

# Applying transformer to imputation of multivariate energy time series data

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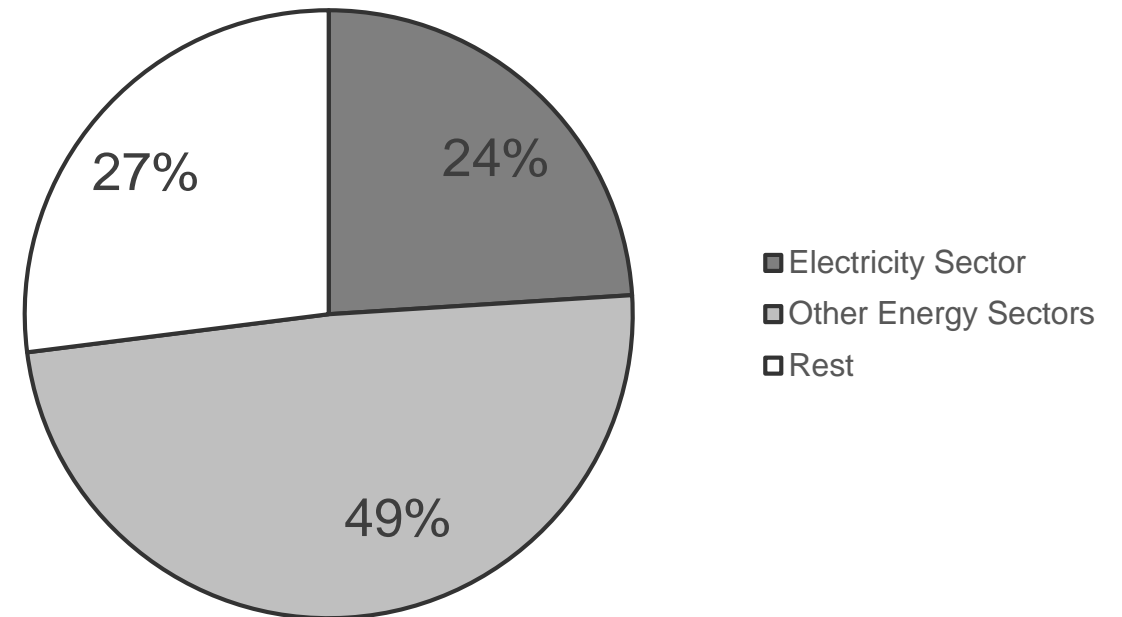
Chair of Energy Economics (Prof Fichtner)  
Research Group 'Energy Policy'



# Motivation

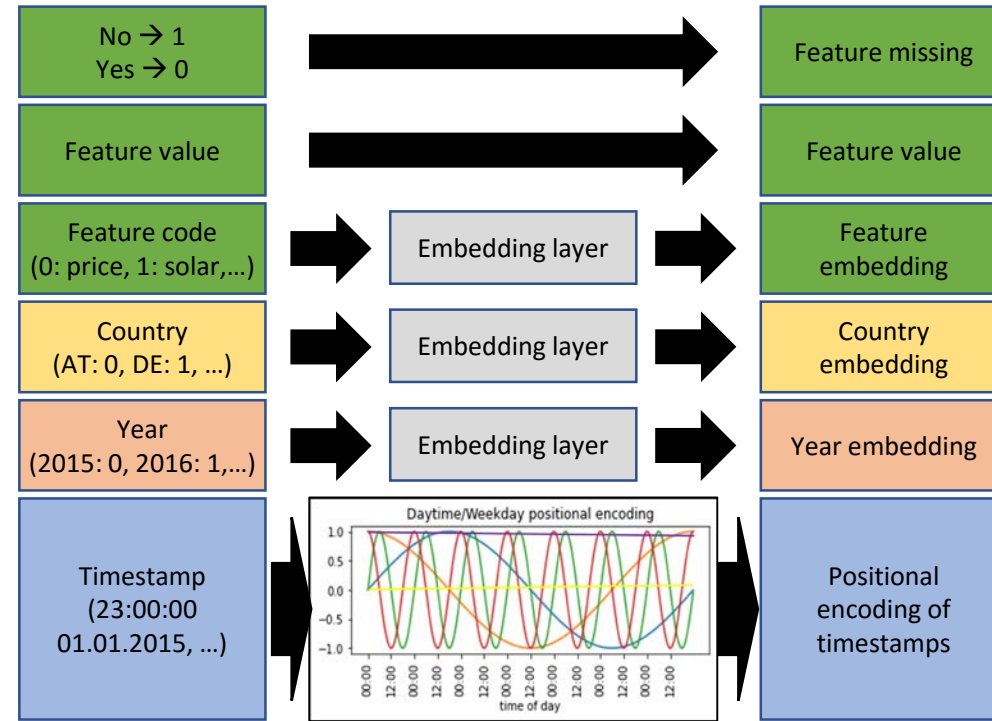
- Global greenhouse gas (GHG) emission
  - 24% electricity sector
  - 49% other energy sectors
- Sector coupling: use of renewable energy sources in other sectors
  - E.g. Electric cars, heat pumps
- Energy data has many data lacks
- Quality of the data matters in many analyses
- Solution:
  - Application of new innovative methods for data imputation: **Transformer**

