

The overnight risk premium in electricity forward contracts

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Abstract

We analyze the risk premium on electricity forward contracts traded for the Nordic and German/Austrian electricity markets. We argue that finding risk premiums by analyzing overnight returns is more relevant than the frequently used ex post approach. The derivatives in these markets can be characterized as trading products and hedging products. Each contract shows a clear increase in trading volume and liquidity when approaching maturity. We link this to a testable hypothesis where financial traders are compensated for holding price risk, and where the sign and magnitude of the risk premium changes depending on the hedging pattern of producers and retailers. Incorporating this in regressions we find that there are higher risk premiums in the period before the forwards become front products, compared to the risk premiums in the front period. Quarterly and monthly contracts show the most significant results.

1. Introduction

We analyze the risk premium in electricity forward markets, specifically the Nordic and German/Austrian market. The risk premium formation in these markets has attracted much attention in the academic community, and much of the focus has been on the ex post risk premium. The ex post risk premium is the difference between the settlement price of the forward contract and the realized average spot price over the corresponding settlement period. The ex post risk premium is investigated in Geman and Vasicek (2001), Shawky et al. (2003), Longstaff and Wang (2004), Redl et al. (2009), Bunn and Chen (2013), Veka (2013), Cartea and Villaplana (2014) and Haugom et al. (2014) amongst others. Botterud et al. (2010) regress the log return of the ex post payoff with respect to explanatory variables following Fama and French (1987). We investigate the risk premium by analyzing the forward contracts directly. Let the price of a forward contract be denoted

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18 by F , and the volatility of the forward contract σ . Then a simple model for the dynamics of the
19 forward contract is

$$\frac{dF}{F} = \mu dt + \sigma dW. \quad (1)$$

20 A forward contract does not require an investment.¹ Since there is no investment, the forward
21 contract should under normal circumstances not command a risk premium, hence μ in eq. (1)
22 should be zero. However, the electricity market is different from many other markets since storing
23 electricity is very costly. In fact, the cost of carry argument used to derive forward prices in other
24 markets does not hold for electricity derivatives.

25 We distinguish between three types of players; producers, retailers and traders. Due to highly
26 volatile electricity prices and the unique non-storable nature of electricity, producers of power will
27 typically want to hedge their physical production a few years ahead (Sanda et al., 2013). If there
28 are no natural buyers in the marketplace, such as retail companies, a financial trader may be the
29 counterpart of the producer. To take on the price risk, the trader may command a risk premium
30 from the producer. That is, to be long in the market, the trader will require a positive risk premium
31 and $\mu > 0$ in eq. (1). Retailers may want to hedge their physical delivery commitments. However,
32 they do not enter the market before they know their sales obligations to end users. When these
33 enter the market, their counterparts are the financial traders that offloaded the price risk from the
34 producers. Financial traders have no incentive to hold contracts over settlement periods since they
35 have no purchase or sales commitments in the spot market. When retailers meet the traders in
36 the marketplace, the risk premium is likely to vanish, and $\mu = 0$ in eq. (1). Another scenario is
37 that retailers will want to offload more of their price risk than the producers. This means that
38 financial traders will be net short after selling to retailers. To hold this price risk the trader will
39 again command a risk premium. In this case, the trader must be compensated for holding a short
40 position, and $\mu < 0$ in eq. (1). Although this is a simplified model for the behavior in the market,
41 it can be formulated as a testable hypothesis:

¹Forward contracts do require a margin paid to the clearing house or as collateral for credit risk in bilateral agreements. The margin account typically pays the risk free interest rate. If the borrowing cost of the investor is the risk free rate, the forward contract will effectively not require an investment. While this is typically not the case, we assume that the entering the forward contract is costless for the investor. This is a standard assumption made in financial theory.

$$\frac{dF}{F} = (\mu + \alpha I)dt + \sigma dW, \quad I = \{0, 1\}. \quad (2)$$

42 In eq. (2), $I = 0$ before the retailers enter the market, and $I = 1$ afterwards. The interpretation
 43 is that the drift for the forward contract is equal to μ in the holding period of the traders, and $\mu + \alpha$
 44 in the hedging period of the retailers. Hence we expect α to be negative. The risk premium in our
 45 model depends on supply and demand for hedging and speculation, which in turn is determined
 46 by the characteristics of the market participants. Note it does not depend on expected spot prices.
 47 We will measure the change in the forward price via the closing prices on consecutive trading days.

48 The next step is to identify when retailers enter the market. Electricity markets show interest-
 49 ing trading patterns. Figure 1 shows log returns and trading volume for the Q2-07 contracts on
 50 NASDAQ OMX and EEX. These are financially settled, where the payoff depends on the difference
 51 between the agreed contract price and the average system (spot) price during the second quarter
 52 of 2007. Notice the sharp increase in trading volume when the Q2-07 contract becomes the front
 53 product, i.e. the quarter product with shortest time to maturity. This sharp increase is interpreted
 54 as the entry point for retailers. The increased trading volume also has implications for any liquidity
 55 premium in the contract price. In fact, μ in eq. (1) might just as well be interpreted as a liquidity
 56 premium, and the sign of the premium is determined by the buyers and sellers in the market. Our
 57 arguments remain valid.

58 Redl et al. (2009) shows that the main characteristics of price formation at the EEX and Nord
 59 Pool forward markets are similar. Electricity prices in both markets are volatile and have occasional
 60 price spikes due to the non-storability of electricity, that makes it difficult to dampen imbalances
 61 between supply and demand. For further details about the electricity market dynamics we refer to
 62 Frestad et al. (2010), Benth et al. (2008) and Huisman and Kilic (2013). Linking the risk premium
 63 in the forward market to supply and demand of contracts has been done in Benth et al. (2008),
 64 Marckhoff and Wimschulte (2008) and Botterud et al. (2010). Geman and Vasicek (2001), Longstaff
 65 and Wang (2004), Weron (2008), Botterud et al. (2010) and Lucia and Torró (2011) find negative
 66 risk premiums on average. Our approach avoids the pitfalls mentioned by Weron and Zator (2014).
 67 In our model the risk premium is negative if $\mu + \alpha < 0$.

