

Smart power-to-heat concepts for district heating system based on data-energy network

Gefördert durch:



Bundesministerium
für Wirtschaft
und Energie

aufgrund eines Beschlusses
des Deutschen Bundestages





1. Background and Motivation
2. Methodology
 - a) P2H Concept for DHS
 - b) Data Analysis
 - c) Heat Load Forecasting
 - d) Implementation P2H Concept
3. Results

Background and Motivation

Heat Transition

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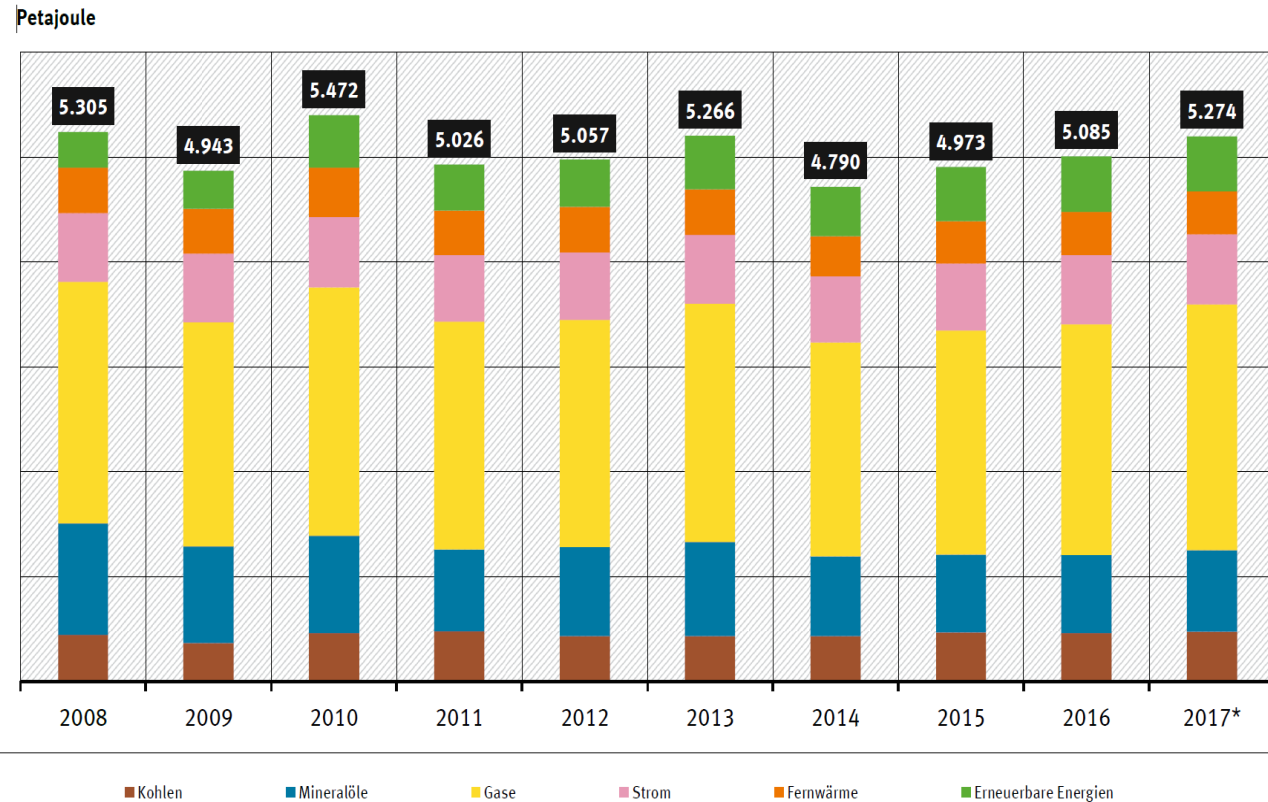