Global GHG Emissions



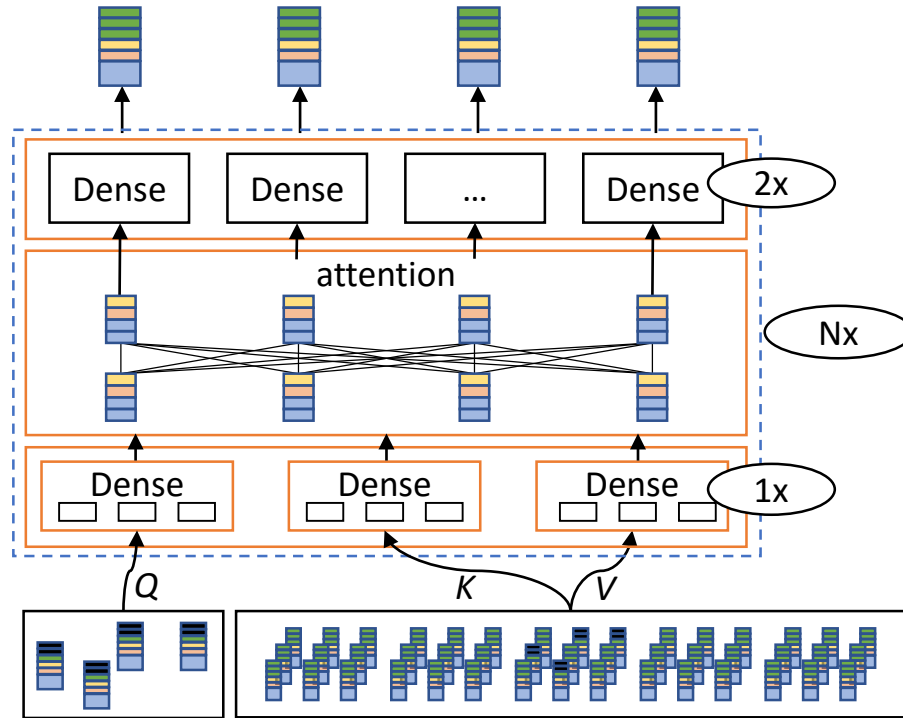
# Data processing

Year	Country	Hour	$S_1$	$S_2$	...	$S_n$
$Y_1$	$C_1$	$H_1$	.	.	.	.
		$H_2$	.	.	.	.
		...	...	...	...	...
		$H_{8760}$	.	.	.	.
	$C_2$	$H_1$	.	.	.	.
		$H_2$	.	.	.	.
		...	...	...	...	...
		$H_{8760}$	.	.	.	.