68 Following the intuition that financial traders command the risk premium, other interesting
 69 implications can be drawn. As the trader only will command a risk premium for risk that he or
 70 she must carry, any ex post estimate of the risk premium is hard to interpret. If a financial trader

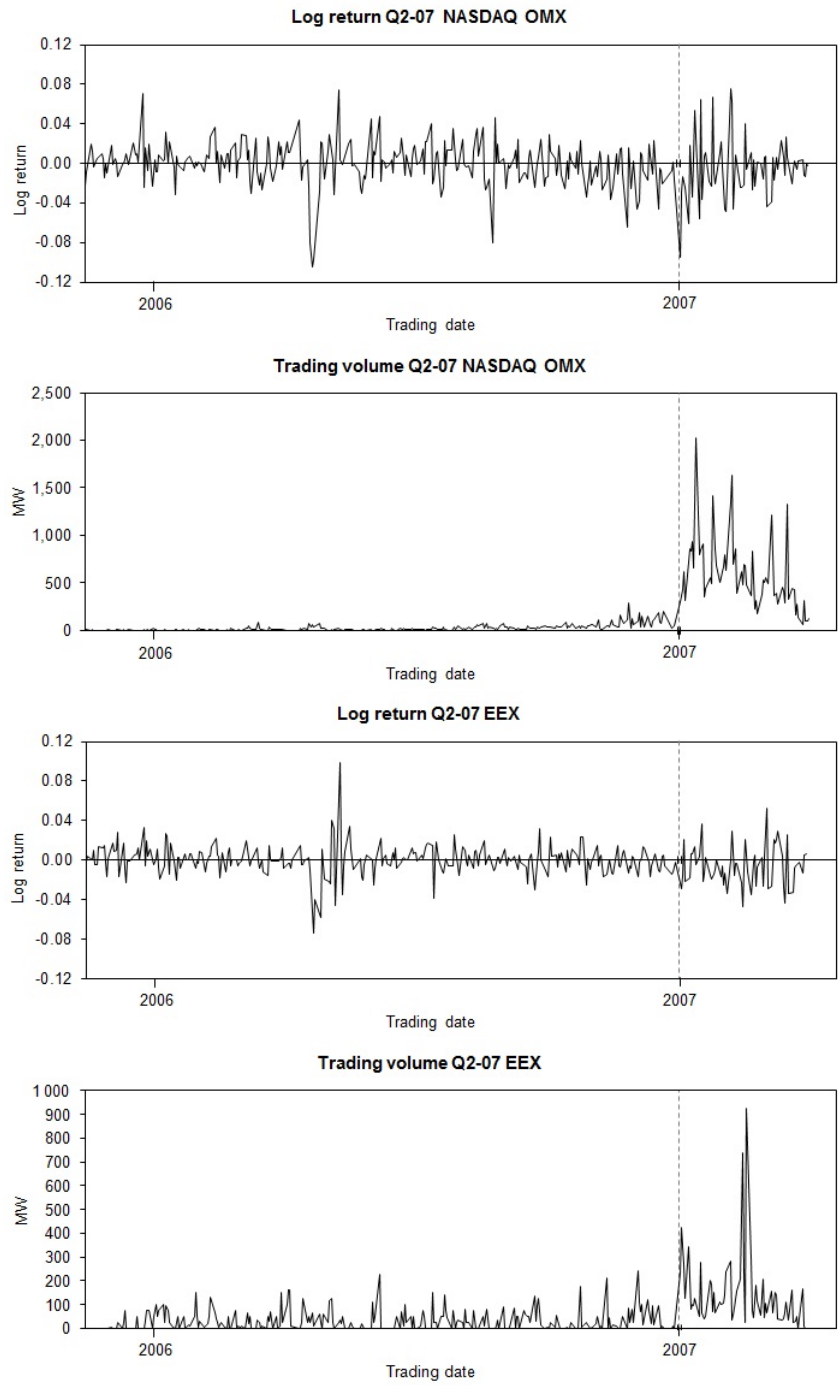


Figure 1: Log return and trading volume for the Q2-07 contracts on NASDAQ OMX and EEX. The dotted line shows when the contract becomes the front product at the exchange.

71 holds an annual forward contract that enters the settlement period, the ex post estimate of the
72 risk premium will be the last closing price minus the realized spot price over the year. However,
73 the trader does not have to hold the entire exposure throughout the year. As January approaches,
74 the trader can simply short the remainder of the year at any day². Summarizing, we hold that
75 the underlying premise of the ex post approach is that the speculators are to buy and hold the
76 contracts until maturity; the apparently relevant benchmark is the realized spot prices during the
77 delivery period of the contracts. However, this kind of thinking does not match the practice of the
78 speculators. Thus the ex post risk premium measures a compensation for risk that does not need
79 to be held, which erodes the interpretation of such calculations.

80 Another problem with ex post analysis of the risk premium is well known. Using realized spot
81 prices in the delivery period of the contracts means that there will be a forecast error component to
82 the estimated risk premiums. Given the amount of time between the date of the relevant contract
83 price and the realization of spot prices, the forecast error might not average to zero even over a few
84 years, leading to uncertainty in parameter estimates. A possible approach to mitigate this problem
85 is to analyze ex ante risk premiums using an explicit spot price model (Benth et al., 2008; Weron,
86 2008; Benth et al., 2013). However, the estimated risk premiums then become dependent on the
87 subjective choice of spot price expectation, for which no consensus model exists.

88 Our contributions include an alternative approach to estimating risk premiums, avoiding previ-
89 ously unrecognized issues with interpretation of ex post analysis. We formulate a testable hypoth-
90 esis that is supported by analyzing electricity forward contracts in the Nordic and EEX electricity
91 markets, thereby shedding new light on how the risk premiums are formed.