Figure 1: Heat consumption (incl. cooling applications) by energy source

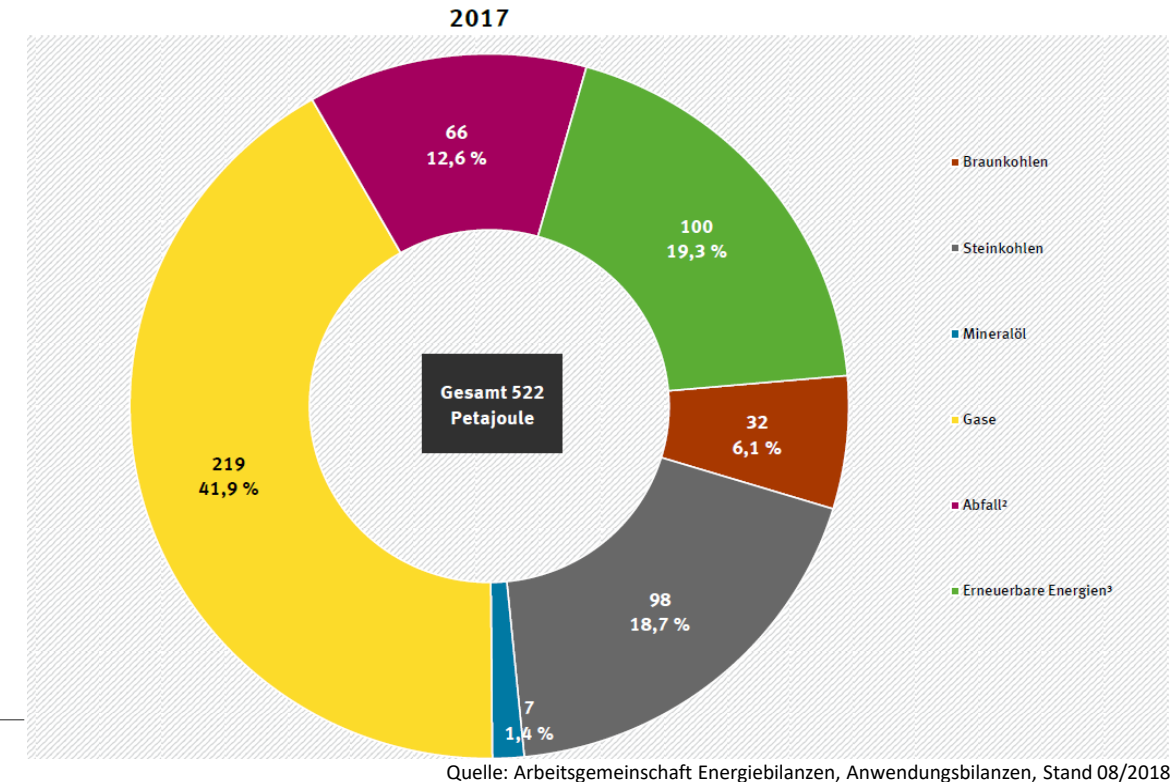


Figure 2: Energy source of district heating system (DHS)

Background and Motivation

P2H in Germany

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P2H in Germany

- At the beginning of 2019, 36 larger P2H modules with an electrical output of between **0.5 and 60 MW** were installed in Germany.
- The total output amounts to approximant **555 MW**.

Table 1: Example of P2H Modules in Operation in Germany

Betreiber	Leistung	Inbetriebnahme
BTB Berlin	6 MW	2015
Energieversorgung Offenbach	2x5 MW	2014
ENRO Ludwigsfelde	15 MW	2014
FHW Berlin-Neukölln	10 MW	2015
Heizkraftwerke Mainz	5 MW	2013
Mainova	8 MW	2015
N-ERGIE	50 MW	2015
Stadtwerke Flensburg	30 MW	2012
Stadtwerke Lemgo	5 MW	2012
Stadtwerke Schwerin	15 MW	2013
Stadtwerke Tübingen	5 MW	2014
VVS Saarbrücken	10 MW	2012