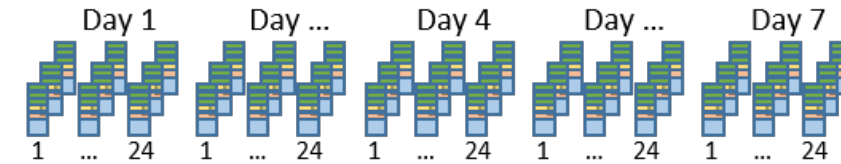


- Aggregated electricity prices, production and electricity demand data for the European countries
- Embedding layer: feature code, country and year
- Positional encoding: data on position of the information in the sequence

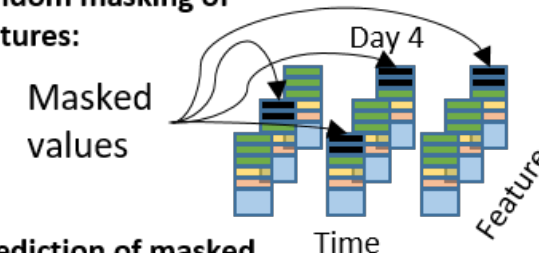
# Transformer model



Input sample:



Random masking of features:



Prediction of masked features:



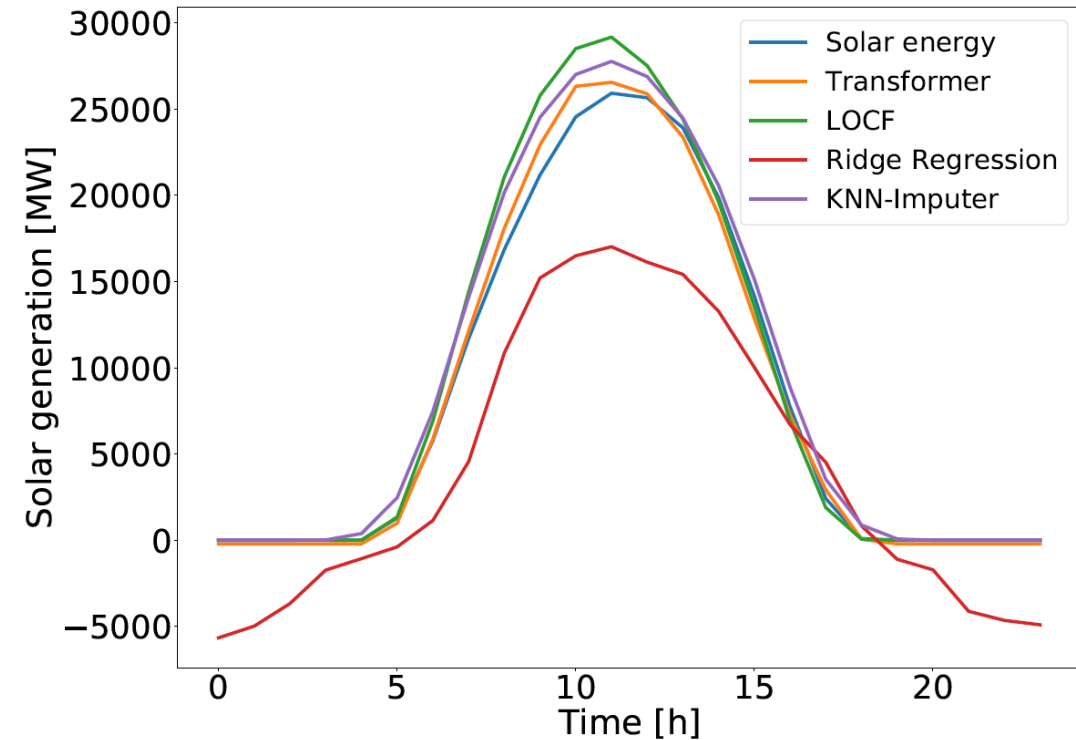
- Training process: mask  $x$  data entries of the middle day of the week
- Decomposition mechanism is necessary.

# Results

- Only with the data for Germany
- Simplified masking
- Methods for comparison:
  - Last observation carried forward (LOCF)
  - K-Nearest neighbors imputer (KNN-Imputer)
  - Ridge regression

## Conclusion:

- Transformer outperforms all other approaches.



	MSE
Transformer	0.0585
LOCF	0.1133
Ridge Regression	0.1827
KNN-Imputer	0.0629

**Thank you for the attention!**