²Consider another example. A trader is long a Q2 contract. At 1 April, this contract has entered its settlement (delivery) period and the trader is exposed to the difference between the contract price and the realized spot price over the remainder of Q2. However, much of the risk can be offloaded, by shorting the May and June contracts, and hold these throughout the quarter. The remaining exposure is to April spot prices, but for this the trader can short the last three weeks of April. Further, there exists day futures to cover daily exposure, meaning the remainder of the week can be hedged. The only risk the trader cannot hedge is the overnight price change. The ex post risk premium measures a compensation for carrying risk that is hedgeable, thus it loses interpretation power. This reasoning carries over to other commodities as well. It does not apply when e.g. trading day-ahead forward contracts in the US (Bessembinder and Lemmon, 2002; Longstaff and Wang, 2004; Haugom and Ullrich, 2012), since these contracts cannot be hedged.

NASDAQ OMX Contracts																
Contract types	Count		Mean, %		Median, %		Max,%		Min, %		Std dev, %		Skewness		Kurtosis	
	B	F	B	F	B	F	B	F	B	F	B	F	B	F	B	F
Y Q M	27879	5229	0.01	-0.08	0.05	0.00	14.09	12.74	-15.97	-16.71	1.95	2.63	-0.32	-0.17	5.25	3.00
Y Q	19379	3514	0.03	-0.03	0.07	0.00	12.73	10.47	-15.20	-15.65	1.60	2.29	-0.48	-0.28	5.85	3.06
Q M	23374	3473	0.00	-0.14	0.05	-0.12	14.09	12.74	-15.97	-16.71	2.07	2.97	-0.29	-0.10	4.57	2.17
Y	4505	1756	0.03	0.02	0.04	0.10	7.74	9.19	-8.97	-9.63	1.16	1.80	-0.62	-0.40	7.95	3.39
Q	14874	1758	0.03	-0.09	0.09	-0.05	12.73	10.47	-15.20	-15.65	1.71	2.69	-0.46	-0.19	5.12	2.07
M	8500	1715	-0.06	-0.18	0.00	-0.20	14.09	12.74	-15.97	-16.71	2.57	3.23	-0.14	-0.03	2.90	1.99

EEX Contracts																
Contract types	Count		Mean, %		Median, %		Max,%		Min, %		Std dev, %		Skewness		Kurtosis	
	B	F	B	F	B	F	B	F	B	F	B	F	B	F	B	F
Y Q M	27665	5232	-0.01	-0.08	0.00	-0.08	14.89	16.27	-22.52	-14.61	1.27	1.66	-0.22	0.10	12.66	8.83
Y Q	18528	3501	0.02	-0.03	0.00	-0.04	10.91	14.63	-22.52	-8.36	1.13	1.33	-0.38	0.29	15.66	7.71
Q M	20921	3458	-0.02	-0.13	0.00	-0.14	14.89	16.27	-22.52	-14.61	1.39	1.86	-0.22	0.15	11.08	7.78
Y	6744	1774	0.04	0.01	0.00	0.00	7.32	8.84	-7.15	-7.05	0.82	1.19	0.11	-0.02	12.45	5.42
Q	11784	1727	0.00	-0.07	0.00	-0.10	10.91	14.63	-22.52	-8.36	1.27	1.45	-0.42	0.49	13.73	8.26
M	9137	1731	-0.06	-0.20	-0.04	-0.18	14.89	16.27	-19.06	-14.61	1.53	2.19	-0.04	0.09	8.73	6.03

Table 1: Descriptive statistics for return on forward contracts traded on NASDAQ OMX and EEX. B/F indicates trading before front period (B) and in front period (F).

92 2. Data

93 We have examined prices from the Nordic market as traded at NASDAQ OMX and Ger-
94 man/Austrian contracts traded at EEX. Our data set spans 2 January 2003 to 30 September 2012.
95 It consists of 7 annual contracts, 28 quarterly contracts and 82 monthly contracts in both markets;
96 33108 observations of NASDAQ OMX prices and 32897 observations of EEX prices. Descriptive
97 statistics of the contracts is given in Table 1.

98 3. Risk premium estimation

99 The formulation in eq. (2) must be discretised to be fitted to observed data. The resulting
100 model can be stated as

$$r_t \sim N((\mu + \alpha I_t)\Delta t, \sigma\sqrt{\Delta t}), \quad (3)$$

101 or

$$r_t = \mu + \alpha I_t + \epsilon, \quad (4)$$

102 where r_t is the first difference of the natural logarithm of the forward price, $\Delta t = 1$ day and
103 $\epsilon \sim N(0, \sigma)$. It is well known that the log returns of financial series are not normally distributed.
104 We employ ordinary least squares to estimate μ and α , thus the estimators are unbiased even

105 without the assumption of normally distributed returns. The annualized risk premium, given daily
 106 observations of the forward curve, is given by

$$e^{(\mu+\alpha I_t)252} \quad (5)$$

107 for 252 trading days per year.

108 Redl and Bunn (2013) and Bunn and Chen (2013) argue that the forward electricity risk premium
 109 is dependent on the underlying fuel. We therefore expand the model to control for fuels, second
 110 and third moment of returns and seasonal effects as in Longstaff and Wang (2004) and Lucia and
 111 Torró (2011). The estimated model includes the following control variables;

- GAS_t Logarithmic return of ICE Natural Gas Index,
- OIL_t Logarithmic return of front line ICE gas oil futures,
- $COAL_t$ Logarithmic return of front line API2 Atlantic Basin, CIF,
- EMI_t Logarithmic return of Argus European Union Allowances Carbon Dioxide front year,
- RES_t Deviation from normal Nordic hydro reservoir level,
- 112 WAV_t Logarithmic return of Nordic water value³,
- $STOCK_t$ Logarithmic return of the stock market,
- VOL_t Change of trading volume,
- VAR_t Spot price variance,
- $SKEW_t$ Spot price skewness,
- $MONTH_{it}$ Seasonal variation, monthly dummies.