Quelle: Power-to-Heat plants in Germany, Power-to-Heat, <http://www.power-to-heat.eu/power-to-heat/>

Challenges of Implementation P2H in DHS

- Complex Options of Implementation P2H
- Cost-effective and Eco-friendly
- Combining P2H and Heat Load Forecasting

P2H Concept for DHS

Location

- Centralized
- Decentralized

Power source

- Renewable Power Plants: PV, Wind Power, Biomass
- Green Power from Grid

Conversion Technology

- Electrical Resistance Heating: Heat Rod, Electric Boiler
- Heat Pump

Storage

- Active: Internal, External
- Passive

Figure 3: P2H Concept in DHS

Example of DHS

- Biomass Cogeneration Plant (CHP) since 2012
- Biomass CHP supplies:
 - Electricity: 16 GWh/a** (4500 households)
 - Heat: 60 GWh/a** (4700 households)
- Line Length: 30 km

Annual Load Duration Curve

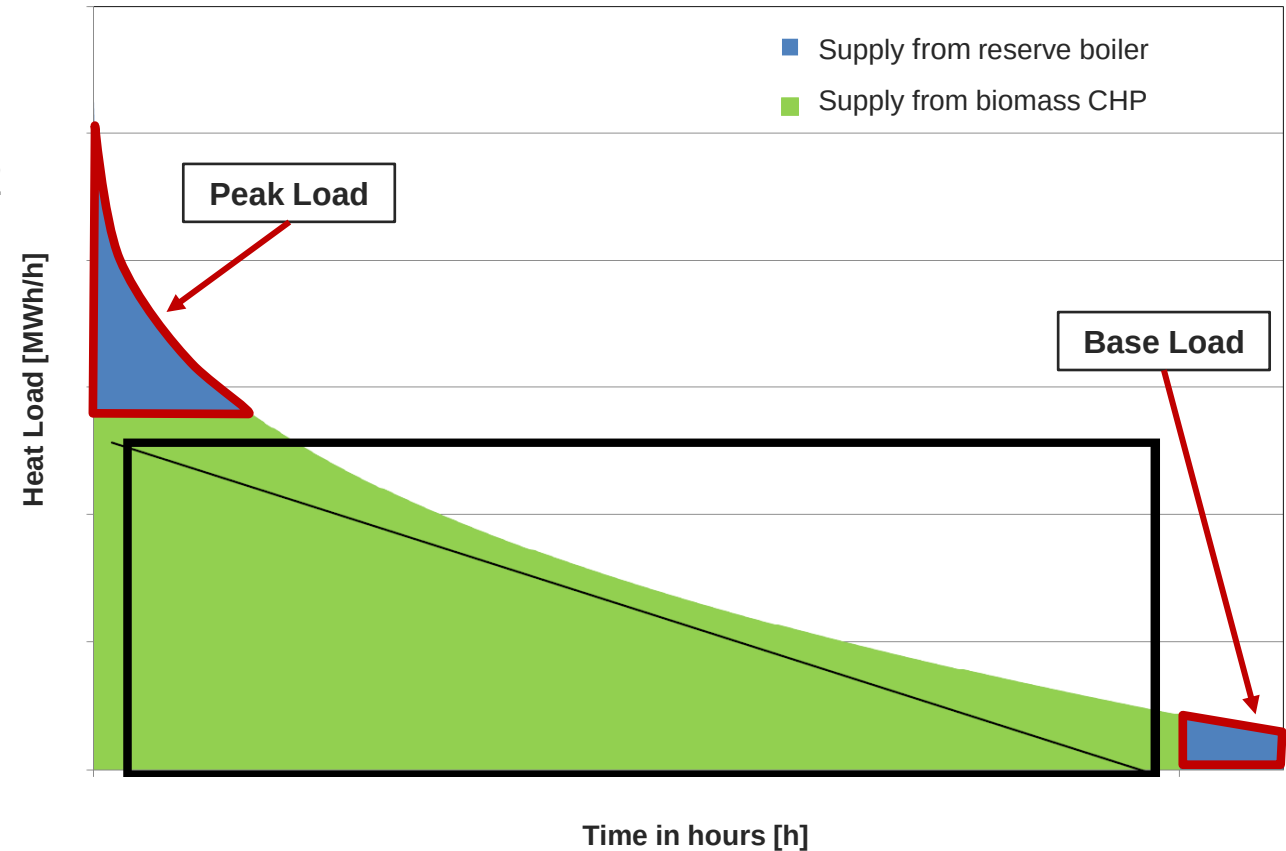


Figure 4: Annual load duration curve of DHS

Heat Load Characteristics

- Thermal Load inversely proportional to outdoor temperature
- Heat load in summer is 20%-30% in winter
- Heat load fluctuation in summer (about 10%) is much smaller than in winter (30%)
- Two peaks: 6:00 - 8:00, 18:00 - 21:00
- Two drops: 14:00 - 16:00, 22:00 - 04:00

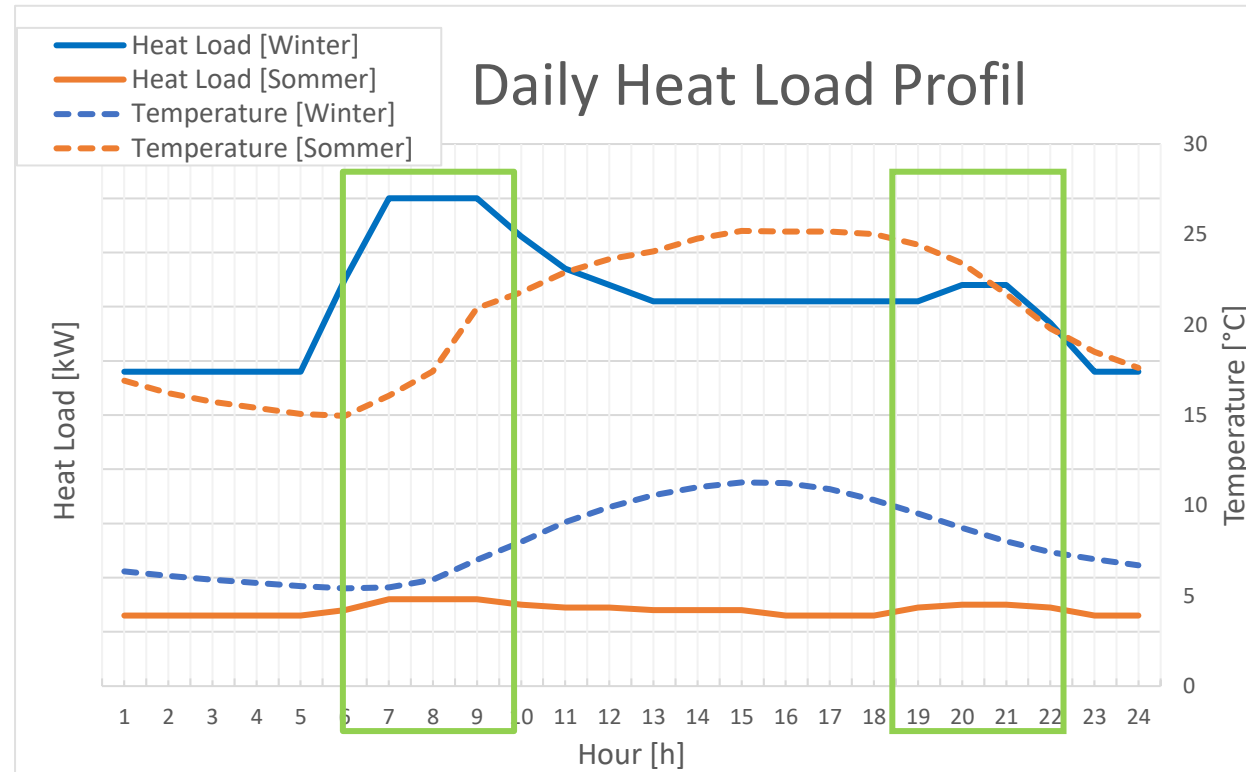


Figure 5: An Example Daily Heat Load Profile for DHS over One Year with Corresponding Outside Temperature

