

# Short-Term Price Elasticities of Heating Demand: A Statistical Analysis of Energy Billing Data in Germany

---

- Master Thesis -

In Partial Fulfillment of the Requirements for the Degree of Master of Science (M. Sc.)

Integrated Natural Resource Management (INRM)

Faculty of Life Sciences, Albrecht Daniel Thaer-Institute

Humboldt University of Berlin

---

*Submitted by:*

**Marc Blauert**

(Matriculation number: 605555)

Berlin, January 31, 2022

---

*Submitted to:*

1<sup>st</sup> Supervisor: Prof. Dr. Karsten Neuhoff (DIW Berlin, TU Berlin)

2<sup>nd</sup> Supervisor: Prof. Dr. Tobias Krüger (HU Berlin)

---

(Written in Cooperation with the Climate Policy Department at DIW Berlin)

# Acknowledgements

People I need to thank: Franziska, Till, Abteilung beim DIW -> Bereitstellung Daten, Betreuer.

# Table of Contents

<b>Chapter 1: Introduction</b>	<b>1</b>
1.1 Background	1
1.2 Relevance of topic	3
1.3 Research questions and objective	6
<b>Chapter 2: Theory and Literature</b>	<b>8</b>
2.1 Price elasticity of demand	8
2.2 Literature review	12
2.3 Conceptual model development	17
<b>Chapter 3: Data and Methods</b>	<b>22</b>
3.1 Data	22
3.1.1 Additional sources of data	23
3.2 Estimation methods and empirical model design	24
3.3 Workflow	26
<b>Chapter 4: Results</b>	<b>28</b>
4.1 Descriptive statistics	28
4.2 Full sample analysis	28
4.3 Subsample analysis	28
<b>Chapter 5: Discussion</b>	<b>31</b>
<b>Chapter 6: Conclusion</b>	<b>32</b>
<b>Appendix A: First Appendix</b>	<b>33</b>
<b>Declaration of Authorship (Eigenständigkeitserklärung)</b>	<b>35</b>
<b>References</b>	<b>36</b>

# List of Tables

2.1	Selected studies on the price elasticity of heating demand . . . . .	14
3.1	Variables and data sources . . . . .	24
4.2	Number of flight connections per destination . . . . .	28
4.1	Summary statistics . . . . .	30

# List of Figures

1.1	BEHG carbon price effects on gas and oil until 2025 . . . . .	4
2.1	Demand curves with different price elasticities of demand . . . . .	10
2.2	Conceptual model as directed acyclic graph . . . . .	18
A.1	Spatial and temporal distribution of buildings observed . . . . .	34

# Abstract

The preface pretty much says it all.

Second paragraph of abstract starts here.

# Chapter 1

## Introduction

### 1.1 Background

Residential energy demand of private households accounted for 20.2% of the overall primary energy consumption in Germany in 2020 (AGEB, 2021). In turn, 68.1% of this 20.2% was used for the purpose of space heating (RWI, 2021). Thus, space heating accounts for the largest share of primary energy consumption by private households in their homes. Far exceeding other categories of energy consumption such as hot water generation (16.1%), process heat (6.0%), room cooling, information technology, or lighting (all below 5%) (RWI, 2021). Furthermore, the two purposes of space heating and hot water supply continue to predominantly rely on energy from fossil sources. In 2020, 80.1% of the energy used for space heating came from either natural gas (45.4%), fuel oil (24.5%), or district heating (10.1%)<sup>1</sup> (RWI, 2021). These figures convey two messages. First, that the level of energy used for space heating strongly determines the total energy demand in the residential buildings sector. And secondly, that space heating is also the largest source of greenhouse gas (GHG) emissions in the residential buildings sector due to its generation from fossil fuels.

Since 1995, GHG emissions in the building sector in Germany have fallen by 34 % (ERK, 2020). This effect is mainly due to lower primary energy consumption through better insulation of the building stock and more efficient heating technologies. In addition, there is an overall lower emission intensity as a result of fuel switching (ERK, 2020). At the population level, however, these efficiency gains at the household

---

<sup>1</sup>Which is at present also to almost 90% generated from fossil fuels (natural gas or coal) or waste incineration (DESTATIS, 2018).

level are partly offset by countervailing effects such as the increase in the number of apartments (smaller average household size) as well as the increase in average living space (BMW<sub>i</sub>, 2021). Furthermore, much of the efficiency gains on the unit-level were already realized in the 1990s and 2000s. For the last decade, on the other hand, several independent data sources indicate that primary energy consumption per square meter in buildings in Germany has largely stagnated when the effective consumption values are adjusted for spatial and temporal differences in climatic conditions (BMW<sub>i</sub>, 2021; Stede, Schütze, & Wietschel, 2020; Techem, 2019).

The recent rates of energy consumption and associated GHG emissions in the residential buildings sector therefore not yet reflect the trajectory needed to achieve ambitious climate targets at national and European level, which – in line with the Paris Agreement – aim to limit global warming to 1.5 or well below 2 degrees Celsius by 2100. For Germany, the latest amendment of the national Federal Climate Change Act 2021 (Bundes-Klimaschutzgesetz, KSG)<sup>2</sup> stipulates that GHG emissions must reach net-zero in 2045. In addition, the interim target for 2030 was set at a 65% reduction in GHG emissions compared to the base year 1990 (previously 55%). Congruently, at the level of the European Union (EU), the Green Deal defines 2050 as the target year to reach net-zero in the whole union (European Commission, 2019). For the residential buildings sector, achieving these targets requires a comprehensive transformation of energy use, with measures to improve the efficiency of the building stock, demand-side electrification of the remaining energy use, as well as supply-side decarbonisation of energy generation (Hennes et al., 2021; Levesque, Pietzcker, Baumstark, & Luderer, 2021).

Basic economic theory relies on the rationale that energy prices represent one of the main drivers for the level of demand for space heat. Against this backdrop, this thesis aims to provide a statistical analysis of price elasticities of heating demand in Germany. Specifically, it is aimed at estimating short-term own-price elasticities of heating demand on the basis of an empirical sample of energy bills in multi-apartment residential buildings in Germany in the period from 2007 to 2019.

---

<sup>2</sup>Federal Climate Change Act of 12 December 2019 (Federal Law Gazette I, p. 2513), as last amended by Article 1 of the Act of 18 August 2021 (Federal Law Gazette I, p. 3905).



## 1.2 Relevance of topic

### Evaluation of price-related energy policies

Estimating price elasticities for demand response in the residential buildings sector is relevant for a number of reasons. First, empirical evidence on price elasticities is generally important for evaluating policies that affect the price of energy. Based on elasticity estimates, the economic, distributional and environmental impacts of policy measures can be better understood. In our time, this is especially applicable with regard to the attainment of ambitious climate targets.

One of the most important measures to induce GHG emission reductions in the building sector in Germany is the introduction of a carbon price for heating fuels, which was rolled out in 2021 on the basis of the Fuel Emissions Trading Act (BEHG).<sup>3</sup> In general, the introduction of a carbon pricing aims to correct the market failure of the previously unrecognized external effects of greenhouse gas emissions (Aldy, Krupnick, Newell, Parry, & Pizer, 2010; Stiglitz, 2019). It is expected that the consideration of external effects will have a steering effect on demand through an increased price level. The recently introduced Fuel Emissions Trading Act specifies a fixed price path for the period until 2025. The path rises from an initial price of 25 euros per ton of carbon dioxide (CO<sub>2</sub>) in 2021 to 55 euros per ton of CO<sub>2</sub> in 2025.

Figure 1.1 exemplifies the effect of carbon prices for the buildings sector through the BEHG. For gas (Panel A) and oil (Panel B) used for residential space heating, the respective left bar (green) represents the average end consumer prices for the five-year period 2016-2020 from national statistics (BMWi, 2021).<sup>4</sup> The two middle bars (yellow) show the absolute price effects that occur based on the carbon intensity of the energy carrier in 2021 (25 euro per ton CO<sub>2</sub>) and in 2025 (55 euro per ton CO<sub>2</sub>).

---

<sup>3</sup>Fuel Emissions Allowance Trading Act (BEHG) of 12 December 2019 (Federal Law Gazette I p. 2728) as amended by Article 1 of the Act of 3 November 2020 (Federal Law Gazette I p. 2291). Emissions from the building sector have not been subject to carbon pricing in Germany in the past, as they do not fall within the scope of the EU Emissions Trading Scheme (ETS). Besides emissions in the building sector, the BEHG also covers emissions from the transport sector. From 2026 onward the fixed pricing will be superseded by an emission trading system (ETS) with a price corridor.