113 For both markets, the regressions are on the form

$$\begin{aligned}
 r_t = & \alpha_1 \cdot I_t + \alpha_2 \cdot GAS_t + \alpha_3 \cdot OIL_t + \alpha_4 \cdot COAL_t \\
 & + \alpha_5 \cdot EMI_t + \alpha_6 \cdot RES_t + \alpha_7 \cdot WAV_t + \alpha_8 \cdot STOCK_t \\
 & + \alpha_9 \cdot VOL_t + \alpha_{10} \cdot VAR_t + \alpha_{11} \cdot SKEW_t \\
 & + \sum_{i=1}^{12} (\alpha_{11+i} \cdot MONTH_{it}) + \epsilon_t,
 \end{aligned} \quad (6)$$

114 however, water value, trading volume and reservoir level was not used for the EEX.

115 We estimate the model for individual contracts and using pooled ordinary least squares to

³As a proxy for the water value, we follow Sandmark and Tennbakk (2010) and use the spot price in the zone NO1 in Norway, which contains only hydropower plants.

116 sharpen statistical inference. We have fitted the model over different time periods to reveal possible
117 changes in parameter values over time.

118 **4. Results**

119 Table .2 and Table .3 show the results for the Nordic market and the German/Austrian market
120 respectively. The first two columns show the time period, while the third column shows which
121 contract types are included in the regression.

122 The first six rows represent models that are estimated on different combinations of contracts
123 spanning the whole data period. In the Nordic market, the four pooled regressions that include the
124 quarterly contracts have statistically significant risk premium parameters at a 5 % confidence level.
125 The estimated risk premium for monthly and quarterly contracts gives support for our hypothesis.
126 We do not see the same effect for annual contracts. One reason for this might be that retailers
127 do not hedge purchases on an annual basis, but focus on shorter term contracts such as weekly,
128 monthly or quarterly.⁴

129 The remaining rows contain results of an analysis of returns for four-year rolling time windows.
130 In most of the regressions that include quarter contracts, the return in the period before front is
131 positive and the return in the front period is negative. We see that our hypothesis has more support
132 in earlier time periods, that is, before 2010. This may represent a structural break caused by the
133 worldwide financial debt crisis. It may also be interpreted as improved market efficiency. While it
134 is difficult to pin down the cause of the observed change, we can provide statistical evidence that
135 a change did occur. To do so, we estimate two models for each market; one where the effect is
136 restricted to be constant over the whole sample, and one model where we include a dummy variable
137 that takes the value of one if the data point is observed before 1. January 2010, and zero otherwise.
138 This allows us to test for parameter stability using an ANOVA test. Under the null hypothesis the
139 test statistic follows an F distribution, with parameters dependent on sample size and the number
140 of imposed restrictions. For the Nordic market the test statistic is F distributed on 28242 degrees
141 of freedom, and takes the value of 8.20. The corresponding p-value is 0.004. The German/Austrian
142 market parameter stability test is F distributed on 27978 degrees of freedom, and takes the value
143 of 21.15. The corresponding p-value is 4.27e-06.

⁴Residential customers use 1-year fixed contracts to a very limited extent according to Mirza and Bergland (2012).

144 The German/Austrian market results offer support for our hypothesis. The effect of the front
145 dummy on the risk premium sign and size is most prominent for monthly contracts in this market.

146 Robustness of the results is checked by adding the control variables to the regressions, i.e.
147 running regressions on the form (6). The results are detailed in Appendix A. The results further
148 strengthen our hypothesis. The coefficient of the front dummy is still negative and statistically
149 significant, and the signs of the coefficients for the control variables are in most cases as expected.

150 The coefficient for the stock return is positive and significant, indicating that the greater the
151 stock return, the greater the return on the forward contracts. The coefficient for the reservoir
152 level variable is negative and significant. This indicates that the risk premium increases with an
153 increase in the negative deviation from the mean level and with a decrease in the positive deviation
154 from the mean level. The positive coefficient associated with trading volume indicates that an
155 increase in trading volume is consistent with a higher return on the forward contracts. Return on
156 water value has a negative and significant effect, indicating that negative returns on water value are
157 consistent with higher returns on the forward contracts. When the water value decreases, producers
158 are more interested in hedging their revenue. Thereby the risk premium increases. The coefficients
159 for oil, coal and natural gas are all positive and significant. Increasing fossil fuel prices will lead
160 to an increase in electricity prices through increased marginal production cost, which in turn will
161 lead to an increased demand for electricity forward contracts. The coefficient for emission rights
162 is also positive, but slightly less significant. For the same reason as the fossil fuels, an increase
163 in the demand for emission rights is consistent with an increased demand for electricity forward
164 contracts. Finally, the coefficients for the skew and variance of the spot price are positive and
165 negative respectively, both highly significant. It is worth mentioning that this is consistent with
166 the findings of Bessembinder and Lemmon (2002) even though our variable definitions are slightly
167 different. The results indicate that the greater the skew of the spot price, the greater the return on
168 the electricity forward contracts. The negative coefficient of the spot price variance indicates that
169 the higher the spot price variance, the lower the return on the electricity forward contracts.

170 As an additional control, we have run the same regressions as in eq. (6) except substituting the
171 front dummy with the daily trading volumes of the contracts in level form. This experiment also
172 confirm our hypothesis, giving a statistically significant negative coefficient for the volume variable,
173 which shows that high volumes (i.e. in the front period) are consistent with lower risk premiums.

174 **5. Conclusion**

175 This paper examines the risk premium on electricity forward contracts traded at NASDAQ OMX
176 and EEX. We have formulated a testable hypothesis where financial traders are compensated for
177 holding price and liquidity risk. The novelty of our approach is that we analyze overnight returns,
178 which is more in line with the exposure that traders in these markets actually hold. We have
179 analyzed empirical data from month, quarter and annual delivery length products, and find strong
180 support for our hypothesis. That is, the risk premiums in the markets are positive, on average,
181 when producers hedge their production, and turn negative when large buyers, e.g. retailers, enter
182 the market. Front contracts, i.e. those nearest to maturity, are trading (speculative) products, while
183 the products further out on the forward curve are hedging products.