Methodology and P2H Concept

Heat Load Forecasting — Data Pre-Processing

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Most ML models **are not able to handle categorical variables**, so we choose the following cyclic encoding:

$$x_{\sin} = \sin\left(\frac{2 \times \pi \times t}{T}\right)$$

$$x_{\cos} = \cos\left(\frac{2 \times \pi \times t}{T}\right)$$

For $t \in T$, with $T = 12$ for months, $T = 7$ for weekdays and $T = 365$ for day in year.

Normalization of numerical values by following equation:

$$z_i = \frac{x_i - \min(x)}{\max(x) - \min(x)}$$

All numerical values are transformed to be between **min = 0 and max = 1**

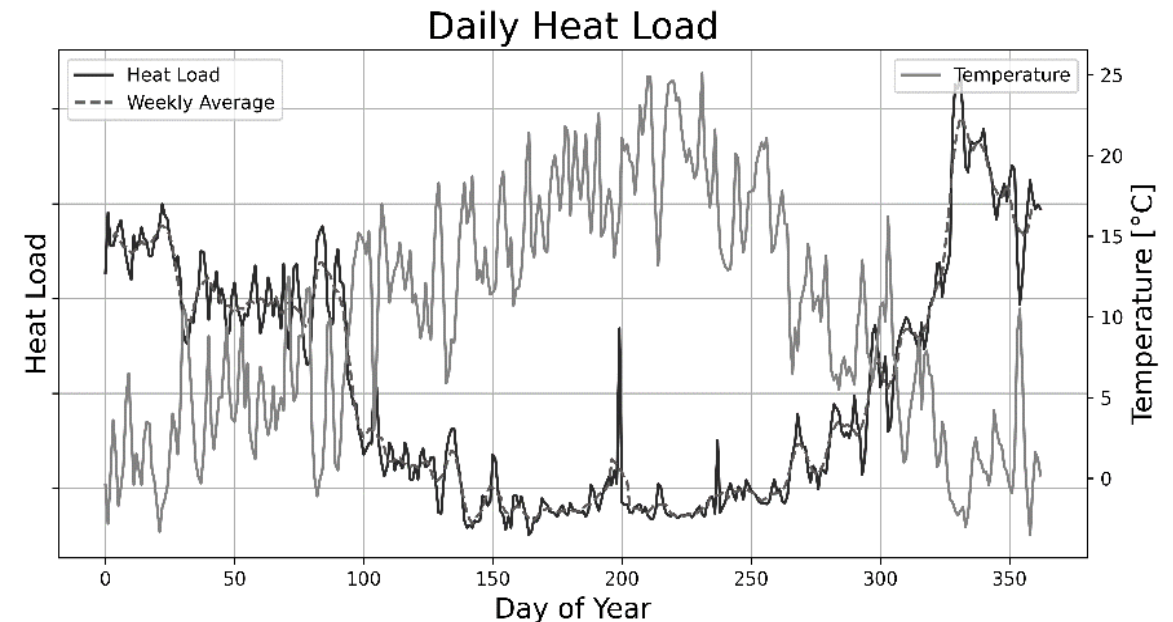


Figure 6: Daily heat load demand for one example building over one year with corresponding outside temperature.