<sup>4</sup>The 5-year time period (2016-2020) was chosen due to the relatively strong volatility of oil prices in the past years. An exclusive consideration of the price for 2020 would therefore have led to a distortion of the relative price effects. In addition, district heating is not considered here, as most district heating power plants were already subject to carbon pricing under the EU ETS. Only plants with a total combustion capacity of less than 20 megawatts were not previously subject to the EU ETS and will now also be subject to BEHG pricing.

The yellow bars are larger for oil due to its higher emission intensity.<sup>5</sup> Assuming otherwise constant prices, the initial CO<sub>2</sub> price in 2021 would thus lead to a price increase of +0.50 cents (+7.4%) per kWh gas and +0.66 cents (+13.4%) per kWh oil. In 2025, the price effects rise to +1.10 cents (+16.3%) per kWh gas and +1.46 cents (+29.6%) per kWh oil.

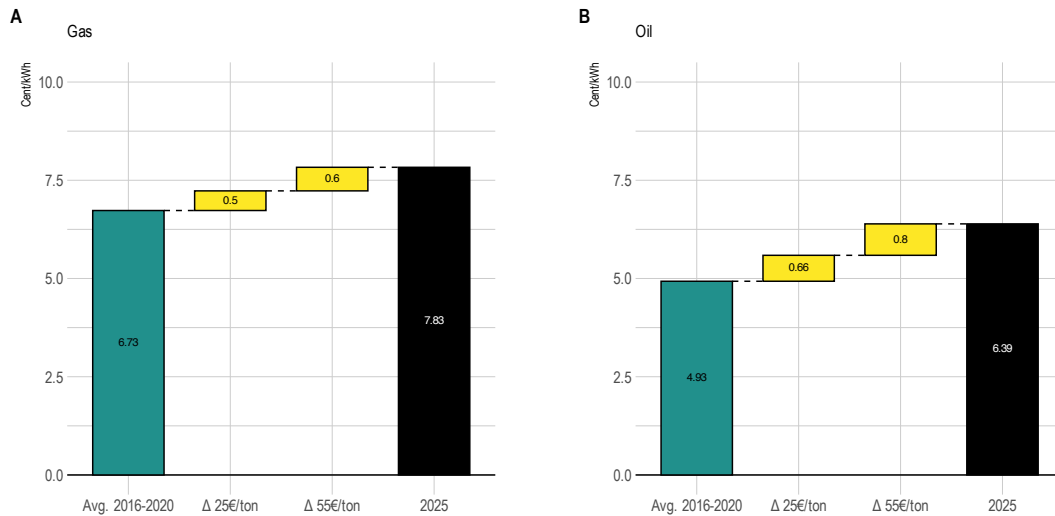


Figure 1.1: BEHG carbon price effects on gas and oil until 2025

These price figures clearly indicate that relevant steering effects induced by the BEHG are likely to occur already in the next years. Higher price levels driven by the structural price component of carbon pricing can be expected to have two impact channels. The first impact channel are short-term price reactions. Consumers adjust their consumption levels for heating fuels due to higher or lower prices (Alberini, Gans, & Velez-Lopez, 2011). Short-term demand responses are therefore of direct relevance to forecasting building sector energy demand and emission levels in the coming years. The expected effect size of these short-term reactions are reflected by short-term price elasticities of heating demand which will be estimated based on empirical data in this thesis.

The second impact channel, on the other hand, is that higher prices for fossil fuels can be expected to induce long-term steering effects for investments in low-carbon heating technologies and energy efficiency. In the literature, long-term investment

<sup>5</sup>Emission factors for gas (0.201 kg CO<sub>2</sub>e per kWh) and oil (0.266 kg CO<sub>2</sub>e per kWh) were taken from (UBA, 2016).

decisions driven by mitigation costs potentials are modeled using integrated assessment models (IAMs) for the buildings sector (e.g., Bürger, Hesse, Köhler, Palzer, & Engelmann, 2019; Levesque et al., 2021). Depending on the model structure, these IAM sector modules are often informed by assumptions on the price elasticities of heating demand as one of their input factor besides other parameters such as the cost of capital or substitution elasticities.<sup>6</sup> Therefore, empirical estimates of price elasticities are also important for further analyses in the literature on the second long-term impact channel.

Additionally, the empirical assessment of the demand effect may be particularly important in the current phase of strong energy price increases following from the economic imbalances after the external shock of the COVID-19 pandemic and Russia’s war of aggression on Ukraine. While the current price spikes can lead to stronger short-term demand responses, only a fraction of the response is due to the structural component of CO<sub>2</sub> pricing and – if the effect does not last long enough to trigger significant long-term investments – would disappear when oil and gas prices might return to lower levels. In order not to overestimate the effect of the structural pricing component of the CO<sub>2</sub> price, empirical estimates from past price reactions could be used to separate the effect into a CO<sub>2</sub> and a commodity price-related element.

### **Different types of data to estimate elasticities**

A second line of argument for the relevance of this study is that the existing empirical evidence on the price elasticity of heating demand is based on different types of data sources, but few studies have been conducted on actual billing data.

Studies on the price elasticities of heating demand rely on different types of data sources. For the case of Germany, ex-post statistical analyses for the residential buildings sector use data from social panel surveys (e.g., Rehdanz, 2007; Schmitz & Madlener, 2020; Schulte & Heindl, 2017). In these studies, household expenditure on heating reported in the survey is usually examined as the relevant variable, but the actual level of energy consumption and energy prices are not observed. This means that, for instance, different energy contract conditions between households may obscure real consumption levels and thus affect the validity of the estimated

---

<sup>6</sup>For a more detailed example of how own price elasticities of demand are taken into account in IAMs, see the model structure description of the ETSAP TIAM model in Loulou & Labriet (2008). Since IAMs vary in their specific structure and relations other models may consider price elasticities differently.

results. The data available for this thesis is different since the energy bills do provide actual consumption and price data. Other studies using data from energy bills have previously been conducted in the United States (US) (e.g., Auffhammer & Rubin, 2018).

In this context it should be noted that the data sample used in this thesis also has its weaknesses. The data does not observe demand responses at the level of individual households but at the level of multi-household buildings. Yet despite this aggregation of multiple households on the building level, I believe the informative value of estimates to be complementary to estimates from other types of data sources such as social surveys.

### 1.3 Research questions and objective

Against the background of the tightened climate targets, the recently introduced carbon pricing, and the sharp price increases for heating fuels, the question of the price elasticity of heating demand is once again moving to the center of public attention. The aim of this work is to investigate the price elasticity of heating demand in multi-family houses in Germany based on a large-scale sample of energy bills. The guiding research questions for the analysis are:

- How does a change in energy price affect the level of energy consumption?
- What other variables might affect energy consumption and need to be taken into account so that their effects are not falsely attributed to energy prices?
- And, are there potential factors for heterogeneity within the sample that would be obscured when only estimating an overall elasticity?

By using energy bills as a data source, the study aims to complement existing studies based on different sources. Furthermore, the period studied is more recent and can therefore complement older studies. Also, the analysis aims not only to estimate an overall elasticity, but also to investigate potential factors for heterogeneity. The statistical analysis aims to use several different modelling approaches, both Frequentist and Bayesian. As far as the author is aware, the Bayesian approach has not been used in previous research in this field. Furthermore, the elasticity estimates from this work may also be relevant for other countries and regions that have comparable conditions to Germany in terms of building stock and energy demand and also face

the challenge of decarbonising the building sector in the coming decades. Here, my results would serve as an empirical guide to the short-term demand effects for any kind of price-related policy instrument.

The thesis is structured in six Chapters. While the first Chapter was used to provide an introduction, Chapter 2 gives an overview on the existing literature on price elasticities of space heating demand. In the following Chapter 3 the data and methods used for the analysis are laid out. Chapter 4 provides the results of the analysis, which are then discussed in Chapter 5. Chapter 6 provides concluding remarks.

# Chapter 2

## Theory and Literature

The objective of this chapter is to provide a short theoretical background on price elasticities as well as an overview of the previous research on price elasticities of heating demand. The text is organized as follows. The first part of the chapter focuses on the theoretical background and also provides a placement of the effects of varying elasticities in connection to heating demand. The second part of the chapter then moves to summarizing the empirical evidence from the literature.