184 We have used data for contracts with delivery periods from 2003–2012. We find that the risk pre-
185 mium decreases over time as we approach the delivery period. The daily returns on future contracts
186 traded for the Nordic electricity market show an annual return of 1.3 % when producers hedge,
187 and -18.7 % after retailers enter the market. The corresponding results for the German/Austrian
188 market are -0.9 % and -7.6 %, respectively. These negative risk premiums confirm previous findings.

189 Quarterly contracts show the most prominent effect in the Nordic market, whereas monthly
190 contracts give the most significant result in the German/Austrian market. Annual contracts show
191 no effect on the risk premium of market participation in either market. We find a higher risk
192 premium in the Nordic market in absolute terms. This could be an indication of market inefficiency
193 in the Nordic Power market and a lack of integration with other financial markets.

194 **6. Acknowledgements**

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196 (RCN grant 209697), and acknowledge financial support from the Research Council of Norway
197 through project 228811.

Table .2: Results from regression $r_t = \mu + \alpha I_t + \epsilon_t$ (4) run for overnight return on forward contracts for the Nordic area, where I_t is a dummy indicating trading in front period (1) or not (0).

Start date	End date	Contract types		Ann. return b/f front, %	Ann. return front, %	μ	α	$\Pr(> t)$	$\Pr(> t)$	$R^2 \cdot 10^3$
		Y	M							
01.01.2003	30.09.2012	Y	Q	1.3	-18.7	0.00005	-0.00088	0.6725	0.0048 **	0.2400
01.01.2003	30.09.2012	Y	Q	8.5	-8.0	0.00033	-0.00066	0.0084 **	0.0368 *	0.1904
01.01.2003	30.09.2012	Q	M	-6.0	-28.4	-0.00025	-0.00109	0.0148 *	0.0000 ***	0.6756
01.01.2003	30.09.2012	Y	Q	7.2	5.7	0.00028	-0.00006	0.1720	0.8790	0.0037
01.01.2003	30.09.2012	Q	Q	8.9	-19.9	0.00034	-0.00123	0.0237 *	0.0082 **	0.4205
01.01.2003	30.09.2012	M	M	-13.4	-36.9	-0.00057	-0.00127	0.0499 *	0.0751 .	0.3101
01.01.2003	31.12.2006	Y	Q	17.3	-6.1	0.00064	-0.00089	0.0009 ***	0.1246	0.2401
01.01.2004	31.12.2007	Y	Q	12.8	-14.0	0.00048	-0.00109	0.0023 **	0.0151 *	0.4022
01.01.2005	31.12.2008	Y	Q	1.0	-24.0	0.00004	-0.00114	0.8180	0.0110 *	0.3571
01.01.2006	31.12.2009	Y	Q	-3.0	-22.3	-0.00012	-0.00089	0.4908	0.0455 *	0.2056
01.01.2007	31.12.2010	Y	Q	2.3	-3.8	0.00009	-0.00025	0.5990	0.5700	0.0170
01.01.2008	31.12.2011	Y	Q	-5.6	-17.1	-0.00023	-0.00052	0.2310	0.2520	0.0774
01.01.2009	30.09.2012	Y	Q	1.1	-13.1	0.00004	-0.00061	0.8340	0.1900	0.1293
01.01.2003	31.12.2006	Q	M	18.2	-22.3	0.00067	-0.00168	0.0056 **	0.0496 *	0.5276
01.01.2004	31.12.2007	Q	M	11.9	-31.0	0.00045	-0.00193	0.0216 *	0.0021 **	0.8475
01.01.2005	31.12.2008	Q	M	-0.3	-38.4	-0.00001	-0.00193	0.9518	0.0015 **	0.7139
01.01.2006	31.12.2009	Q	M	-4.3	-31.4	-0.00018	-0.00133	0.3945	0.0204 *	0.3483
01.01.2007	31.12.2010	Q	M	2.4	-10.0	0.00009	-0.00052	0.6350	0.3530	0.0559
01.01.2008	31.12.2011	Q	M	-5.3	-23.5	-0.00022	-0.00085	0.2990	0.1300	0.1581
01.01.2009	30.09.2012	Q	M	0.4	-19.6	0.00002	-0.00089	0.9420	0.1050	0.2223
01.01.2003	31.12.2006	Y	Y	14.3	17.0	0.00054	0.00009	0.0347 *	0.8708	0.0105
01.01.2004	31.12.2007	Y	Y	16.1	18.0	0.00060	0.00007	0.0030 **	0.8800	0.0065
01.01.2005	31.12.2008	Y	Y	6.5	5.9	0.00025	-0.00022	0.3050	0.9680	0.0004
01.01.2006	31.12.2009	Y	Y	3.3	-0.3	0.00013	-0.00014	0.6360	0.7960	0.0166
01.01.2007	31.12.2010	Y	Y	1.8	9.8	0.00007	0.00030	0.8130	0.5970	0.0800
01.01.2008	31.12.2011	Y	Y	-7.9	-2.8	-0.00033	0.00022	0.4680	0.7620	0.0368
01.01.2009	30.09.2012	Y	Y	10.7	5.3	0.00041	-0.00020	0.5030	0.8160	0.0363
01.01.2003	31.12.2006	Q	Q	25.9	-10.8	0.00092	-0.00138	0.0001 ***	0.1500	0.3849
01.01.2004	31.12.2007	Q	Q	22.7	-18.4	0.00082	-0.00163	0.0000 ***	0.0200 *	0.6993
01.01.2005	31.12.2008	Q	Q	12.3	-28.7	0.00047	-0.00182	0.0222 *	0.0075 **	0.7824
01.01.2006	31.12.2009	Q	Q	5.4	-23.4	0.00021	-0.00127	0.3457	0.0616 .	0.3685
01.01.2007	31.12.2010	Q	Q	5.5	-6.0	0.00021	-0.00046	0.3190	0.4860	0.0512
01.01.2008	31.12.2011	Q	Q	-1.9	-16.4	-0.00008	-0.00064	0.7470	0.3490	0.1033
01.01.2009	30.09.2012	Q	Q	4.6	-11.3	0.00018	-0.00066	0.4790	0.3220	0.1506
01.01.2003	31.12.2006	M	M	-2.3	-33.7	-0.00009	-0.00155	0.8870	0.3810	0.3979
01.01.2004	31.12.2007	M	M	-10.9	-42.5	-0.00046	-0.00175	0.3420	0.1590	0.5806
01.01.2005	31.12.2008	M	M	-21.5	-47.2	-0.00097	-0.00158	0.0318 *	0.1647	0.3923
01.01.2006	31.12.2009	M	M	-18.8	-38.6	-0.00083	-0.00112	0.0412 *	0.2632	0.2104
01.01.2007	31.12.2010	M	M	-2.7	-13.9	-0.00011	-0.00049	0.7850	0.6120	0.0431
01.01.2008	31.12.2011	M	M	-10.3	-30.0	-0.00043	-0.00100	0.2640	0.2920	0.1855
01.01.2009	30.09.2012	M	M	-4.7	-27.0	-0.00019	-0.00106	0.6050	0.2320	0.2702

