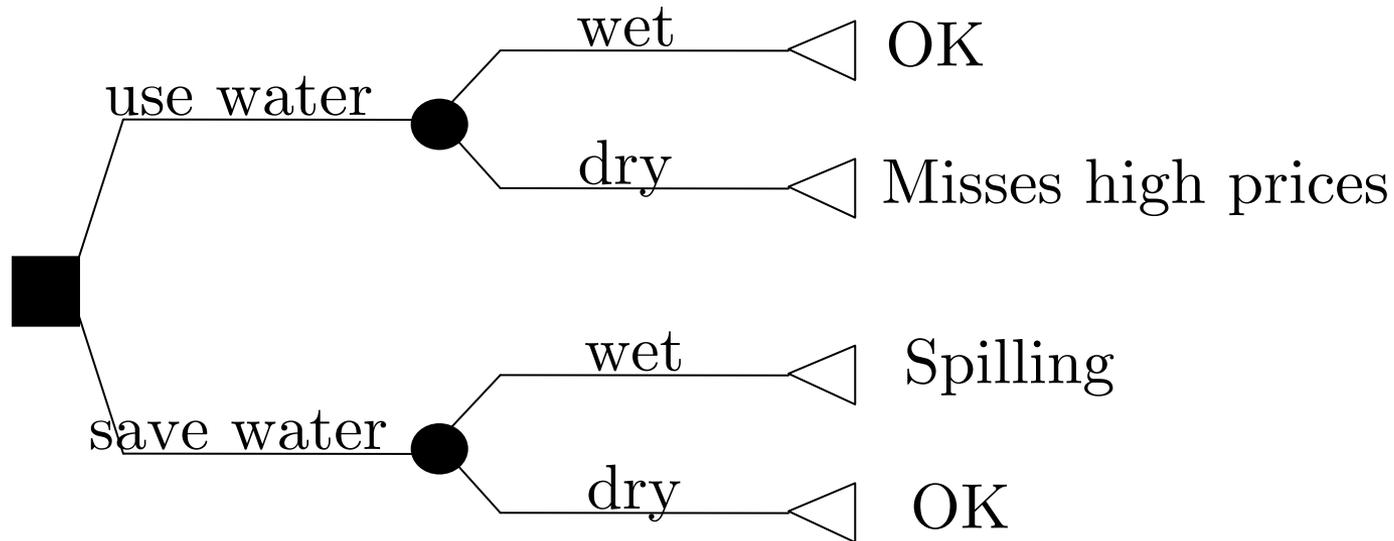
How to optimise reservoirs with

- stochastic inflow
- stochastic spot- and forward prices
- multi-year storage capacity?

Power station and reservoir



Scheduling problem



- “Marginal costs” are opportunity costs of discharging water
- Avoid spilling, discharge when prices are high

Scheduling problem

How to calculate the expectations? Forecasts or forward curve?

$$\max E_{\pi, p} \left[\sum_{t=1}^T \frac{\pi_t p_t}{(1+k)^t} + \frac{V(l_T, \pi_T)}{(1+k)^T} \right]$$

subject to

hydro balance

lower and upper bounds on
reservoir and discharge

Notation:

π = price

p = generation

k = discount interest rate

V = value at end of horizon

l = reservoir volume

Hydro scheduling – hierarchy

(Fosso et al., 1999)

Reservoir management
Horizon: 2-3 years

Time step: 1 week

- ❑ Scheduling discharges
- ❑ The horizon depends on the size of the reservoir compared to the annual inflow
- ❑ There may also be a medium term model

Short term planning
Horizon: 24-168 h

Time step: 1 h

- ❑ Detailed generation allocation with signals from the long term models
- ❑ Bidding into the physical day-ahead market

Production and information

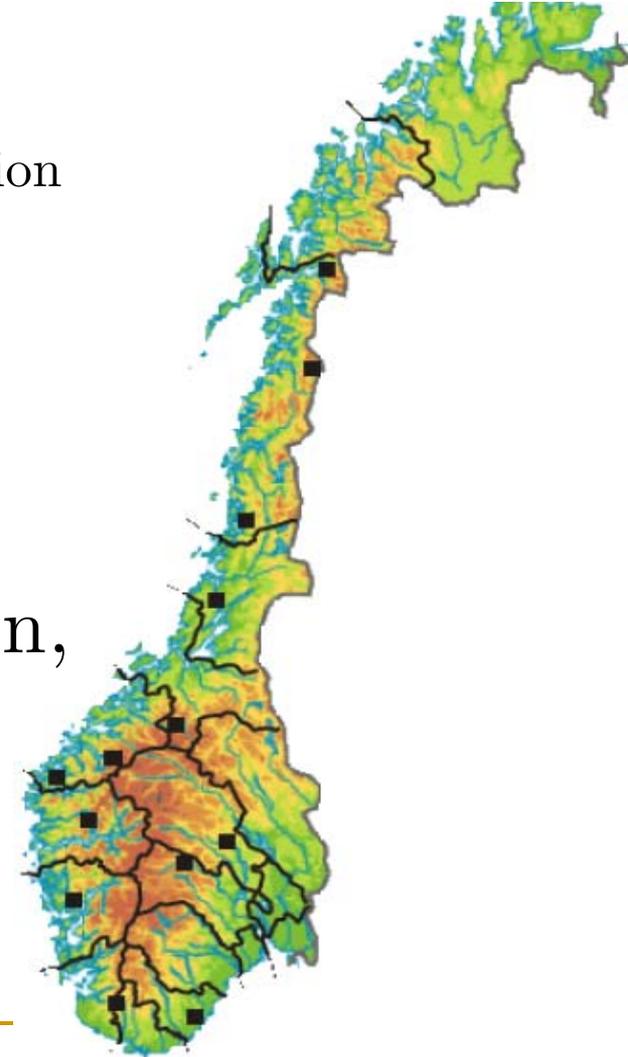
- Hydropower producer should consider
 - (i) current spot price and expected future prices
 - (ii) water reservoir level and expected inflow
 - (iii) production constraints
 - For instance,
 - The higher the forward prices the more should be produced later
 - The higher the water level the more should be produced now
 - Producers have continuous access to spot and forward price information
 - Inflow forecasts are not reliable beyond one week ahead
 - Daily inflow forecasting, price forecasting, bidding
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Empirical questions