### 2.1 Price elasticity of demand

Elasticities are one of the key concepts in micro-economic theory. They are used to express the sensibility of one variable to a change in another. The own price elasticity of demand is one type of elasticity. It represents the magnitude of a change in consumption of a good following from a change in its own price (Pindyck & Rubinfeld, 2018). Formally, the own price elasticity of demand  $\epsilon$  can be expressed as

$$\epsilon = \frac{\frac{\Delta Q}{Q}}{\frac{\Delta P}{P}} = \frac{P \Delta Q}{Q \Delta P} \quad (2.1)$$

where  $\Delta P/P$  represents the percentage change in the price  $P$  of a good and  $\Delta Q/Q$  the corresponding percentage change in the quantity  $Q$  demanded of the same good. Due to budget constraints, consumers tend to consume less of a good when its price increases, which implies that under normal conditions the price elasticity of demand is negative. But the responsiveness of the demand reaction may vary across different goods. A common assumption in the econometric literature is that elasticities are

constant (Varian, 2010). Leading to a demand function that can be expressed by

$$Q = AP^\epsilon \quad (2.2)$$

where  $A$  represents an arbitrary constant and  $\epsilon$  again is the price elasticity of demand. Taking the logarithms of the demand function removes  $\epsilon$  from the exponent and yields

$$\ln(Q) = \ln(A) + \epsilon \ln(P) \quad (2.3)$$

which can be referred to as the elasticity case and will reappear at a later point, when constructing the regression models to estimate price elasticities from the sample data.

To make the assumption of constant elasticity more intuitive, it seems useful to return to the subject of interest and show the behavior of demand curves for space heating under varying price elasticities of demand. Generally, the literature distinguishes between three different types of elasticities. If the demand response for a good is greater than the change in its price ( $\epsilon < -1$ ), it is called price elastic. If the demand response is exactly equal to the change in price ( $\epsilon = -1$ ), it is said to be unit-elastic. Finally, if the demand response is smaller than the price change ( $0 > \epsilon > -1$ ) – and therefore indicating a lower responsiveness – it is called price inelastic (Pindyck & Rubinfeld, 2018).<sup>1</sup>

Figure 2.1 visually represents these different price elasticities for the demand of space heating.<sup>2</sup> To construct the demand curves, a common arbitrary intersection point is chosen for the curves. Specifically, this intersection has a consumption value of 130 kilowatt hours (kWh) per square metre (sqm) and a price of 6 cents per kWh. These values can be considered realistic for an average building in Germany, as we will see later in this thesis when working with the sample data.

---

<sup>1</sup>It should be noted that in the literature the minus of the demand elasticity  $\epsilon$  is sometimes omitted, as it is assumed that it is generally negative. However, I personally find it more intuitive not to do so – especially when working empirically where positive estimates are a possibility – and will therefore continue to use the negative values in this thesis.

<sup>2</sup>Please note that in the graph, energy price as the independent variable is plotted on the horizontal axis and energy demand (the level of consumption) as the response variable on the vertical axis. I consider this convention – which is common in most of science – to be more intuitive than the traditional Marshallian cross diagram in Economics, where the price is plotted on the vertical axis.

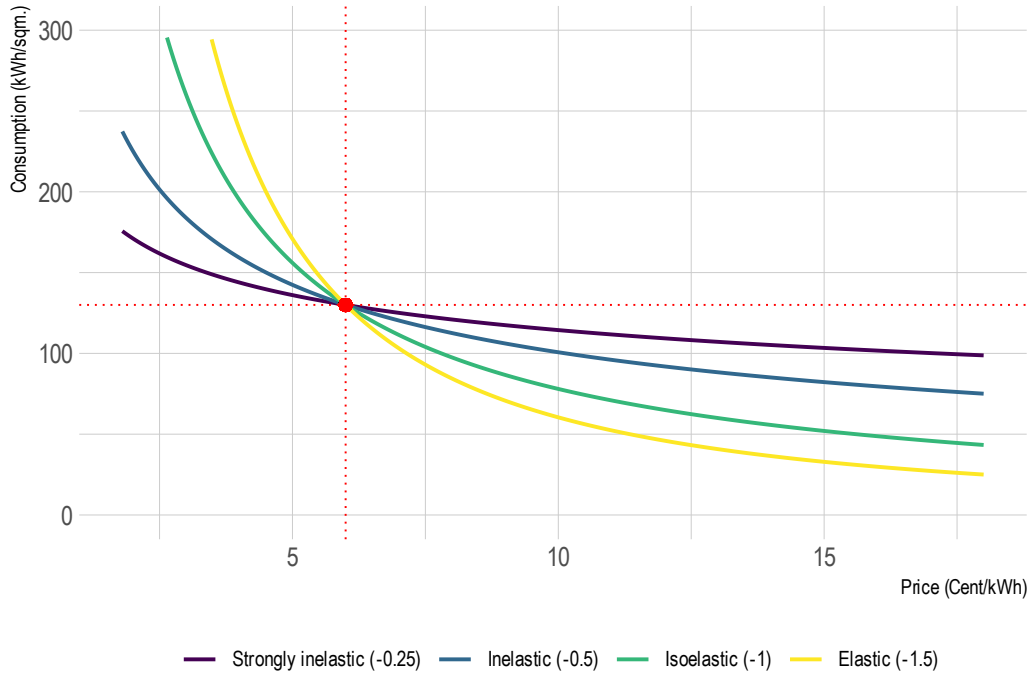


Figure 2.1: Demand curves with different price elasticities of demand

The graph shows a total of four demand curves. In addition to the unit-elastic curve ( $\epsilon = -1$ ), there is also an elastic ( $\epsilon = -1.5$ ) and two inelastic ( $\epsilon = -0.25$ ,  $\epsilon = -0.5$ ) demand curves. The special feature of the unit elastic demand curve is that it gives the same result for total expenditure on heating costs for all combinations of price times demand. This is not the case with the other elasticities. First, a look to the right of the intersection (higher prices). In the elastic case (yellow curve), total expenditures fall when the price is higher, which means that households can adjust their demand downwards more. In the opposite, inelastic cases (purple and blue curves), on the other hand, households would reduce their demand less and therefore also have higher total expenditure. Conversely, to the left of the intersection point (lower prices), more elastic demand (yellow curve) leads to a higher level of consumption and thus also to higher total expenditure than in the inelastic cases (purple and blue curves), where consumption changes only gradually.

If we now take a step back and consider what kind of good the demand for space heating represents and what price elasticity of demand one might expect, it seems reasonable to assume that the elasticity is rather inelastic. There are two indications



leading to this assumption. First, heating energy is to be regarded as an essential good that is associated with a minimal level of demand – especially in the winter months. Essential goods tend to follow a more inelastic demand pattern than other types of goods (Gwartney, 1976). Households can only reduce their demand to a certain extent (right side of the intersection), but on the other hand, they do not have a strong rationality for a large increase in heating demand beyond a comfortable level of warmth (left side of the intersection). The second indication for a rather inelastic demand is that due to the dependency on the existing heating system – especially for the tenant households investigated in this thesis – there is not really a viable option to substitute space heating from one source with space heating from another.

Another important dimension for distinguishing different price elasticities of demand is time. More precisely, the time that elapses between a price signal and the measurement of the demand response. Here, the literature distinguishes between short-term and long-term price elasticities (also called short- and long-run elasticities) (Pindyck & Rubinfeld, 2018). Short-term elasticities are those that are estimated in temporal proximity to the price change – in this case, that would be the energy price change within a billing year and the associated change in consumption in the same period (or in the year after the price change). Long-term elasticities, on the other hand, imply that more time has elapsed for consumers to fully adjust to a price change and also to make long-term investment decisions that would change the overall level of consumption (Pindyck & Rubinfeld, 2018). For the subject at hand, such decisions may include changes in the isolation of a building or an exchange of the heating system. Thus, long-term demand elasticities tend to be larger than their short-term counterparts (Schmitz & Madlener, 2020). Therefore, the two types can be seen as mirroring the two different channels through which a pricing policy can influence consumer behavior (see Section 1.2).

In the literature, long-term price elasticities are usually estimated using dynamic models, where the elasticity estimated between two periods is only a fraction of the long-term elasticity (e.g., Alberini et al., 2011). However, I have chosen follow the example of Schmitz & Madlener (2020) and only estimate short-term elasticities. This is because I believe that the long-term approaches do not apply well to the given context in which I am analyzing a subset of the total residential building segment consisting of rental buildings where households have relatively few opportunities to make long-term adjustments that would significantly affect the energy efficiency or carbon intensity of the energy of their apartments.

