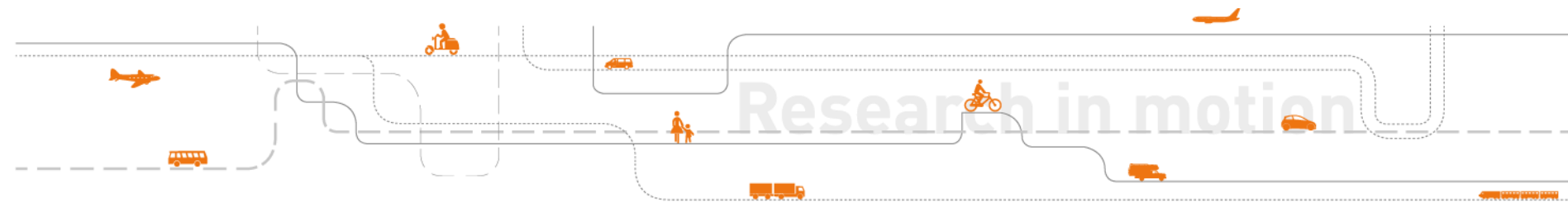


# Good Chargers? The impact of electric vehicle density on local grid costs

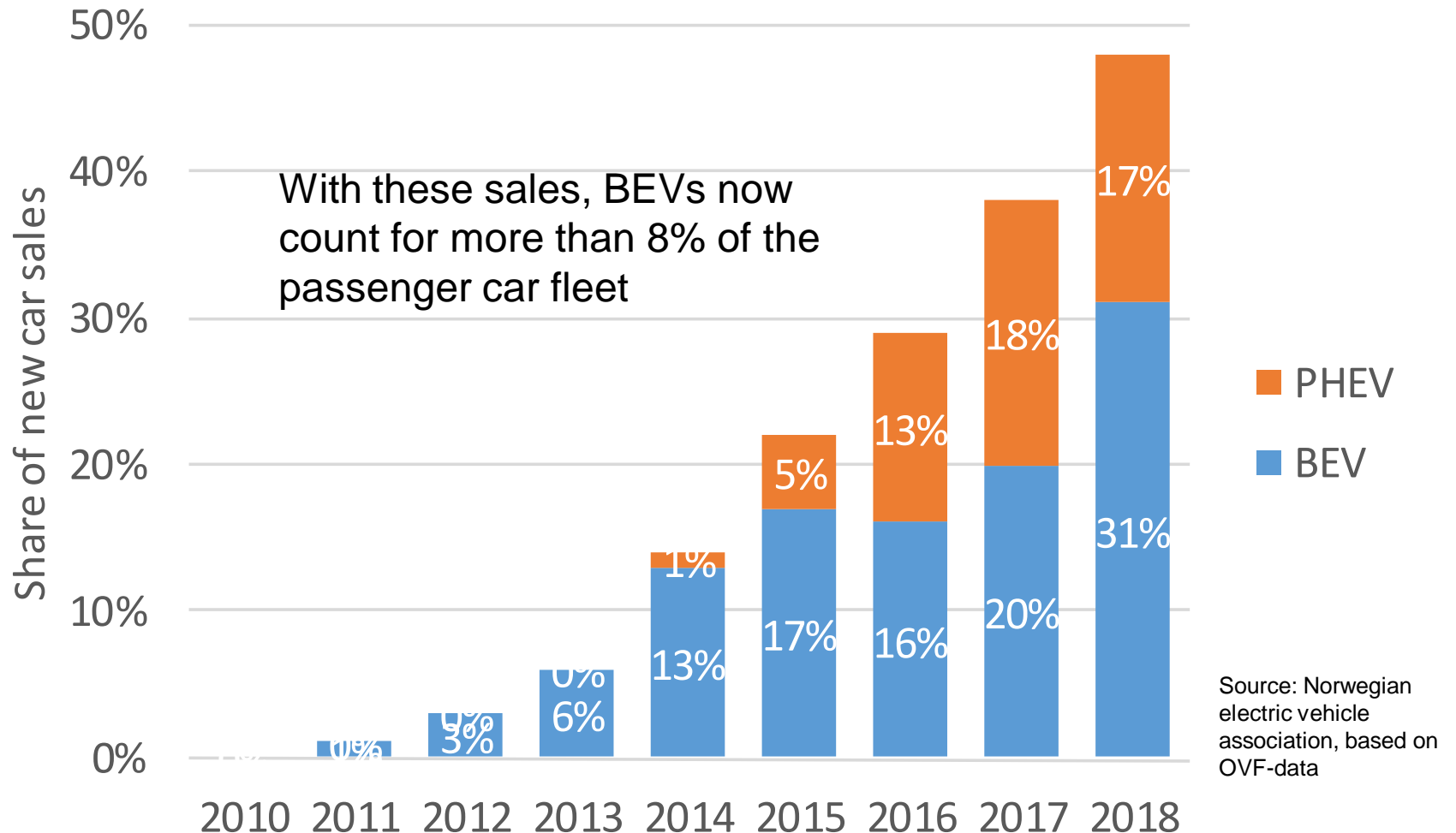
Paal B. Wangsness (TØI/NMBU)