Note .p < .1, *p < .05, **p < .01, ***p < .001.

Table 3: Results from regression $r_t = \mu + \alpha I_t + \epsilon_t$ (4) run for overnight return on forward contracts for the German/Austrian area, where I_t is a dummy indicating trading in front period (1) or not (0).

Start date	End date	Contract types		Ann. return front, %	Ann. return back, %	μ	α	$\Pr(> t)$	$\Pr(> t)$	$R^2 \cdot 10^3$
		Y	M							
01.01.2003	30.09.2012	Y	Q	-0.9	-7.6	-0.00009	-0.00076	0.2613	0.0002 ***	0.4230
01.01.2003	30.09.2012	Y	Q	1.7	-2.7	0.00017	-0.00045	0.0488 *	0.0361 *	0.1993
01.01.2003	30.09.2012	Q	M	-2.4	-11.4	-0.00025	-0.00109	0.0148 *	0.0000 ***	0.6756
01.01.2003	30.09.2012	Y	Q	4.1	1.1	0.00039	-0.00028	0.0004 ***	0.2470	0.1574
01.01.2003	30.09.2012	Q	M	0.4	-6.3	0.00004	-0.00072	0.7412	0.0304 *	0.3470
01.01.2003	30.09.2012	Y	Q	-5.7	-15.7	-0.00062	-0.00138	0.0004 ***	0.0015 **	16.5000
01.01.2003	31.12.2006	Y	Q	8.5	6.3	0.00077	0.00018	0.0000 ***	0.6620	0.0210
01.01.2004	31.12.2007	Y	Q	3.1	-8.6	0.00030	-0.00126	0.0118 *	0.0002 ***	1.0570
01.01.2005	31.12.2008	Y	Q	6.8	-2.7	0.00063	-0.00091	0.0000 ***	0.0034 **	0.5469
01.01.2006	31.12.2009	Y	Q	-3.8	-11.7	-0.00040	-0.00099	0.0022 **	0.0023 **	0.5412
01.01.2007	31.12.2010	Y	Q	-4.9	-10.2	-0.00053	-0.00065	0.0000 ***	0.0317 *	0.2698
01.01.2008	31.12.2011	Y	Q	-4.1	-6.7	-0.00043	-0.00030	0.0005 ***	0.3051	0.0648
01.01.2009	30.09.2012	Y	Q	-5.6	-8.6	-0.00060	-0.00037	0.0000 ***	0.1800	0.1280
01.01.2003	31.12.2006	Q	M	9.9	-0.2	0.00088	-0.00090	0.0001 ***	0.2080	0.3378
01.01.2004	31.12.2007	Q	M	2.0	-17.2	0.00019	-0.00244	0.2880	0.0000 ***	2.6230
01.01.2005	31.12.2008	Q	M	6.3	-9.2	0.00058	-0.00163	0.0003 ***	0.0003 ***	1.2480
01.01.2006	31.12.2009	Q	M	-5.7	-15.8	-0.00061	-0.00139	0.0001 ***	0.0011 **	0.8285
01.01.2007	31.12.2010	Q	M	-5.9	-13.2	-0.00063	-0.00097	0.0000 ***	0.0118 *	0.4628
01.01.2008	31.12.2011	Q	M	-4.3	-8.4	-0.00046	-0.00048	0.0006 ***	0.1804	0.1303
01.01.2009	30.09.2012	Q	M	-5.7	-10.4	-0.00061	-0.00058	0.0000 ***	0.0694 .	0.2636
01.01.2003	31.12.2006	Y	Q	7.0	14.3	0.00065	0.00057	0.0000 ***	0.1430	0.4914
01.01.2004	31.12.2007	Y	Q	4.9	7.7	0.00046	0.00024	0.0000 ***	0.4290	0.1191
01.01.2005	31.12.2008	Y	Q	8.1	9.7	0.00074	0.00013	0.0000 ***	0.6890	0.0317
01.01.2006	31.12.2009	Y	Q	3.1	-1.8	0.00029	-0.00048	0.1130	0.2260	0.3392
01.01.2007	31.12.2010	Y	Q	0.0	-3.1	0.00000	-0.00032	0.9990	0.4390	0.1766
01.01.2008	31.12.2011	Y	Q	-1.7	-2.9	-0.00017	-0.00013	0.5930	0.7950	0.0274
01.01.2009	30.09.2012	Y	Q	-3.7	-3.6	-0.00039	0.00001	0.3250	0.9880	0.0002
01.01.2003	31.12.2006	Q	M	-0.9	-7.6	-0.00009	-0.00076	0.2613	0.0002 ***	0.4230
01.01.2004	31.12.2007	Q	M	6.0	-4.6	0.00056	-0.00105	0.0024 **	0.0831 .	0.6168
01.01.2005	31.12.2008	Q	M	8.6	3.2	0.00078	-0.00047	0.0000 ***	0.3600	0.1347
01.01.2006	31.12.2009	Q	M	-2.3	-10.0	-0.00023	-0.00091	0.2095	0.0864 .	0.4103
01.01.2007	31.12.2010	Q	M	-3.6	-9.0	-0.00038	-0.00064	0.0248 *	0.1808	0.2336
01.01.2008	31.12.2011	Q	M	-3.0	-5.4	-0.00032	-0.00026	0.0626 .	0.5860	0.0391
01.01.2009	30.09.2012	Q	M	-3.4	-8.5	-0.00036	-0.00061	0.0323 *	0.1569	0.3102
01.01.2003	31.12.2006	M	M	6.8	-8.1	0.00063	-0.00153	0.2010	0.2560	0.8159
01.01.2004	31.12.2007	M	M	-4.0	-25.1	-0.00042	-0.00352	0.2535	0.0002 ***	4.4090
01.01.2005	31.12.2008	M	M	3.1	-17.9	0.00030	-0.00267	0.3281	0.0005 ***	2.6800
01.01.2006	31.12.2009	M	M	-9.7	-20.2	-0.00111	-0.00169	0.0001 ***	0.0135 *	1.0780
01.01.2007	31.12.2010	M	M	-8.6	-16.8	-0.00097	-0.00122	0.0001 ***	0.0490 *	0.6413
01.01.2008	31.12.2011	M	M	-5.9	-11.1	-0.00064	-0.00066	0.0024 ***	0.2202	0.2427
01.01.2009	30.09.2012	M	M	-8.0	-12.1	-0.00089	-0.00055	0.0000 ***	0.2590	0.2109