## 2.2 Literature review

The study of the price elasticity of energy demand is a broad field of academic research which also has a long history dating back to the mid-twentieth century (e.g. Cutler, 1941; Houthakker, 1951). In the more recent past, as well numerous studies have been published. These studies differ considerably in terms of their focus and design. Several conceptual factors can systematically influence the elasticity estimates. The most important of these factors is the type of energy application analysed: The price responsiveness of household space heating demand has only limited informative value for the price responsiveness of electricity consumption or transport fuel demand. The same is true for varying consumer groups: Households are likely to react differently to energy price changes than industrial consumers. Beyond, other conceptual factors which may as well affect estimates are the statistical methods used (time-series analysis, panel analysis, cross-sectional analysis), varying sources of data (national accounts, aggregate sector-level data, micro-data), varying spatial and temporal focus, measurement heterogeneity, and price trends (Csereklyei, 2020; Miller & Alberini, 2016). While many of the earlier studies either had only a national granularity or, when using micro data, focused predominantly on the US or European countries, a considerable number of studies have also been published recently that cover emerging countries (especially China) as well as other countries.

### **Meta-estimates as first indication of elasticity magnitudes**

To organize this comprehensive body of literature with a wide range of energy applications, meta-studies do serve as a good starting point. The recently published meta-analysis by Labandeira, Labeaga, & López-Otero (2017) is particularly suitable for this purpose, as it covers a broad spectrum of energy uses and thus goes beyond previous meta-analyses that primarily focus on the elasticity of petrol demand. I will therefore first delve into the aforementioned meta-analysis before turning to a subset of selected studies that take a closer look at the specific issue of price elasticities for space heating demand in the residential buildings sector.

In their analysis, Labandeira et al. (2017) include a total of 428 papers, all published between 1990 and 2016. For the demand of natural gas, they find an average short-term (long-term) price elasticity of -0.18 (-0.57) based on 230 individual estimates. And for heating oil they find a comparable average short-term (long-term) elasticity of -0.19 (-0.54), which, however, is based on only 44 individual estimates (Labandeira et al., 2017). An average elasticity for space heating – which is generally

less focused on in the literature due to its relatively low prevalence – is not given.

The above average estimates of the price elasticity of demand for gas and oil, however, come with an important caveat. While they are broken down by energy carrier, they are drawn from a wide range of consumer groups. This means that the underlying studies include not only those that examine demand response patterns for residential consumption, but also others that examine these patterns for the commercial building stock and/or the industrial sector. Labandeira et al. (2017) also provide average price elasticities stratified by consumer groups, but these are on the other hand not clustered by energy source. When differentiated along the dimension of consumer groups, they report that the average estimates are slightly higher in the residential household sector (short-term: -0.22; long-term: -0.62) than in the industrial sector (short-term: -0.17; long-term: -0.51), which seems reasonable given that households likely have more ability to change their consumption levels than the industrial sector with more fixed production patterns and the ability to forward costs. Overall, however, the average elasticity estimates are all in a similar size range and together convey the message that demand is strongly inelastic in the short-term, confirming the theoretical arguments that energy is an essential good in many applications and also has limited substitutability (see section 2.1).

The meta-estimates are useful in providing a high-level overview. However, as already mentioned, they include estimates from a wide range of energy applications that not necessarily reflect the demand response for the specific application of space heating in residential buildings. Furthermore, they are average estimates which may have the effect of masking discrepancies and variations between studies. Therefore, the following section presents individual studies that explicitly focus on the space heating demand of the residential sector.

### **Selected studies on space heating in the residential buildings sector**

The space heating demand of the residential sector represents a subset of the full literature stream. The selected studies were chosen based on the criteria that they correspond to the focus of this work and are of good quality (peer-reviewed journal articles). A summary of the studies is provided in Table 2.1.

There are three earlier studies that have a spatial focus on Germany. The first study was conducted by Rehdanz (2007), who estimates price elasticities for different energy carriers for residential heating demand. She uses social survey data on household-level heating expenditures from the Socio-Economic Panel (SOEP) in the years 1998 and

Table 2.1: Selected studies on the price elasticity of heating demand

Study	Type of data used	Energy carrier	Short-term price elasticities
<i>Studies within Germany</i>			
Rehdanz (2007)	Social survey, panel data (SOEP), all types of buildings, 1998 and 2003	Gas Oil	-0.44 to -0.63 -1.68 to -2.03
Schmitz and Madlener (2020)	Social survey, panel data (SOEP), all types of buildings, 1996-2014	(All)	-0.31 to -0.43
Schulte and Heindl (2017)	Social survey, panel data (EVS), all types of buildings, 1993-2008	(All)	-0.50
<i>Studies outside of Germany</i>			
Alberini et al. (2011)	US, household-level social survey, 50 metropolitan areas, panel data, 1995-2007, only single-family homes and duplexes	Gas	-0.57 to -0.69
Auffhammer and Rubin (2018)	US, energy billing data, household-level, only California, panel data, 2003-2014	Gas	-0.17 to -0.23
Leth-Petersen and Togeby (2001)	Denmark, apartment-block level (>1,500 sqm), panel data, 1984-1995	Oil	-0.08
		District heating	-0.02
Meier and Rehdanz (2010)	UK, household-level social survey, panel data 1991-2005	Gas Oil	-0.34 to -0.56 -0.40 to -0.49
Metcalf and Hassett (1999)	US, household-level, panel data, 1984, 1987 and 1990	Gas	-0.48 to -0.71

2003. Methodologically, a cross-sectional statistical analysis with dummy variables is conducted for the two years – in other words, no panel structure is used. In contrast to the average meta-estimates presented in the previous section, the study finds that demand for heating oil has a high elasticity of -1.68 to -2.03, depending on the model specification. For gas, the results indicate a less elastic demand between -0.44 and -0.63, also depending on the model used. Therefore the study concludes that the type of fuels can have a strong influence on the price sensitivity of households (Rehdanz, 2007). In a more recent study, also based on the SOEP social survey data but covering a more comprehensive time period between 1996 and 2014, Schmitz & Madlener (2020) find a price elasticity of heating and hot water demand of -0.31 to -0.43, depending on model specifications. They do not differentiate the elasticities by energy carrier. The price elasticities are derived from household expenditure elasticities, as demand is not directly observed. In contrast to Rehdanz (2007), the study uses a panel structure where observations are clustered by household and year using fixed effects. This produces much lower price elasticity estimates, which are more consistent with the meta-estimates presented in the previous section. Moreover, the study discovers that price elasticity is heterogeneous across different socio-economic groups. Given the household-level information at their disposal, they find that higher-income households are less sensitive to energy price changes than lower-income households. Likewise,

homeowners show less sensitivity than renters (Schmitz & Madlener, 2020).

Also for Germany, Schulte & Heindl (2017) investigate the price and expenditure elasticities of private energy demand in Germany between 1993 and 2008. For their analysis, they use survey data from the *Einkommens- und Verbrauchsstichprobe (EVS)* and analyse them with a demand system approach (in particular, they estimate a quadratic expenditure system). For space heating, they find an own-price elasticity of -0.50 across all households. They further observe that the price elasticity changes with the level of total household expenditure, with households in higher expenditure strata reacting more strongly to price changes (Schulte & Heindl, 2017). Put differently, this means that following an increase in price level low income households tend to decrease energy consumption to a lesser extent as compared to households with a higher income. Thus, in relation to the income dimension the results must be understood as contradictory to those of Schmitz & Madlener (2020).

All three studies on Germany rely on data from household-level social surveys. Therefore, the studies take a strong perspective on the impact of household income on energy price sensitivity. As the data for this study is aggregated at the building-level and not at the household-level, however, the focus of this thesis is to some extent different. The focus is rather on the overall price response based on actual price and consumption values, which are not available in the social surveys, and on the response in interaction with other characteristics and changes at the building level.

Besides the studies focusing on Germany, Table 2.1 also reports the evidence from five selected international studies focusing on the price elasticities of space heating demand. Alberini et al. (2011) examine household demand for gas (and electricity) in single-family homes and duplexes using household-level social survey data in 50 metropolitan areas in the United States (US) over the period 1995-2007. As a modelling approach, they use static and dynamic panel models. For the static models reflecting the approach taken in this thesis, they find an own price elasticity of private gas demand between -0.57 and -0.69, depending on the model specification. These estimates are in their magnitude comparable to the results of an earlier study by Metcalf & Hassett (1999) who use the 1984, 1987 and 1990 waves of the US Residential Energy Consumption Survey (RECS) to examine homeowners' investments in energy efficiency measures and as part of their study find price elasticities of residential gas demand to range between -0.48 and -0.71. In contrast to these estimates, a recent study by Auffhammer & Rubin (2018) finds that the short-term price elasticity of residential gas demand is even more inelastic, ranging from -0.17 to -0.23, not for the

whole US but for California in particular. The study by Auffhammer & Rubin (2018) is especially interesting for this thesis because it is the only one that also uses energy billing data. The data analysed covers the period 2003-2014 and consists of monthly energy bills at the household-level. So although the type of data source used to derive the estimates is the most similar to the approach taken in this thesis, there are still two relevant differences: firstly, gas tariffs in the US may change monthly rather than annually, as is the case in the German dataset used for this analysis, and secondly, the data in Auffhammer & Rubin (2018) relates to individual households rather than the aggregate building-level.