Askill H. Halse (TØI)

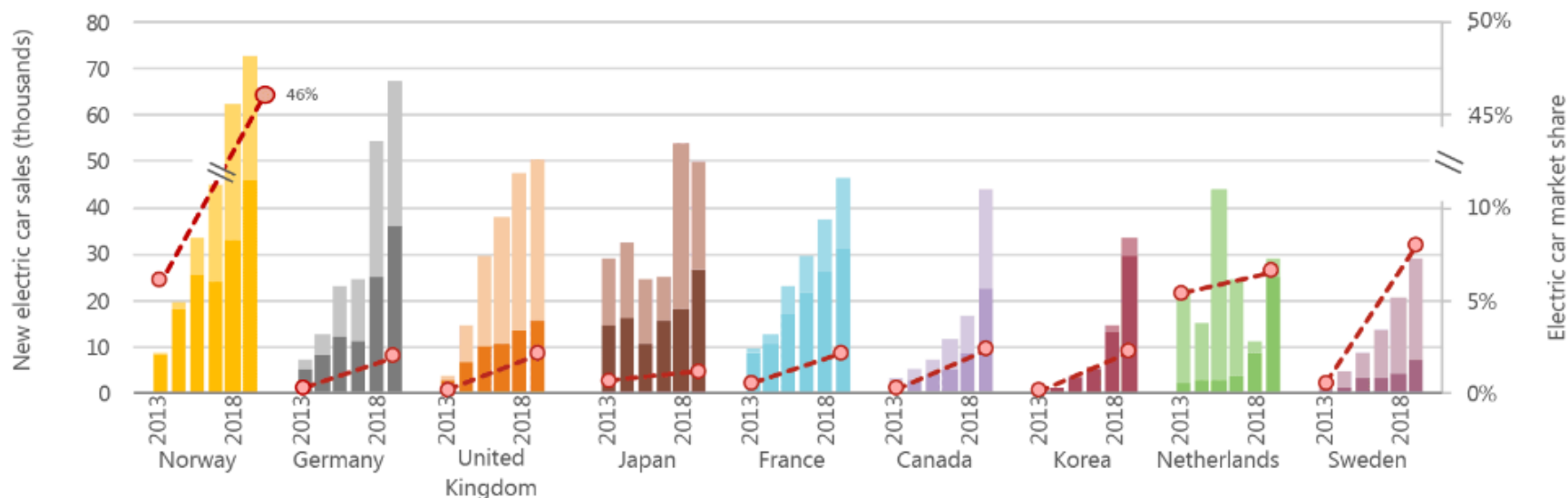


# MOTIVATION AND BACKGROUND

# Almost half of all new cars in Norway were plug-in electric in 2018

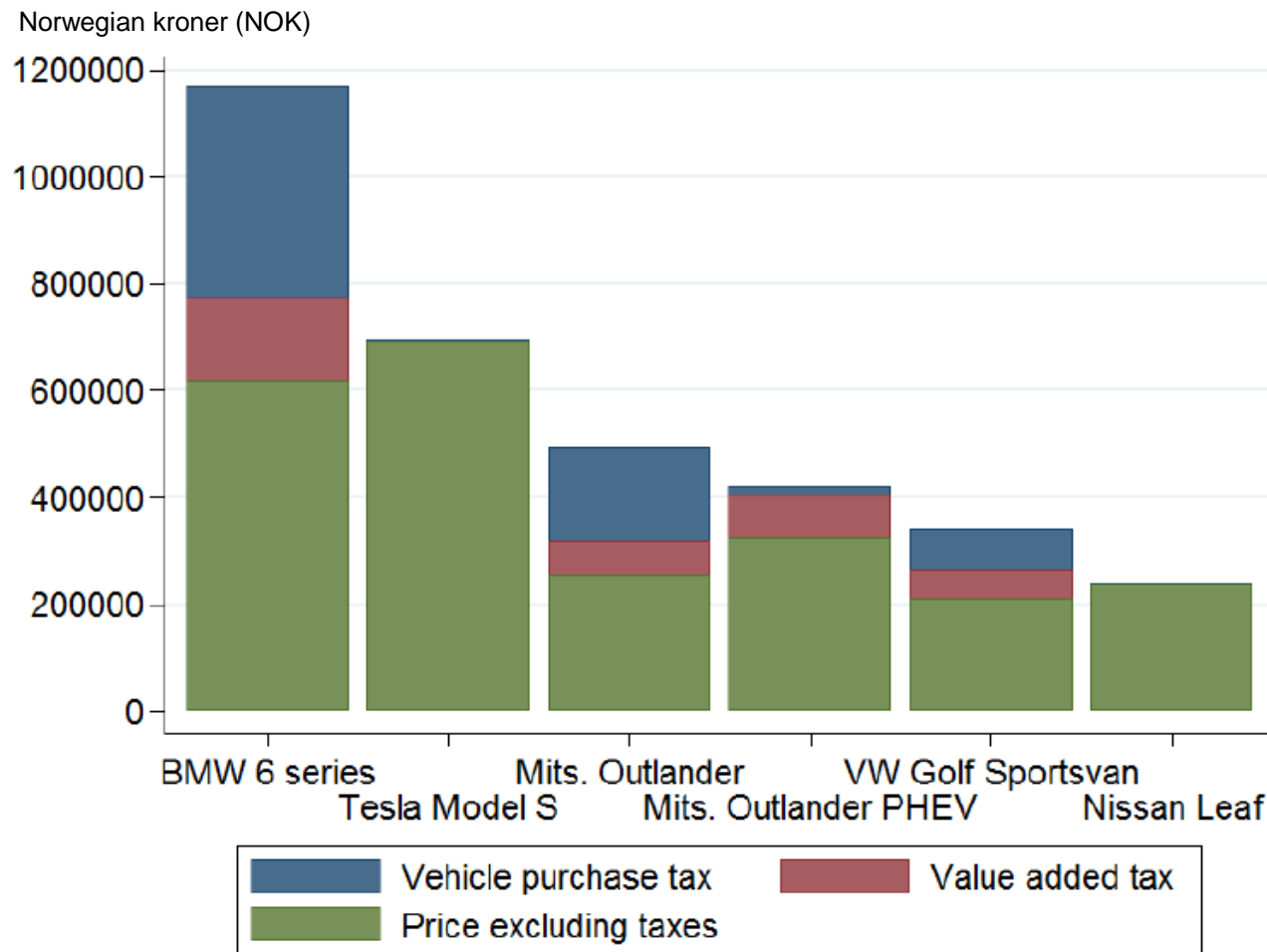


# This gives Norway the highest market share of EVs in the world, and the highest absolute number of EVs in Europe



Source: IEA (2019), "Global EV Outlook 2019": IEA analysis based on country submissions, complemented by ACEA (2019); EAFO (2019); EV Volumes (2019); Marklines (2019); OICA (2019)