- Is derivative price information used in hydropower scheduling?
 - Do forward prices explain realized production schedules?
 - Does it help to use forward prices?
 - Which factors drive generation scheduling?
 - Prices, inflow, reservoir levels, ...
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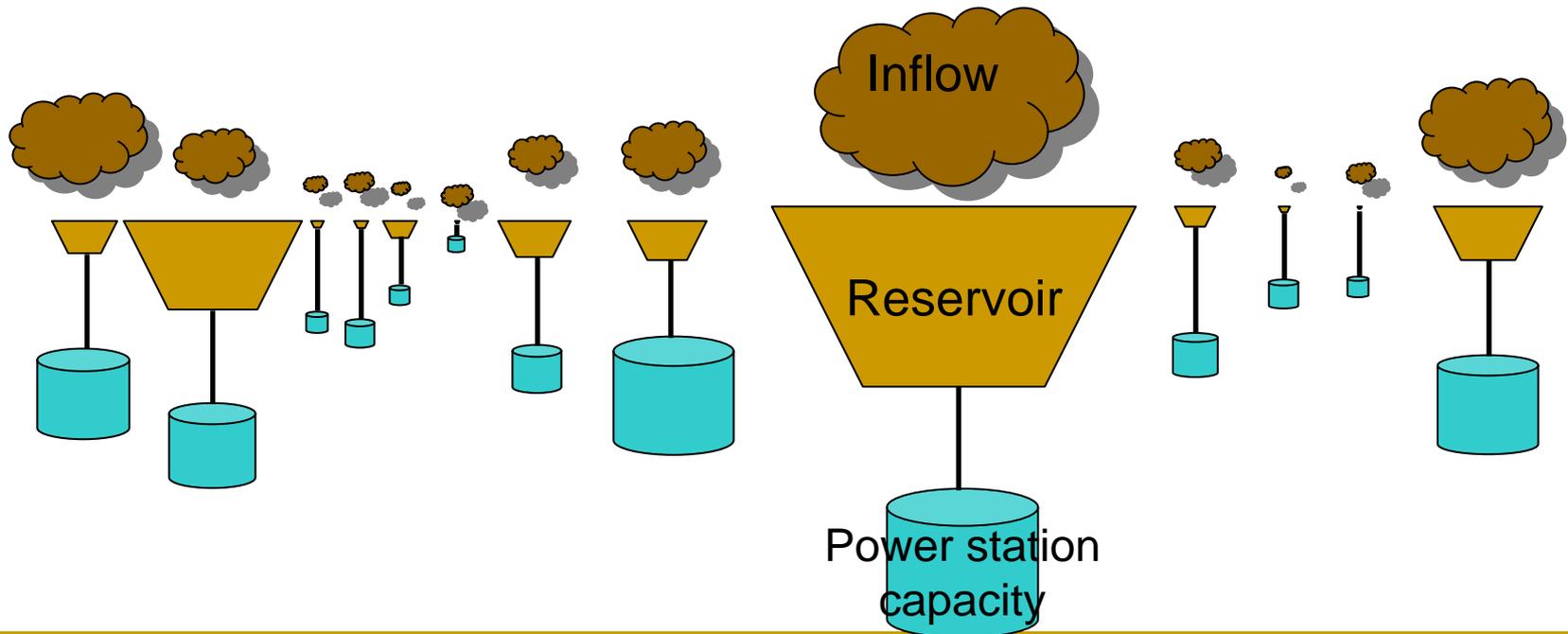
Data

- 13 Norwegian plants, having one main reservoir
 - 9 producers say that they use forward information
 - 4 producers use their own forecasts
- The largest producers (Statkraft, Hydro) are not represented
 - We consider only price takers
- Weekly data 2000-2006: generation, reservoir level, inflow
- Nord Pool prices
 - Elspot (day ahead) and Eltermin (futures and forwards)