In addition to these three studies focusing on the US, table 2.1 also contains the results of a Danish study by Leth-Petersen & Togeby (2001). They use a panel data approach and analyse data from the period 1984-1995. Interestingly, they also use the building-level as the level of analysis. More specifically, they study apartment buildings with a base area of more than 1,500 square meters (sqm) in Denmark. As an empirical strategy, they rely on fixed effects models. As a result, they find very inelastic demand responses for space heating. For oil, the short-term price elasticity is found to be -0.08 and for district heating as low as -0.02, meaning almost no demand response to a changing price at all. Another study by Meier & Rehdanz (2010) looks at the United Kingdom (UK) and analyses a household-level social survey panel for the 15-year period 1991-2005. For their model they use a log-linear approach. They find that the short-term price elasticity of demand for gas ranges from -0.34 to -0.56, depending on the specification. For residential oil consumption, the results are in the same order of magnitude, but narrower, between -0.40 and -0.49. In addition, they find that the type of occupant is a relevant dimension for a different price responsiveness. While renters adjusted their consumption more, homeowners showed a more inelastic response.

### **Synthesis of estimates from the literature**

If one aims to summarize the literature on the price elasticity of residential space heating demand presented previously, commonalities and differences emerge. The most important commonality is that the price elasticity of energy demand in general and household space heating demand in particular is inelastic in almost all empirical estimations. However, when looking at the individual studies, differences in the magnitude of the estimates appear. While the meta-results by Labandeira et al. (2017) as well as a number of individual studies indicate a strongly inelastic demand response (e.g. Auffhammer & Rubin, 2018; Leth-Petersen & Togeby, 2001), other individual

studies still indicate an inelastic but stronger demand response (e.g., Alberini et al., 2011; Meier & Rehdanz, 2010; Metcalf & Hassett, 1999; Rehdanz, 2007; Schmitz & Madlener, 2020; Schulte & Heindl, 2017). These differences in magnitude may be partly due to the different local and temporal conditions of the studies or due to differences in the data and methodology. However, other aspects such as climatic conditions, the building stock or behavioral aspects related to a specific population may also be important. Therefore, it should be noted that all three studies that focused on Germany found somewhat more sensitive price elasticities ranging from -0.3 to -0.6.<sup>3</sup>

Another common feature of all the individual studies presented is that they do not focus on one-off extreme price shocks as an identification strategy, but rather on gradual price developments. They are mostly built around a panel data set covering a longer period of time. The period studied in this paper is also a longer period with rather gradual price developments. This should therefore be seen as an important aspect that will facilitate the comparability of the results of this study with the literature estimates presented.

## 2.3 Conceptual model development

In the previous two subsections, the theory around price elasticities of demand (see section 2.1) and evidence from the prior literature (see section 2.2) in the context of space heating demand have been presented. To complete this foundation, this subsection uses the theory, but more importantly the approaches of prior studies, to identify relevant variables in the relationship between space heating demand and energy price. The aim is to create a conceptual model, which will then be translated into a statistical model and empirically examined in the further course of this thesis.

### Energy price affecting space heat demand

The starting point for the conceptual model is the relationship between space heating demand and energy price. Here, the basic assumption from the theory of price and demand is that the demand for space heating is influenced by the energy price. When the price goes up, demand moves down along the demand curve and vice

---

<sup>3</sup>Note that the extreme results of Rehdanz (2007) for oil as an energy source, which come only from a cross-sectional regression analysis, are omitted here. Since only two isolated periods are considered, these results must be considered more prone to extreme results than is the case with a longer panel.

versa. Transferred to a statistical terminology this implies that space heat demand is the response variable of a model and energy price the main explanatory variable. To support the development of the conceptual model visually, Figure 2.2 depicts it as a directed acyclic graph (DAG).<sup>4</sup> The heating demand, whose variance we want to explain, is shown in the center (yellow bubble). Since the effect of energy price (purple bubble) on space heating demand is to be regarded as the price elasticity, it is marked with a red arrow.

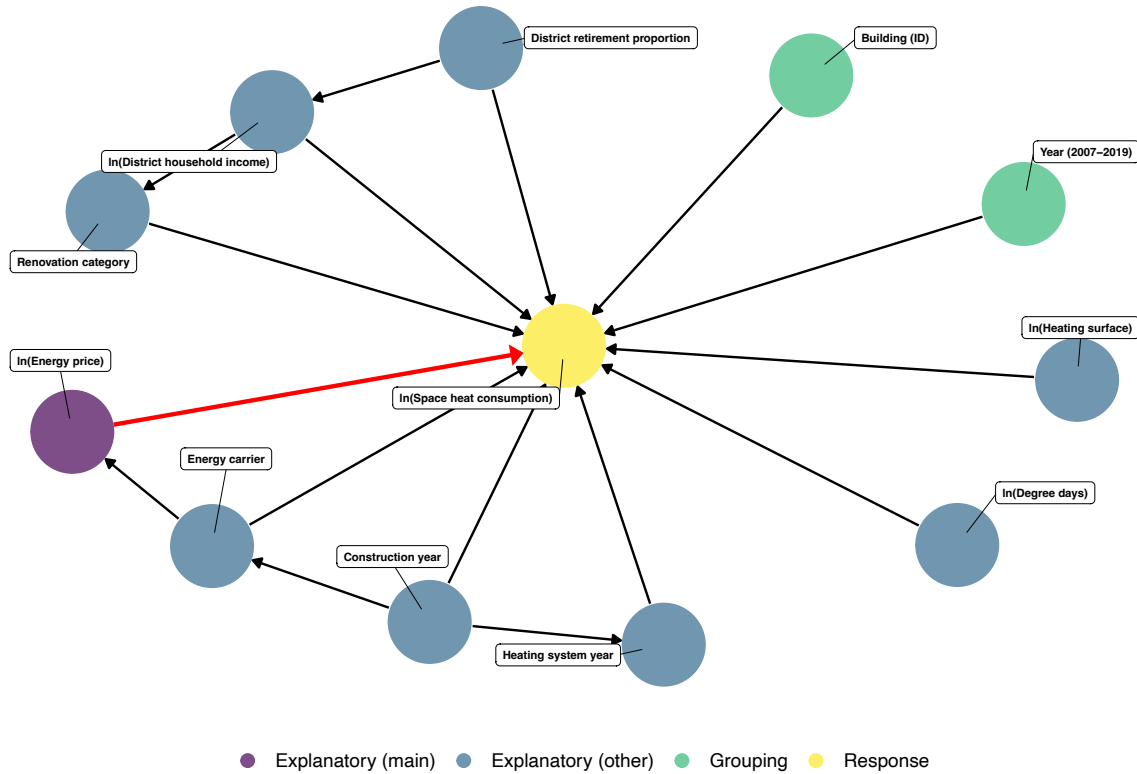


Figure 2.2: Conceptual model as directed acyclic graph

In addition to the energy price, there are a whole range of other variables that may influence the level of space heating demand. These are shown in the DAG as additional explanatory variables (blue bubbles) and as grouping variables (green bubbles). In order to establish the additional variables relevant for the model, I used a structured procedure. First, I used the recent work of Schmitz & Madlener (2020) as a basis to create a list of potential variables. Secondly, I used the additional studies presented in section 2.2 to cross-validate their relevance and to screen for further variables. This

<sup>4</sup>Please note that the arrows between heat demand and the variables used to explain demand are of relevance here. Other relationships between the explanatory variables are hinted at for the sake of completeness, but are not examined in the course of the thesis.



structured approach leads to the variables presented in the following.

### **Climatic condition**

Firstly, the consideration of the local climatic condition in a billing period is of importance, as it can have a strong influence on the amount of heating demand (hesse20?). In the previous literature using micro-level panel data, climatic conditions are always considered and are usually approximated by the outdoor temperature, measured in degree days. In line with (vdi13?), degree days in this study are defined as the temperature difference between a mean room temperature of 20°C and the daily mean of the outdoor temperature, provided it is below the heating limit of 15°C.<sup>5</sup> The intuition is that with lower mean outdoor temperatures, transmission and heat losses increase and thus the demand for space heating increases to compensate for the heat losses. Lower mean outdoor temperatures are associated with more degree days in a period. Thus, one would expect a higher space heating demand when the degree days in an annual heating cycle are higher and vice versa. Varying degree days may occur on a temporal (one year is warmer/cooler than another) but also on a spatial scale (places at higher altitudes or in the interior of the continent have structurally lower temperatures).