# The main reason: Conventional cars are heavily taxed



Source: Johansen (2019), based on OVF-data

# EVs needed to reach climate goals, but can lead to higher grid costs and tariffs



# There exists a literature that warns that EV charging will cause substantial future costs to the local grid

Examples:

De Hoog et al (2015). Optimal charging of electric vehicles taking distribution network constraints into account. *IEEE Transactions on Power Systems*

Haidar et al (2014). Technical challenges for electric power industries due to grid-integrated electric vehicles in low voltage distributions: A review. *Energy Conversion and Management*

Hattam et al (2017). Green neighbourhoods in low voltage networks: measuring impact of electric vehicles and photovoltaics on load profiles. *Journal of Modern Power Systems and Clean Energy*

# Our motivation and contribution

- If indeed uncoordinated charging leads to higher costs to DSOs, then Norwegian data would be the first place to investigate.
- To our knowledge, such an empirical analysis has not been done before on real data on the world's highest level of EV density to date.
- Implications for regulation, pricing and assessing GHG abatement costs



# Based on this motivation, our research questions are:

1. What are the *marginal costs* inflicted on DSOs when the number of EVs increases?
2. Through which *mechanisms*, i.e. which of the DSOs cost components, do we find the cost associated with a larger EV stock?

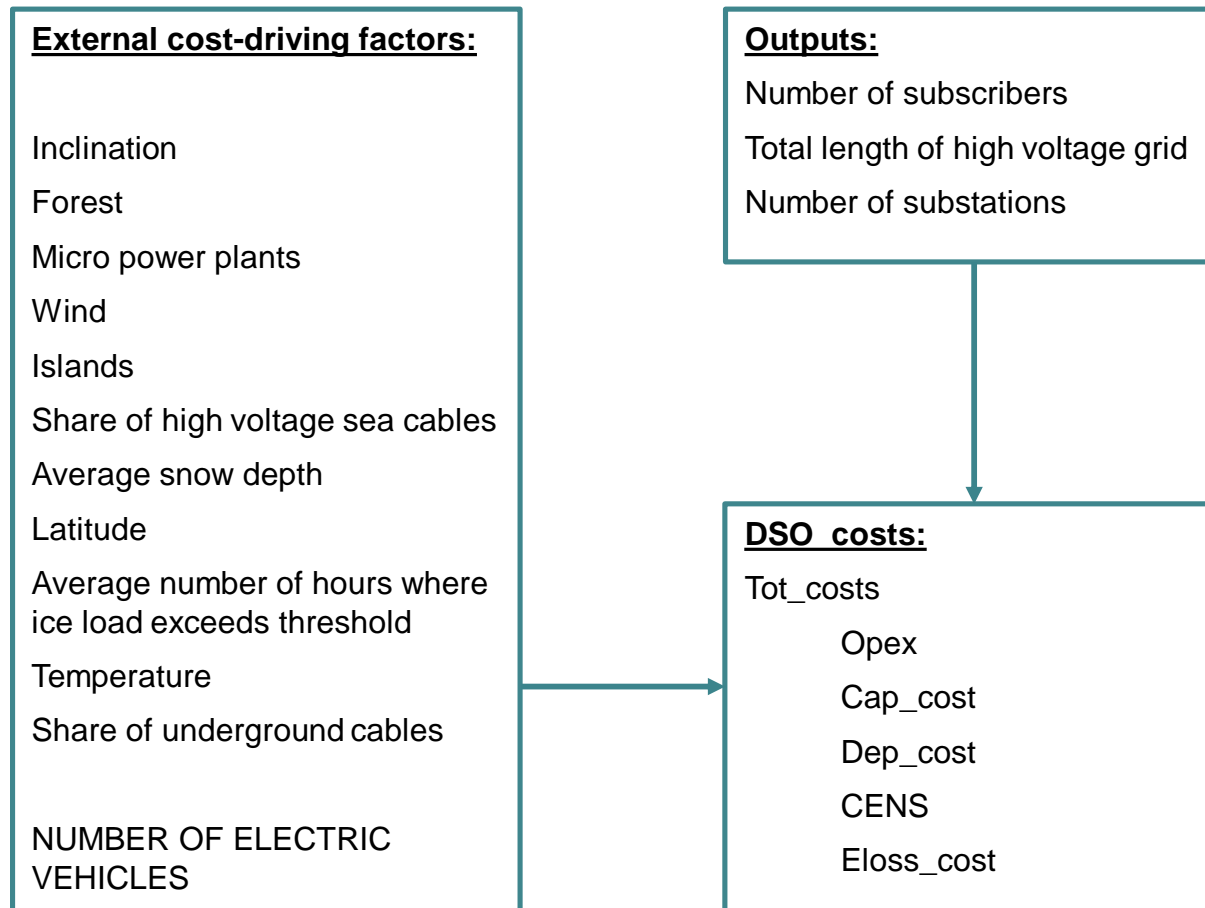
# DATA

# Panel dataset of 107 DSOs for the years 2008-2017 – 1070 observations

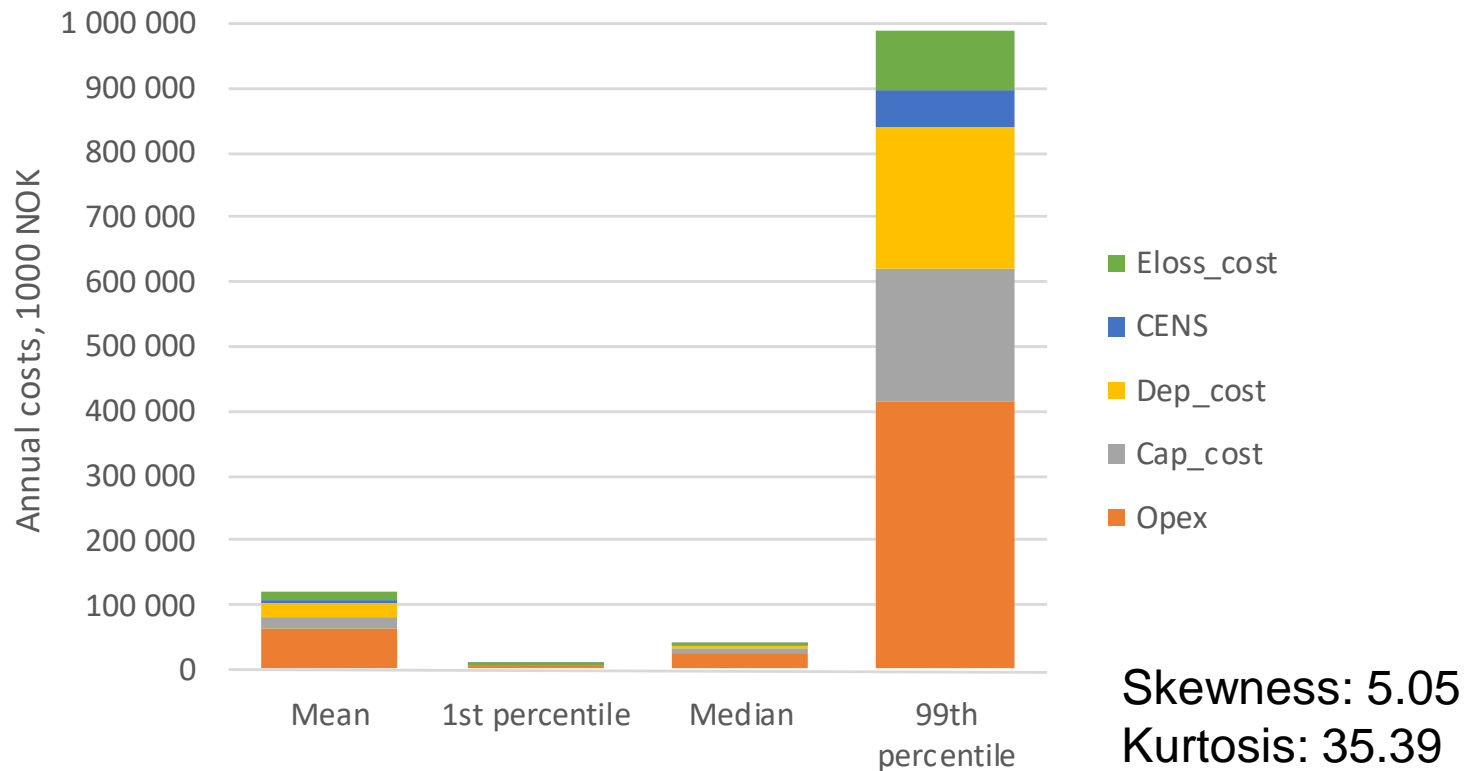
We match and merge together 3 datasets:

- 1) NVE's data for DSO costs and outputs applied for regulation
- 2) NVE's data for the DSOs legal operational area
- 3) Statistics Norway's data over registered cars at municipal level

# The variables, and the direction of impacts from outputs and external cost-driving factors to costs



# Descriptive statistics: DSO costs



We transform the variables to log-form:  
Skewness: 0.83  
Kurtosis: 3.67

# METHOD

# A fixed effects regression model on a balanced panel

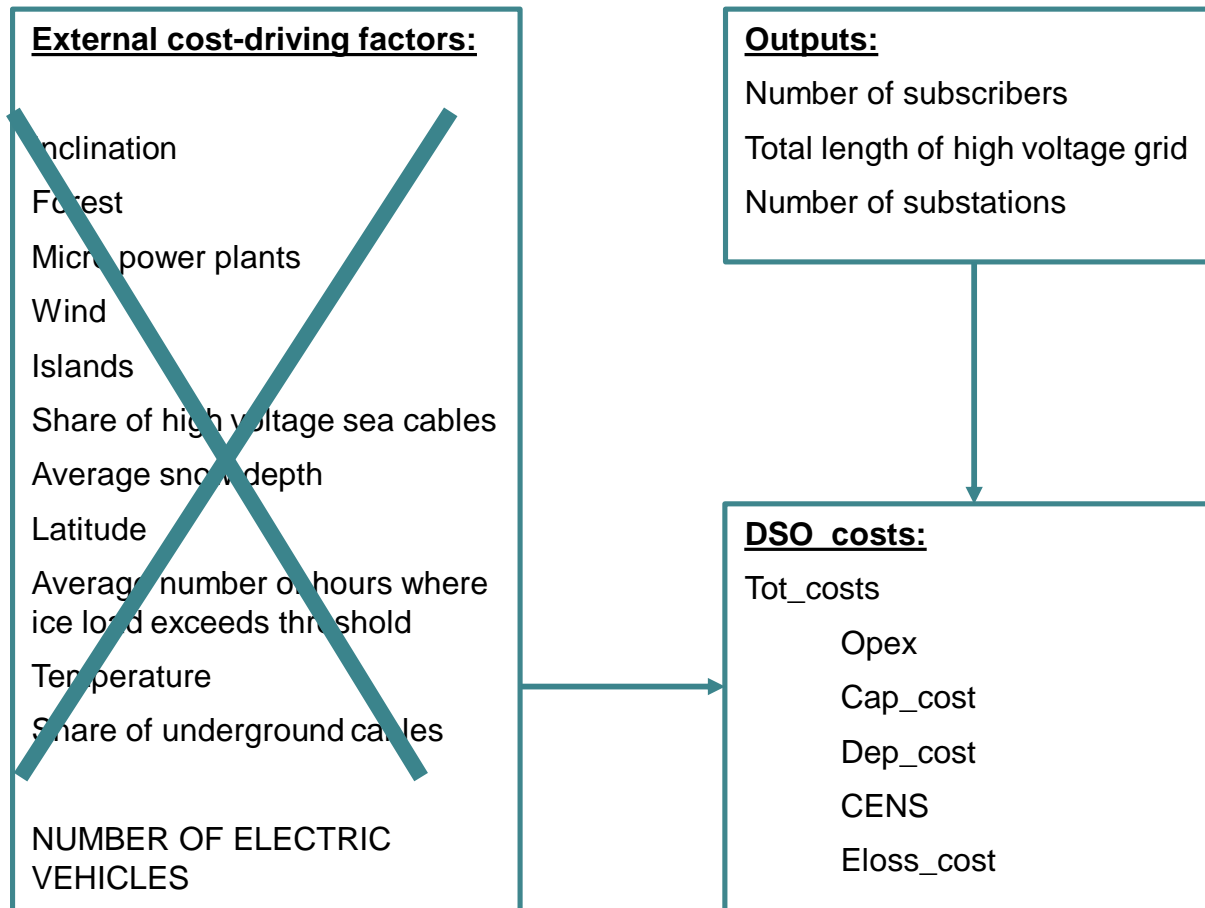
- Avoid issues of unmeasured time-invariant variables that we expect have an effect on both our explanatory variable of interest and the endogenous variable
  - *E.g. temperature, sprawl etc.*
- FE reduces the risk of omitted variable bias, making it more likely that the relationship we infer between EV density and DSO costs to be causal