Note: $p < .1$, * $p < .05$, ** $p < .01$, *** $p < .001$.

198 **Appendix A. Control regressions**

199 The control variables are data series for gas oil, coal, natural gas, emission rights, stock returns,
200 reservoir level, trading volume and skewness and variance in spot price. Gas oil is taken as daily
201 returns on the 1st position Gas Oil Futures traded at ICE from the Reuters EcoWin Pro database.
202 Coal is taken as the daily returns on the 1st position in a monthly rollover series, API2 Atlantic
203 Basin CIF, provided by Statoil. Natural gas is taken as daily returns on the ICE Natural Gas
204 Index, from Reuters EcoWin Pro. Emission rights are taken as the 1st position in a yearly rollover
205 series, the European Union Allowances Carbon Dioxide Yearly Rollover Series Argus Mid. The
206 emission rights have only been traded in the market since 2005, consequently this time series is
207 limiting the regressions time span. As a proxy for general market return we have used stock returns
208 from the OMXS30 index for the Nordic area and the DAX30 index for the German and Austrian
209 area, both from Reuters EcoWin Pro. For the reservoir level we calculated the daily difference
210 between average reservoir level and the historical reservoir level, from Reuters EcoWin. Both
211 the average and historical daily numbers were linearly interpolated from weekly numbers, using 7
212 days a week. Trading volumes for the future contracts are extracted from Montel. Skewness and
213 variance are calculated using a 90 days historical rolling window on the Nord Pool system spot
214 price for the Nordic area and the Phelix system spot price for the German/Austrian area. This is
215 an approximation to the skewness and variance variables introduced by Bessembinder and Lemmon
216 (2002) which calculates the variable on the deviation from expected spot price.

Nordic area				
Variable	Coeff.	SE	t-stat	Pr(> t)
Front dummy	-0.00093	0.00031	-2.99	0.0028 **
Stock returns, OMXS30	0.09297	0.00747	12.44	0.0000 ***
Δ Trading volume ¹	0.02108	0.01088	1.94	0.0527 .
Water value	-0.00435	0.00113	-3.85	0.0001 ***
Spot price variance ¹	-0.02228	0.00989	-2.25	0.0243 *
Spot price skewness	0.00020	0.00008	2.41	0.0160 *
Δ Reservoir level	-0.00162	0.00058	-2.82	0.0049 **
Return on gas oil	0.17197	0.00608	28.29	0.0000 ***
Return on coal	0.15369	0.00651	23.60	0.0000 ***
Return on emission	0.06475	0.00216	30.00	0.0000 ***
Return on natural gas	0.01422	0.00327	4.35	0.0000 ***
January	-0.00025	0.00041	-0.62	0.5350
February	-0.00131	0.00043	-3.07	0.0022 **
March	-0.00059	0.00040	-1.50	0.1332
April	0.00169	0.00042	4.04	0.0001 ***
May	0.00124	0.00043	2.91	0.0037 **
June	0.00068	0.00043	1.59	0.1125
July	-0.00121	0.00042	-2.86	0.0042 **
August	0.00128	0.00041	3.11	0.0019 **
September	-0.00126	0.00040	-3.20	0.0014 **
October	0.00053	0.00041	1.29	0.1976
November	-0.00170	0.00041	-4.19	0.0000 ***
December	0.00136	0.00046	2.98	0.0029 **

Note . $p < .0.1$, * $p < .05$, ** $p < .01$, *** $p < .001$. ¹The variable is scaled by 10,000.

Table A.4: Results from regression $r_t = \sum_{i=1}^n (\alpha_i \cdot X_i) + \epsilon_t$ (6) run on forward data for the Nordic area. Start date is 1 January 2005, end date is 28 September 2012. All contract types (month, quarter, year) are pooled.