### **Building characteristics**

In addition, various building characteristics are included in the model as additional explanatory variables. The heating area of a building can influence the heat demand. As the ratio of the building's external surface area to the heating surface area decreases in larger buildings, heat loss decreases and accordingly a relatively lower space heating demand can be expected. The intuition is therefore that in buildings with a large heating surface, the space heating demand should be lower.<sup>6</sup> The type of energy carrier is another building characteristic, which was used in almost all previous studies (see Section 2.2). The intuition here is that different types of carriers (gas, oil, district heating) are associated with different heating technologies, which can lead to structural differences in energy use and efficiency. This effect could be amplified by the fact that certain heating technologies were introduced in certain periods in the past, which means that, for example, the average oil heating system may be older and therefore less energy efficient than the average gas heating system.

---

<sup>5</sup>While this paper uses heating area as a continuous variable, other studies with a less detailed database often use building size categories, which can be seen as a similar approach.

<sup>6</sup>While this paper uses heating area as a continuous variable, other studies with a less detailed database often use building size categories, which can be seen as a similar approach.

Additionally, also the three building-level characteristics construction year of the building, heating system installation year, and comprehensive renovation are included. Those variables are only available for a subset of the overall energy billing sample (see Section 3.1). Thus, they are considered but not in all models. The intuition for the building construction year is that older buildings were subject to either no or less stringent new construction energy efficiency standards which most importantly include the *Wärmeschutzverordnung (WSVO)* which was introduced in 1977 and later replaced by the waves of the *Energieeinsparverordnung (EnEV)*. Thus, the intuition is that older buildings have a higher space heat consumption. Similarly, older heating systems are also less energy efficient and therefore lead to a higher space heat consumption. Furthermore, the information is also important to observe if a heating system has been exchanged in the period observed. Lastly, the comprehensive renovation variable is used to observe if a major renovation on multiple components of a building have taken place during the observed period. The intuition here is that with a more comprehensive renovation the energy efficiency of a building is likely to improve and therefore the heat demand would decline.

### **District socio-economic variables**

Besides these building-related characteristics, additional socio-economic variables can also influence heating demand. In the literature that draws its data from social surveys, household-level socio-economic information such as income, number of inhabitants, age, education level or employment status is usually part of the core survey data. Since the present study relies on building-level billing data, this approach is not feasible because the data is not available and because the level of analysis is the building and not the household level. Nevertheless, socio-economic variables could also play a role at this aggregated level as well. Therefore, three socio-economic variables at the district-level are included in the model to capture potential overlaying effects. Those are, the average regional household income where in line with most of the literature the intuition would be that with a lower income the heat demand would be lower. However, it may also be that income on a regional level rather serves as an approximation of the overall affluence of the district and therefore effects such as less economic means available for investments into building efficiency dominate the effect, so that a lower regional income is rather associated with higher space heat demand. Additionally, the regional retirement share is considered as a variable. Here, the intuition is that with high retirement rates in a region more people are at home most of their time which in turn might lead to higher space heat demand. Lastly, the

---

population density of a district is considered to observe if there may be structural differences in heating demand between metropolitan areas on the one hand and rural areas on the other.

# Chapter 3

## Data and Methods

### 3.1 Data

#### Energy bills panel data

Several data sources were used for the analysis. The data on space heat consumption and energy prices – which is the backbone of the analysis – was obtained from a large-scale building-level panel of energy bills made available through the Climate Policy Department at DIW Berlin<sup>1</sup>. The sample contains information on an average of 260,000 multi-unit residential buildings in Germany from 2003 to 2019. This means that the smallest buildings observed have two living units; single-family houses with just one unit, on the other hand, are not observed.

The sample is structured around annual heating bills at the building-level. While the consumption data in the panel already start in 2003, the energy prices are only observable with a more extensive data delivery at the end of 2008. However, since a relevant number of observations billed in 2008 already have the majority of the billing period in 2007, the panel data used in the analysis ranges from 2007 to 2019. Spatially, the data have a good coverage and spreads over all German federal states.

In the sample, the consumption data is given in kilowatt hours (kWh) consumed per building and per annual billing. Importantly, the data set also includes information

---

<sup>1</sup>The data originates from the energy- and billing service provider ista Deutschland GmbH and was provided to DIW Berlin for scientific use. Due to the sensitive nature of the data, it is classified as confidential and was only accessed using the internal servers at DIW Berlin. Making the data available to an external party for own scientific purposes or reproduction of results is therefore not possible.

on the relative share of energy consumed for space heating and the relative share for hot water production. While one group of buildings also uses central heating for hot water production, other buildings do not produce hot water. In order not to distort the overall picture of energy consumed, only the share of space heating that is relevant for all buildings is considered. In order to obtain a comparable metric for the different sized buildings, the total consumption at building level is divided by the total heating area in square meters. This results in the space heating consumption in kWh per square meter as the consumption variable (kWh/sqm.).

The data structure for prices in the data set is similar to that for consumption data. Total annual costs in euros are given on the building-level. Based on the relative consumption shares for space heating and hot water generation, the share of space heating in the total costs can be isolated. In a second step, the proportional costs for space heating is converted into per-unit costs by dividing them by the respective amount of space heating consumed in kWh. This gives a per-unit cost in cents per kWh (cents/kWh). A breakdown according to fixed and variable cost components is not reported in the data. Prices are provided as nominal prices.

Besides consumption and price data, the panel also features additional building characteristics that must be considered relevant for the analysis. This includes the exact dates of the billing period which allows for assigning a bill to a year. Additionally, the data also features an unique identifier per building. This is important as it allows to identify changes on the building-level over time and to use buildings as a unit for clustering (panel structure). Furthermore, the data set also provides important information on the individual properties of the buildings, including the type of energy carrier used in a billing period, total building living space in square meters, number of living units per building, as well as its spatial location on the postal code level.

### 3.1.1 Additional sources of data

123<sup>2</sup>

**Degree days** Degree days data was retrieved from

---

<sup>2</sup>The data originates from the energy- and billing service provider ista Deutschland GmbH and was provided to DIW Berlin for scientific use. Due to the sensitive nature of the data, it is classified as confidential and was only accessed using the internal servers at DIW Berlin. Making the data available to an external party for own scientific purposes or reproduction of results is therefore not possible.

Table 3.1: Variables and data sources

Variable		Type	Unit	Data source
<i>Demand</i> ( <i>building-level</i> )	Space heat demand	Continuous	ln(kWh/sqm)	Ista
<i>Price</i> ( <i>building-level</i> )	Energy price	Continuous	ln(cents/kWh)	Ista
<i>Billing data</i> ( <i>building-level</i> )	Heating surface	Continuous	ln(sqm)	Ista
	Energy carrier	Categorical	Gas / Oil / District heating	Ista
	Building ID	Categorical	-	Ista
	Year	Categorical	-	Ista
<i>EPC data</i> ( <i>building-level</i> )	Construction year	Continuous	-	Ista (sub-sample)
	Heating system installation year	Continuous	-	Ista (sub-sample)
	Renovation	Categorical	-	Ista (sub-sample)
<i>Climatic conditions</i> ( <i>postal code area</i> )	Degree days	Continuous	ln(degree days)	IWU (2021)
<i>Socio-economic factors</i> ( <i>district-level</i> )	Average household income	Continuous	ln(Euros/a)	Statistische Ämter (2021)
	Retirement rate	Continuous	%	DESTATIS (2021)
	Population density	Continuous	ln(inhabitants/sq. km)	DESTATIS (2021)

- As part of the project, a “Gradtagzahlen-Deutschland.xlsx” tool was developed that is based on DWD data from 1991 onwards.
- The following are determined: Monthly temperature averages, heating days, degree day figures or heating degree days for all postcodes in Germany.
- Degree day figures are determined as a weighted average from the measurement data of the three geographically closest stations
- Use of temperature measurements as well as global radiation data for different inclinations
- Correction for altitude values of the locations

Translated with [www.DeepL.com/Translator](http://www.DeepL.com/Translator) (free version)

## 3.2 Estimation methods and empirical model design

### Establishing causality

#### Endogeneity issues

Prices: Form nominal  $\rightarrow$  real

There is considerable debate in the literature whether household energy demand depends on the marginal or average price, which may differ in the presence of fixed fee and block pricing. Shin (1985) argues that households will respond to average price, which is easily calculated from the electricity bill, rather than to the actual block marginal price, which is costly to determine, and develops an empirical strategy for testing this conjecture. The average price is also used in Metcalf and Hassett (1999),<sup>5</sup> and Borenstein (2008, 2009) and Ito (2010) find no evidence that consumers “bunching up” around the block where the price changes, as one would expect if consumers truly respond to marginal price. If demand is assumed to depend on the marginal block price, then price and consumption are simultaneously determined, and instrumental variable estimation techniques must be used (Burtless and Hausman, 1978; McFadden et al., 1978; Wilder and Willenborg, 1975; Hewitt and Hanemann, 1995; Reiss and White, 2005). For lack of exact information about the block rates faced by the consumers, however, we are forced to use average price.