# *A priori* reasons to believe that EV density can be considered an exogenous regressor

- Even if higher EV density led to higher costs for DSOs and higher grid rent, dramatic price hikes would be needed to make noticeable changes in EV demand.
- Electricity costs make about 15% of the distance-based cost for EVs, and grid rent makes up less than half of the bill before taxes.
- Not certain that the DSO can pass on all of their cost increase to their customers due to regulation.
- In other words, we expect EVs to affect grid costs, and have very little feedback the other way around.



# We are left with a few time-variant variables



# Model specification

$$\log\_tot_{it} = \alpha_i + \beta_1 \log\_subscribe_{it} + \beta_2 \log\_voltline_{it} + \beta_3 \log\_EV_{it} + \beta_4 \log\_substations_{it} + \beta_5 event_{it} + \beta_6 Year\_Dummy_t + \varepsilon_{it}$$

- Log\_tot: Log of total costs
- Log\_subscribe: Log of number of subscribers
- Log\_voltline: Log of km of high voltage line
- Log\_EV: Log of EV stock
- Event: Number of extreme weather events
- Log\_substations: Log of number of substations

# RESULTS

# The effect of EV density on costs seems to be economically significant, but imprecisely estimated

	(1) log_tot	(2) log_tot	(3) log_tot	(4) log_tot	(5) log_tot (removed outliers)
log_subscribe	0.338* (0.192)	0.222 (0.247)	0.289 (0.243)	0.276 (0.245)	0.250 (0.254)
log_voltline	0.318** (0.147)	0.305** (0.147)	0.293** (0.147)	0.287* (0.148)	0.271* (0.149)
event	0.007*** (0.002)	0.008*** (0.002)	0.004* (0.002)	0.004* (0.002)	0.004 (0.002)
log_ev	0.011*** (0.004)	0.007 (0.006)	0.011 (0.007)	0.010 (0.007)	0.007 (0.007)
year		0.004 (0.005)			
log_substation				0.025** (0.012)	
_cons	5.835*** (1.595)	-1.399 (8.079)	6.420*** (2.191)	6.417*** (2.197)	6.873*** (2.256)
Year dummies	No	No	Yes	Yes	Yes
N	1070	1070	1070	1070	1030
r2_w	0.212	0.214	0.279	0.279	0.289
r2_b	0.982	0.980	0.982	0.982	0.979
r2_o	0.977	0.975	0.977	0.977	0.973
rho	0.948	0.975	0.968	0.968	0.973

Standard errors clustered at DSO level in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

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# The economic significance

- The median number of customers were 7282, they had in total 78 EVs and total costs were 44 mill NOK for 2017
- Let's say the number of EVs doubles, *ceteris paribus*
- Model predicts a cost increase of 462 000 NOK (46 200 €) or about 5920 NOK per EV