German/Austrian area

Variable	Coeff.	SE	t-stat	Pr(> t)
Front dummy	-0.00068	0.00020	-3.36	0.0008 ***
Stock returns, DAX	0.03235	0.00546	5.93	0.0000 ***
Spot price variance ¹	-0.00109	0.00375	-0.29	0.7702
Spot price skewness	0.00015	0.00007	2.34	0.0192 *
Return on gas oil	0.09621	0.00398	24.16	0.0000 ***
Return on coal	0.12610	0.00425	29.67	0.0000 ***
Return on emission	0.06868	0.00146	47.22	0.0000 ***
Return on natural gas	0.01242	0.00214	5.81	0.0000 ***
January	-0.00137	0.00029	-4.73	0.0000 ***
February	-0.00103	0.00029	-3.59	0.0003 ***
March	0.00014	0.00028	0.51	0.6075
April	0.00200	0.00027	7.42	0.0000 ***
May	-0.00012	0.00027	-0.45	0.6530
June	0.00096	0.00026	3.61	0.0003 ***
July	-0.00163	0.00027	-6.07	0.0000 ***
August	0.00005	0.00028	0.19	0.8528
September	-0.00090	0.00027	-3.35	0.0008 ***
October	0.00065	0.00028	2.34	0.0192 *
November	-0.00151	0.00027	-5.65	0.0000 ***
December	0.00096	0.00030	3.17	0.0015 **

Note $p < .0.1$, $*p < .05$, $**p < .01$, $***p < .001$. ¹The variable is scaled by 10,000.

Table A.5: Results from regression $r_t = \sum_{i=1}^n (\alpha_i \cdot X_i) + \epsilon_t$ (6) run on forward data for the German/Austrian area. Start date is 1 January 2005, end date is 28 September 2012. All contract types (month, quarter, year) are pooled.

- 217 Benth, F. E., R. Biegler-König, and R. Kiesel (2013). An empirical study of the information
218 premium on electricity markets. *Energy Economics* 36, 55–77.
- 219 Benth, F. E., A. Cartea, and R. Kiesel (2008). Pricing the forward contracts in power markets by
220 the certainty equivalence principle: Explaining the sign of the market risk premium. *Journal of*
221 *Banking and Finance* 32(10), 2006–2021.
- 222 Bessembinder, H. and M. L. Lemmon (2002). Equilibrium pricing and optimal hedging in electricity
223 forward markets. *The Journal of Finance* 57(3), 1347–1382.
- 224 Botterud, A., T. Kristiansen, and M. D. Ilic (2010). The relationship between spot and futures
225 prices in the Nord Pool electricity market. *Energy Economics* 32(5), 967–978.
- 226 Bunn, D. W. and D. Chen (2013). The forward premium in electricity futures. *Journal of Empirical*
227 *Finance* 23, 173–186.
- 228 Cartea, Á. and P. Villaplana (2014). An analysis of the main determinants of electricity forward
229 prices and forward risk premia. In F. E. Benth, V. A. Kholodnyi, and P. Laurence (Eds.),
230 *Quantitative Energy Finance*, pp. 215–236. Springer.
- 231 Fama, E. F. and K. R. French (1987). Commodity futures prices: Some evidence on forecast power,
232 premiums, and the theory of storage. *Journal of Business* 60(1), 55–73.
- 233 Frestad, D., F. E. Benth, and S. Koekebakker (2010). Modelling term structure dynamics in the
234 Nordic electricity swap market. *The Energy Journal* 31(2), 53–86.
- 235 Geman, H. and O. Vasicek (2001). Plugging into electricity. *Risk Magazine August*, 93–97.
- 236 Haugom, E., G. A. Hoff, M. Mortensen, P. Molnár, and S. Westgaard (2014). The forecasting power
237 of medium-term futures contracts. *Journal of Energy Markets* 7(4), 1–23.
- 238 Haugom, E. and C. J. Ullrich (2012). Market efficiency and risk premia in short-term forward
239 prices. *Energy Economics* 34(6), 1931–1941.
- 240 Huisman, R. and M. Kilic (2013). A history of European electricity day-ahead prices. *Applied*
241 *Economics* 45(18), 2683–2693.

- 242 Longstaff, F. A. and A. W. Wang (2004). Electricity Forward Prices: A High-Frequency Empirical
243 Analysis. *Journal of Finance* 59(4), 1877–1900.
- 244 Lucia, J. J. and H. Torró (2011). On the risk premium in Nordic electricity futures prices. *Inter-*
245 *national Review of Economics and Finance* 20, 750–763.
- 246 Marckhoff, J. and J. Wimschulte (2008). Locational price spreads and the pricing of contracts for
247 difference: Evidence from the Nordic market. *Energy Economics* 31, 257–268.
- 248 Mirza, F. M. and O. Bergland (2012). Pass-through of wholesale price to the end user retail price
249 in the Norwegian electricity market. *Energy Economics* 34(6), 2003–2012.
- 250 Redl, C. and D. W. Bunn (2013). Determinants of the premium in forward contracts. *Journal of*
251 *Regulatory Economics* 43(1), 90–111.
- 252 Redl, C., R. Haas, C. Huber, and B. Böhm (2009). Price formation in electricity forward markets
253 and the relevance of systematic forecast errors. *Energy Economics* 31(3), 356–364.
- 254 Sanda, G. E., E. T. Olsen, and S.-E. Fleten (2013). Selective hedging in hydro-based electricity
255 companies. *Energy Economics* 40, 326–338.
- 256 Sandsmark, M. and B. Tennbakk (2010). Ex post monitoring of market power in hydro dominated
257 electricity markets. *Energy Policy* 38(3), 1500–1509.
- 258 Shawky, H. A., A. Marathe, and C. L. Barrett (2003). A first look at the empirical relation between
259 spot and futures electricity prices in the United States. *Journal of Futures Markets* 23(10),
260 931–955.
- 261 Veka, S. (2013). Additional evidence on the forward premium in the Nordic electricity market.
262 *Working Paper, Lillehammer University College and the Norwegian University of Science and*
263 *Technology*.
- 264 Weron, R. (2008). Market price of risk implied by Asian-style electricity options and futures. *Energy*
265 *Economics* 30(3), 1098–1115.
- 266 Weron, R. and M. Zator (2014). Revisiting the relationship between spot and futures prices in the
267 Nord Pool electricity market. *Energy Economics* 44, 178–190.