Zu Blockpreisen: Blockpreise stellen eine gesonderte Ausprägungsform von Paketpreisen dar, und sind beispielsweise bei Stromlieferanten vorzufinden (ebd.). Hier ändern sich die Preise je Mengenintervall, d.h. im Beispiel der Stromversorger nimmt der Preis mit steigendem Mengenintervall ab. Die Idee dahinter ist die, dass die Nachfrage nach Energieversorgung zwar mit zunehmender Haushaltsgröße steigt, die Zahlungsbereitschaft für die gewünschte Menge beim Ausgangspreis jedoch sinkt, so dass durch Blockpreise dieses Dilemma ausgeglichen werden kann und somit zusätzliche Gewinne generiert werden können (ebd., S.288).

Aus Alberini (2011): An additional concern is whether usage decisions depend on the price in the current (billing) period, on that of earlier periods, or a moving average of the prices of recent periods (Poyer and Williams, 1993). For good measure, in what follows we experiment with current price, as well as price of the previous period.

### **Frequentist estimation approaches**

In line with the previous literature (Meier & Rehdanz, 2010; Schmitz & Madlener, 2020), the fixed-effects model takes the following form:

$$\ln(D)_{it} = \alpha + \beta \ln(P)_{it} + \gamma X_{it} + \delta_i + \eta_t + \varepsilon_{it} \quad (3.1)$$

### **Bayesian estimation approaches**

### 3.3 Workflow

Since we model conditional demand only, the results should be considered as short-term elasticities. In contrast, medium- and long-term income elasticities tend to be higher. Households with increased income are likely to move into larger homes over the medium term, which would increase their heating expenditures as well (Nesbakken 1999). → Short-run elasticities as a lower bar estimate that produces a conservative estimate.

Csereklyei (2020): It is widely anticipated that responses in electricity consumption to changes in prices will be rather slow, as it takes time for energy-using durable stock to change (Miller and Alberini, 2016). Thus, short-run or annual price elasticity estimates are expected to be rather inelastic, while one anticipates significantly larger long-run estimates.

Miller and Alberini (2016) furthermore report that the use of IV estimates can change the reported elasticities by up to 80%.

All from Labanderira et al.

discrete decision to purchase durable goods that consume energy and the decision to consume energy is rarely considered. renovations may disturb the picture the panel data provides

On the other hand, most empirical studies in this area have used single-equation econometric models that require separability restrictions. This is a severe disadvantage as it is not possible to estimate cross-price effects between different energy products or consider the effects of non-energy products on the price elasticity of energy goods.

Sample period. It is widely accepted that the economic cycle has a strong influence on energy consumption due to income and (indirect cycle-related) price effects. In the case of economic crises, for example, a depression of energy prices may occur; reduced disposable income may lead agents to reduce consumption through improvements in energy efficiency, adjustments to other types of consumption or changes towards other more inexpensive energy goods.

From Aufhammer wegen IV approach: We instrument the utilities' consumer-facing prices with the weekly average spot price of natural gas at a major natural gas distribution hub in Louisiana (the Henry Hub). This instrument is valid, as we know the formula of how utilities pass-through the price (providing a strong first stage), and



the price is determined prior to within-bill consumption (strengthening the exclusion restriction).

No change in the energy taxation (Energiesteuerergesetz) in the observed period; letzte Neufassung aus 2006 mit Effekten für 2007

Miller Alberini (2016): We also find that aggregating our data can result in both higher and lower price elasticity estimates, depending on the dataset used, and that controlling for unit-level fixed effects with panel data generally results in more inelastic demand functions. Addressing the endogeneity of price and/or measurement error in price with instrumental variables has a small but noticeable effect on the price elasticities. Finally, controlling for housing characteristics and capital stock produces a lower price elasticity.

# Chapter 4

## Results

### 4.1 Descriptive statistics

### 4.2 Full sample analysis

Notes (copied from other papers):

-Accounting for unobserved heterogeneity reduces the elasticities by 15% to 32% relative to a model that does not include any effects at all, where the own price elasticity of electricity demand is -1.).

### 4.3 Subsample analysis

Table 4.2: Number of flight connections per destination

Destination	Number of connections
Albuquerque International Sunport	30
Bob Hope	261
Boise Air Terminal	29
Charlotte Douglas Intl	45
Chicago Midway Intl	104
Chicago Ohare Intl	524
Dallas Fort Worth Intl	441
Denver Intl	905
Detroit Metro Wayne Co	55

Table 4.2: Number of flight connections per destination (*continued*)

Destination	Number of connections
General Edward Lawrence Logan Intl	70
George Bush Intercontinental	228
Hartsfield Jackson Atlanta Intl	410
Honolulu Intl	180
John F Kennedy Intl	137
John Wayne Arpt Orange Co	247
Kahului	171
Kansas City Intl	89
Klamath Falls Airport	82
Kona Intl At Keahole	90
Lihue	52
Long Beach	263
Los Angeles Intl	912
Mahlon Sweet Fld	189
Mc Carran Intl	564
Metropolitan Oakland Intl	451
Minneapolis St Paul Intl	270
Newark Liberty Intl	91
Norman Y Mineta San Jose Intl	574
Ontario Intl	196
Palm Springs Intl	105
Phoenix Sky Harbor Intl	888
Reno Tahoe Intl	94
Roberts Fld	252
Ronald Reagan Washington Natl	87
Sacramento Intl	404
Salt Lake City Intl	538
San Diego Intl	265
San Francisco Intl	1265
Santa Barbara Muni	89
Seattle Tacoma Intl	569
Ted Stevens Anchorage Intl	254
Tucson Intl	90
Washington Dulles Intl	89

<sup>1</sup> This table was created based on the flights dataset.<sup>2</sup> Source: R Packages.

Table 4.1: Summary statistics

Variable	Unit	Median (IQR)
Space heat consumption, eff.	[kWh/sqm.]	113 (86, 145)
Energy price, real	[Cents/kWh]	6.49 (5.77, 7.69)
Degree days		3,446 (3,214, 3,733)
Building heating surface	[sqm]	404 (260, 707)
Building housing units		6 (3, 10)
District household income	[€/a]	20,695 (18,786, 22,568)
District retirement share	[%]	0.207 (0.193, 0.220)
District population density	[inhabitants/sq. km]	572 (217, 1,960)
<i>Energy carrier group</i>		
Gas		1,647,563 (61%)
Oil		802,451 (30%)
District heating		268,232 (9.9%)
<i>(N = 2,718,246; No missings)</i>		

# Chapter 5

## Discussion

NOTES - Energy taxation and climate change: - Fuel consumption demand is determined by demand functions that depend primarily on income and prices - Those who disfavor fuel taxes often claim they are strongly regressive. Earlier studies have shown that this depends on the country studied and on the details of the methods used, for instance if lifetime or temporary income is used, if substitution or other reactions are allowed for in the analysis. There is a tendency to progressivity in low income countries but regressivity in high income countries. - Indirect channel of long-term price change and anticipation of those changes leading to investments in low-carbon UBA: Remediation measures drawn by lot through the BEHG are partly subsidized through the BEG or the tax incentive and are accounted for there (CO<sub>2</sub> price as door opener for subsidy)

# Chapter 6

## Conclusion

(2.1)