# Heterogeneity test with sample splits along 3 dimensions

	(1) log_tot Lower half customers	(2) log_tot Upper half customers	(3) log_tot Lower half EV density	(4) log_tot Upper half EV density	(5) log_tot Lower half cost per customer	(6) log_tot Upper half costs per customer
log_subscribe	0.336 (0.315)	0.491 (0.361)	0.124 (0.304)	0.638** (0.263)	0.280 (0.270)	0.490 (0.377)
log_voltline	0.459*** (0.139)	-0.317 (0.351)	0.490*** (0.160)	-0.009 (0.240)	0.075 (0.288)	0.375** (0.142)
event	0.010*** (0.003)	0.001 (0.003)	0.001 (0.004)	0.006* (0.003)	0.003 (0.003)	0.006 (0.004)
log_ev	0.035*** (0.010)	0.003 (0.009)	0.025** (0.011)	0.002 (0.010)	0.011 (0.009)	0.027** (0.011)
_cons	4.828* (2.454)	8.931** (3.871)	4.755* (2.757)	7.849** (3.109)	8.060*** (2.713)	4.135 (3.008)
Year dummies	Yes	Yes	Yes	Yes	Yes	Yes
N	535	535	540	530	540	530
r2_w	0.157	0.094	0.289	0.311	0.251	0.343
r2_b	0.977	0.978	0.933	0.973	0.993	0.972
r2_o	0.974	0.975	0.921	0.966	0.987	0.964
rho	0.957	0.909	0.831	0.986	0.990	0.753

Standard errors clustered at DSO level in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

# EVs affect cost components positively but statistically insignificantly, with the exception of grid-losses

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	log_opex	log_cap	log_cens	log_depres	log_gridloss	log_regulat_assets	log_copaid_assets
log_subscribe	0.338 (0.386)	0.120 (0.380)	0.080 (1.210)	0.146 (0.372)	1.244*** (0.454)	0.120 (0.380)	-1.034 (1.125)
log_voltline	0.181 (0.202)	0.446** (0.217)	1.146 (0.751)	0.259 (0.272)	0.594*** (0.211)	0.446** (0.217)	1.348* (0.764)
event	0.006 (0.004)	0.005* (0.003)	0.007 (0.012)	-0.002 (0.005)	0.000 (0.006)	0.005* (0.003)	-0.006 (0.009)
log_ev	0.002 (0.010)	0.014 (0.011)	0.018 (0.030)	0.010 (0.013)	-0.034** (0.015)	0.014 (0.011)	0.019 (0.035)
_cons	6.104* (3.311)	5.177 (3.399)	-0.698 (10.190)	6.018 (3.976)	-6.576* (3.878)	7.775** (3.399)	10.752 (9.774)
<i>N</i>	1070	1070	1070	1070	1070	1070	1068
r2_w	0.134	0.833	0.125	0.507	0.195	0.687	0.543
r2_b	0.962	0.935	0.917	0.965	0.957	0.935	0.049
r2_o	0.951	0.916	0.809	0.942	0.940	0.923	0.076
rho	0.938	0.974	0.350	0.976	0.968	0.974	0.964



# DISCUSSION

# A few things that have surprised us

- Economically significant, but not statistically significant
- The effect per EV seems to be diminishing as the EV density increases in a DSO's operational area
- A negative, but statistically significant impact on grid-losses

# The wide confidence interval and other caveats

- $\beta_3$  has a 95% CI of [-0.0028 , 0.0238]
- Would want more precision before including it into calculations for DSO regulation?
- We estimate the relationship between costs and *registered EVs*
  - *We do not have data on charging behavior*
  - *Costs may accrue elsewhere, e.g., in areas with many cabins*

# CONCLUSION

# We would not recommend EV density as a variable in the DSO regulation quite yet

- Statistically insignificant, but point estimates fairly large
- The effect per EV seems to be diminishing as the EV density increases
- Cautious optimist's interpretation: A shift from conventional cars to electric cars have social costs. However, as of now the costs are *not* in any statistically significant degree coming in the form of added cost to the local grid.

# Thank you for your attention!

For any questions or comments not covered in the following Q&A, you may reach me at [pbw@toi.no](mailto:pbw@toi.no)

I will also be happy to send you a complete working paper when it is finished in the next few months. Just let me know!