Producers

	1	2	3	4	5	6	7	8	9	10	11	12	13
Capacity MW	128	120	30	40	28	23	68	167	210	62.1	41	29	140
kWh/m3	1.16	1.32	1.15	1.27	0.67	0.16	1.25	1.09	1.46	1.5	0.95	0.91	1.36
Reservoir GWh	228.1	624.4	47.1	51.8	118.9	14	255	272.5	1270	142	42.6	12.4	380.8
Inflow GWh/y	641.2	380.8	106.6	139.9	87.8	153	272.3	414.4	1250.5	231.8	81.3	147.2	662.9
Relative reservoir	0.356	1.64	0.442	0.37	1.35	0.092	0.937	0.642	1.015	0.613	0.953	0.084	0.574
Capacity factor %	57.2	36.2	40.5	39.9	35.8	76	45.7	28.3	68	42.6	22.6	57.9	54



Regression model variables

- Dependent variable is weekly production relative to the capacity
- Main explanatory variables:
 - Inflow relative to capacity
 - Spot price relative to forward price (nearest season or quarter), we call this as Basis
 - Seasonality dummies: months and filling season (weeks 18-39)
 - Relative production in the previous week
- Additional effects through dummy variables:
 - Reservoir level $>$ average level: Production should be higher.
 - Reservoir level is high or low (over/below 90%/10% of the max level): Production should depend less on the market prices.
 - Reservoir level $>$ 90% of the max level: Production should depend more on inflow.
 - Spot price $>$ 95% of the max price: Production should be high.
 - Spot price volatility $>$ 95% of the max volatility: Production should be low.
 - Producer claims to use forward prices in the scheduling: Production should depend more on the market price.

Regression model

- Granger causality test:
 - Controlling for seasonality
 - Basis Granger causes aggregate production of the 13 power plants
 - The aggregate production does not Granger cause Basis
 - OLS estimation procedure
 - Fixed effects: A dummy on the intercept for each producer
 - Lagged production as a covariate, all the other covariates are assumed to be strictly exogenous
 - Each producer in the model is allowed to have its own sensitivity towards inflow, seasonal inflow, and lagged production (only own lagged production)
 - In-sample period: week 5, 2000 – week 52, 2004;
out-of-sample period: week 1, 2005 – week 52, 2006
 - Out-of-sample R^2 is used as criterion
 - Typical model:
Production week $t = \text{constant} + \text{dummies} + \text{inflow} + \text{spot price relative to forward price} + \text{lagged production}$
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Best model

- Best out-of-sample model for the relative production (producer i and week t):

$$p_{i,t} = \alpha_i + \underset{(0.025)}{0.084} \cdot Basis_t + \beta_{1,i} inflow_{i,t} + \beta_{2,i} S_t inflow_{i,t} + \beta_{3,i} p_{i,t-1} \\ + \sum_{k=2}^{12} \hat{\beta}_k M_{k,t} + \sum_{k=1}^6 \tilde{\beta}_k H_{k,i,t} + \varepsilon_{i,t}$$

where S_t , $M_{k,t}$, and $H_{k,t}$ are the filling season, month, and the hypothesis dummies

- Out-of-sample R^2 is 78%

Best model, Cont'd

- The higher the spot price relative to the forward prices, the higher the production
 - The higher the inflow the higher the production
 - Less so in the filling season (if $S_t=1$)
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Additional effects

- A higher reservoir than normal increases production (confirmed)
- When reservoirs are nearly full or nearly empty, market prices are less important (confirmed)
- Inflow is more important when reservoirs are nearly full (confirmed)
- Production is high at the highest prices (opposite is found – they had low reservoir levels)
- Production decreases when spot price volatility is very high (confirmed)
- Producers that claim to use forward price information are more sensitive against market price changes (confirmed)

Production changes

- Best out-of-sample model:

$$\Delta p_{i,t} = 6.05 + 0.03 \Delta inflow_{i,t} + 5152.68 \Delta Basis_{i,t}$$

and its R^2 is 3%.

- The R^2 is consistent with the best empirical work in financial time series (see, e.g., Campbell and Thompson (2008))
 - R^2 is lower since we model differences
 - The forward price is also in this model
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More on the use of forwards

- 4/13 of the producers report that they do not use forward prices to guide scheduling
 - They instead use their own forecasts
- This is confirmed by the data:
 - This difference is significant: The four use significantly less forward information than the nine
- The group which uses forwards have significantly higher production volatility (608% vs. 575%, annualized)

More on the use, Cont'd

- Cash flows normalized wrt production capacity are not significantly different:
 - With forward information: average = 10.78, standard deviation = 10.09
 - Without forward information: average = 12.24, standard deviation = 12.67
 - Performance measures that avoid valuation of water may be hard to come by
 - Does it really help using forward price information?
 - Data indicates the case is not clear
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Conclusion

- Forward prices are significant in driving production scheduling
 - Our model simplifies hydro scheduling in practice
 - 4/13 do not use forward information, the rest say they use
 - Forward prices explain significantly more the production of the nine companies
 - Those using forward info are not performing significantly better than those who use own forecasts
 - Large variance in spot prices decreases production
 - This is due to the value of waiting
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