Methodology and P2H Concept

Heat Load Forecasting — Model Development and Evaluation

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1.095 daily values: **80% for train and 20 % test**

Training phase:

- approximate a function $f: \mathbf{X} \rightarrow \mathbf{Y}$,
- $\mathbf{X}$ is composed of the features *yesterday's heat load, temperature, humidity, month, weekday, day of year*
- $\mathbf{Y}$ is the day-ahead prediction of the building's heat load.
- three models *Random Forest, Neural Network, Support Vector Machine*

trained models are evaluated with the remaining 20 % data

Model
evaluation
metrics

$$R^2 = \frac{\sum_{n=0}^N (\hat{y}_n - \bar{y}_n)}{\sum_{n=0}^N (y_n - \bar{y}_n)},$$

$$MAE = \frac{1}{N} \sum_{n=0}^N |\hat{y}_n - y_n|,$$

$$RMSE = \sqrt{\frac{\sum_{n=0}^N (\hat{y}_n - y_n)^2}{N}}.$$

Table 2: Model evaluation metrics on test set.

	Random Forest	Support Vector Machine	Neural Network
R^2	0.98	0.96	0.95
MAE	18.21	28.27	32.50
RMSE	34.75	49.25	52.51

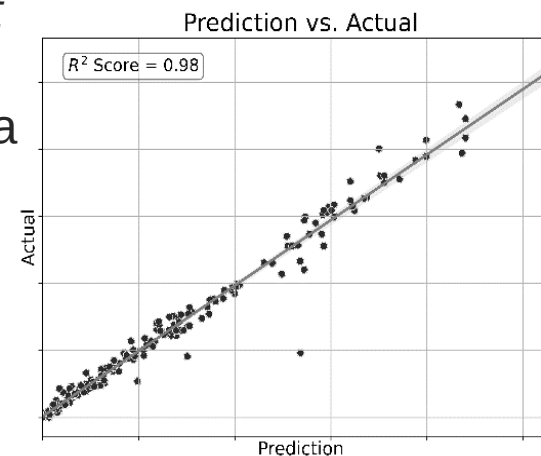


Figure 7: Accuracy of model prediction with an R^2 score of 0.98

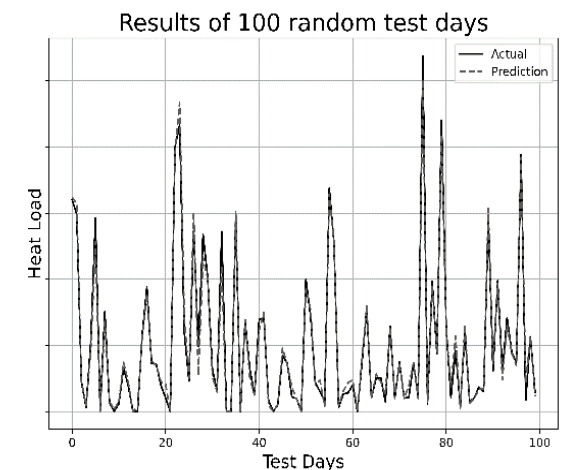


Figure 8: Comparison between actual heat load and predicted value for 100 random days

Methodology and P2H Concept

Implementation P2H Concept

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Decentralized systems:

- Purchase, installed, maintained cost
- Retrofitting difficulty
- Installed additional meters
- Space requirement
- Contract adjustment

=> Heat boilers, heat rods

Centralized systems:

- Heat losses in heat transfer

=> Superpower Heat pump, multiple small heat pumps

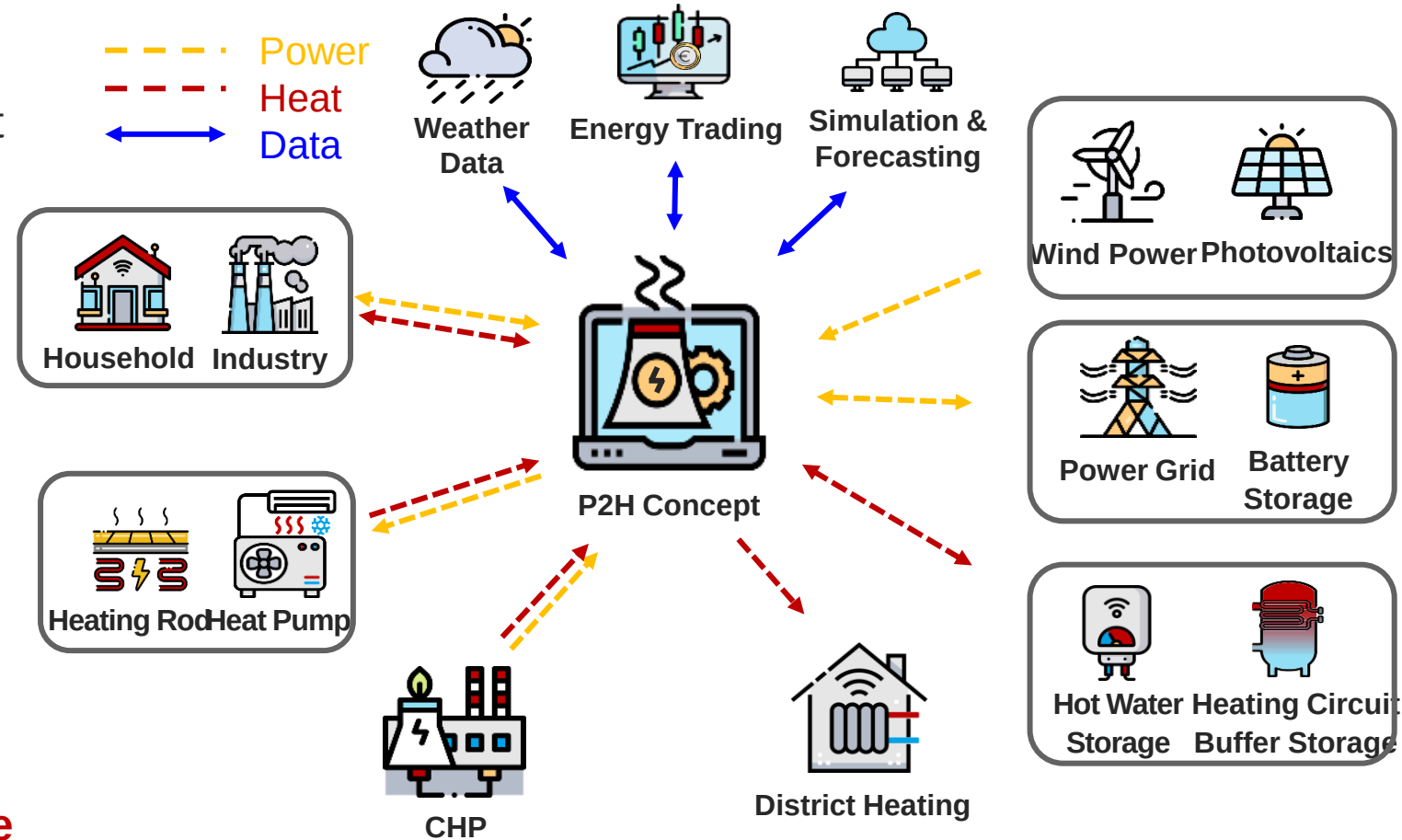


Figure 9: P2H Concept for DHS Including Data and Energy Flows

Results

Cost Analysis

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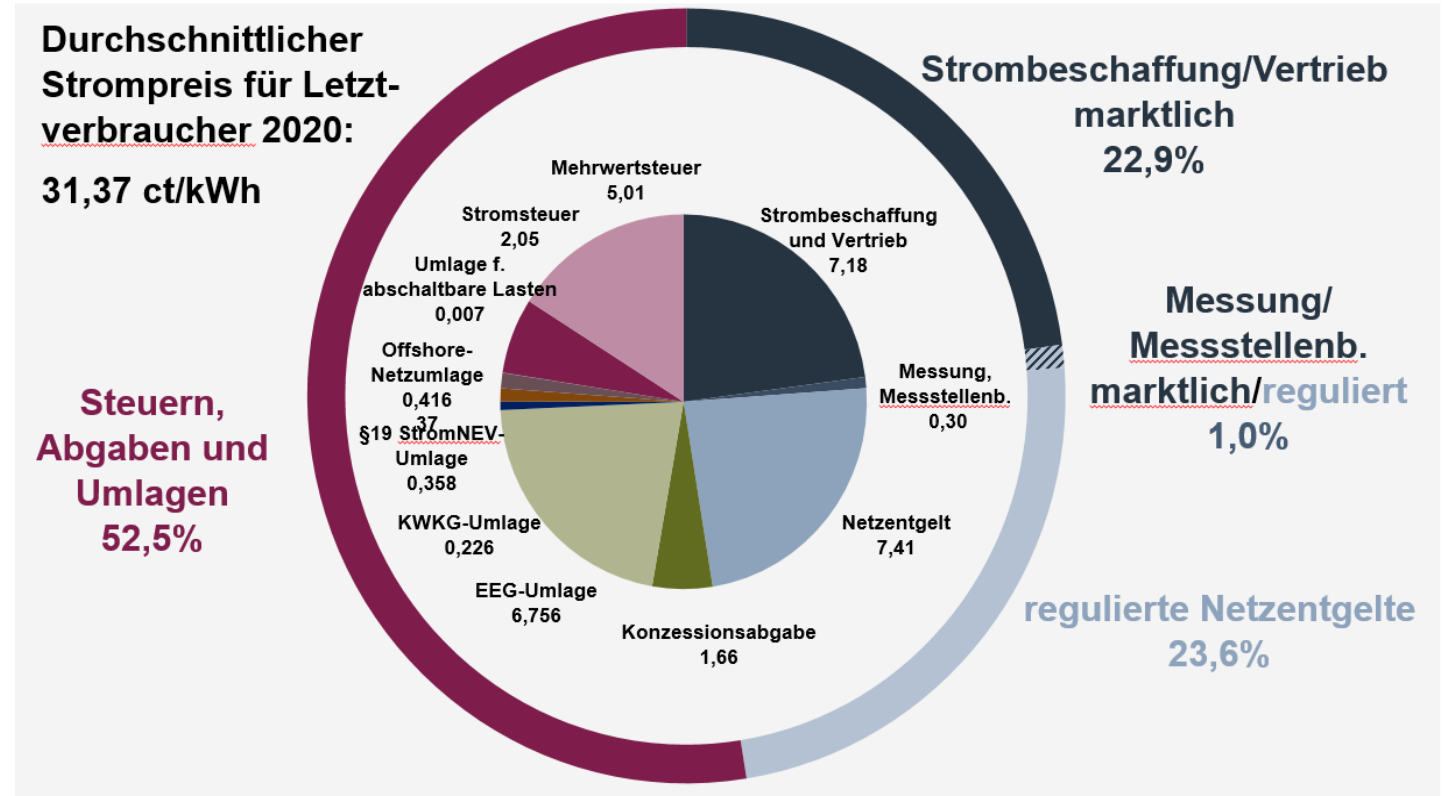


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Cost Analysis

- Purchase cost of heat pump is about **7 times** that of a heat boiler
- Cost-balancing electricity price (over 5 years) for heat pumps and electric boilers is **23 cents/kWh**
- The replacement of gas boiler with P2H technology is **NOT** cost-effective in the current situation



Quelle: BDEW, Angaben in ct/kW bei einem Verbrauch von 3.500kWh/a, Stand: 01/2020

Figure 10: Average Electricity Price for P2H Operators

Thanks for your attention!

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