# Appendix A

## First Appendix

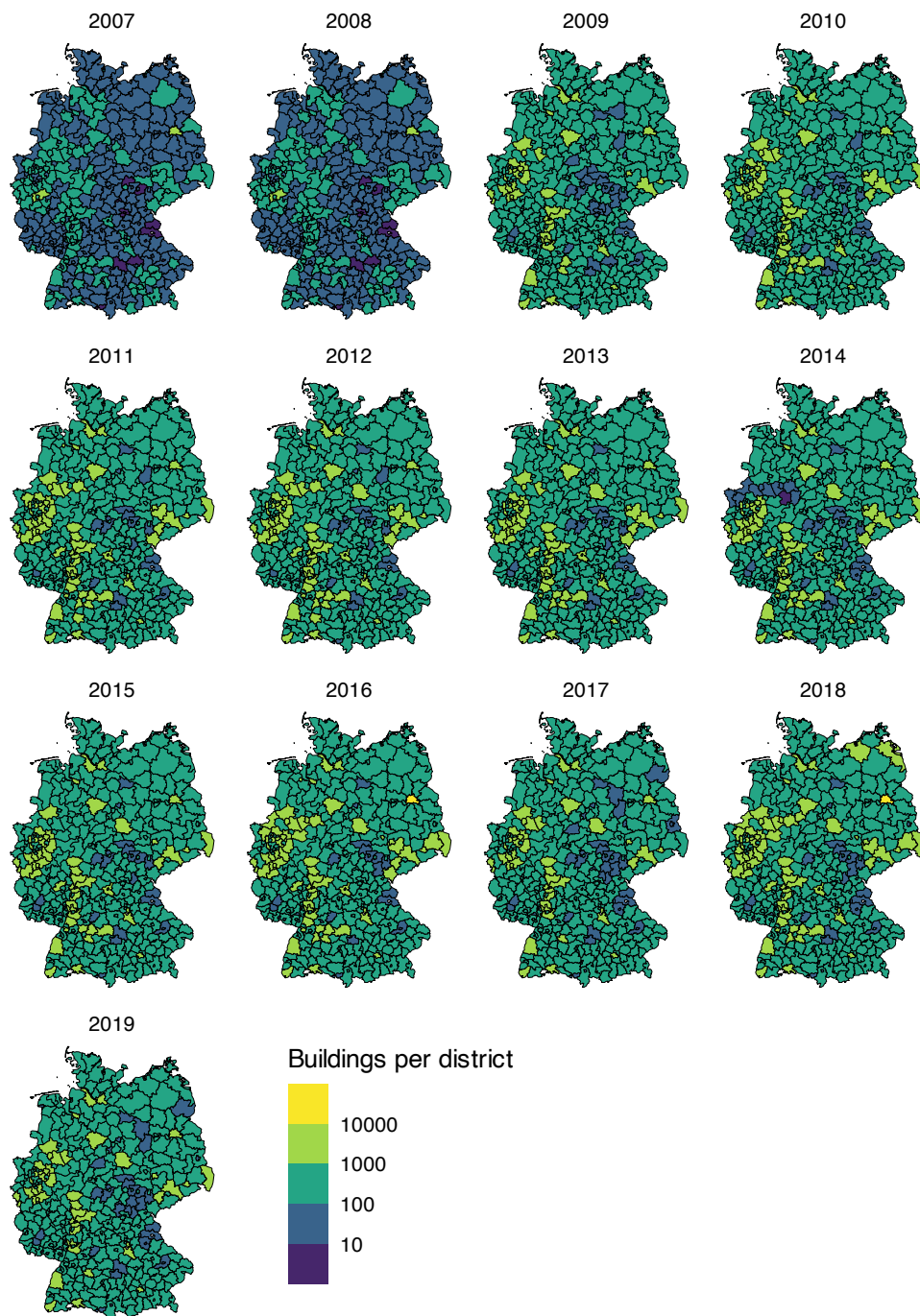


Figure A.1: Spatial and temporal distribution of buildings observed



# Declaration of Authorship (Eigenständigkeitserklärung)

Hiermit erkläre ich, dass ich die vorliegende Arbeit selbständig verfasst habe und sämtliche Quellen, einschließlich Internetquellen, die unverändert oder abgewandelt wiedergegeben werden, insbesondere Quellen für Texte, Grafiken, Tabellen und Bilder, als solche kenntlich gemacht habe.

Ich versichere, dass ich die vorliegende Abschlussarbeit noch nicht für andere Prüfungen eingereicht habe.

Mir ist bekannt, dass bei Verstößen gegen diese Grundsätze ein Verfahren wegen Täuschungsversuchs bzw. Täuschung gemäß der fachspezifischen Prüfungsordnung und/oder der Fächerübergreifenden Satzung zur Regelung von Zulassung, Studium und Prüfung der Humboldt-Universität zu Berlin (ZSP-HU) eingeleitet wird.

Berlin, den 28.02.2022

---

*Ort, Datum*

---

*Unterschrift*

# References

- AGEB. (2021). *Auswertungstabellen zur Energiebilanz 1990 bis 2020*. Berlin. Retrieved from <https://ag-energiebilanzen.de/daten-und-fakten/auswertungstabellen/>
- Alberini, A., Gans, W., & Velez-Lopez, D. (2011). Residential consumption of gas and electricity in the U.S.: The role of prices and income. *Energy Economics*, 33(5), 870–881. <http://doi.org/10.1016/j.eneco.2011.01.015>
- Aldy, J. E., Krupnick, A. J., Newell, R. G., Parry, I. W. H., & Pizer, W. A. (2010). Designing Climate Mitigation Policy. *Journal of Economic Literature*, 48(4), 903–934. <http://doi.org/10.1257/jel.48.4.903>
- Auffhammer, M., & Rubin, E. (2018). *Natural Gas Price Elasticities and Optimal Cost Recovery Under Consumer Heterogeneity: Evidence from 300 million natural gas bills* (No. w24295) (p. w24295). Cambridge, MA: National Bureau of Economic Research. Retrieved from <http://www.nber.org/papers/w24295.pdf>
- BMWi. (2021). *Gesamtausgabe der Energiedaten*. BMWi. Retrieved from [https://www.bmwi.de/Redaktion/DE/Binaer/Energiedaten/energiedaten-gesamt-xls.xlsx?\\_\\_blob=publicationFile&v=133](https://www.bmwi.de/Redaktion/DE/Binaer/Energiedaten/energiedaten-gesamt-xls.xlsx?__blob=publicationFile&v=133)
- Bürger, V., Hesse, T., Köhler, B., Palzer, A., & Engelmann, P. (2019). German Energiewende—different visions for a (nearly) climate neutral building sector in 2050. *Energy Efficiency*, 12(1), 73–87. <http://doi.org/10.1007/s12053-018-9660-6>
- Csereklyei, Z. (2020). Price and income elasticities of residential and industrial electricity demand in the European Union. *Energy Policy*, 137, 111079. <http://doi.org/10.1016/j.enpol.2019.111079>
- Cutler, H. A. (1941). The Elasticity of the Residential Demand for Electricity. *The Journal of Land & Public Utility Economics*, 17(2), 242–245. <http://doi.org/10.2307/3158428>

- DESTATIS. (2018). Fernwärmeversorgung 2017: Abgegebene Wärmemenge leicht gesunken. Retrieved January 19, 2022, from [https://www.destatis.de/DE/Presse/Pressemitteilungen/2018/11/PD18\\_434\\_434.html](https://www.destatis.de/DE/Presse/Pressemitteilungen/2018/11/PD18_434_434.html)
- ERK. (2020). *Bericht zur Vorjahresschätzung der deutschen Treibhausgasemissionen für das Jahr 2020*. Retrieved from [https://expertenrat-klima.de/content/uploads/2021/04/210415\\_Bericht\\_Expertenrat\\_Klimafragen\\_2021-2.pdf](https://expertenrat-klima.de/content/uploads/2021/04/210415_Bericht_Expertenrat_Klimafragen_2021-2.pdf)
- European Commission. (2019). *The European Green Deal* (No. COM(2019) 640 final). Brussels. Retrieved from <https://eur-lex.europa.eu/legal-content/EN/TXT/HTML/?uri=CELEX:52019DC0640&from=EN>
- Gwartney, J. D. (1976). Demand and Consumer Choice. In *Economics Private and Public Choice* (pp. 289–309). Elsevier. <http://doi.org/10.1016/B978-0-12-311050-3.50021-8>
- Hennes, O., Jeddi, S., Madlener, R., Schmitz, H., Wagner, J., Wolff, S., & Zinke, J. (2021). Auswirkungen von CO<sub>2</sub>-Preisen auf den Gebäude-, Verkehrs- und Energiesektor. *Zeitschrift für Energiewirtschaft*, 45(2), 91–107. <http://doi.org/10.1007/s12398-021-00305-0>
- Houthakker, H. S. (1951). Some Calculations on Electricity Consumption in Great Britain. *Journal of the Royal Statistical Society. Series A (General)*, 114(3), 359–371. <http://doi.org/10.2307/2980781>
- Labandeira, X., Labeaga, J. M., & López-Otero, X. (2017). A meta-analysis on the price elasticity of energy demand. *Energy Policy*, 102, 549–568. <http://doi.org/10.1016/j.enpol.2017.01.002>
- Leth-Petersen, S., & Togeby, M. (2001). Demand for space heating in apartment blocks: measuring effects of policy measures aiming at reducing energy consumption. *Energy Economics*, 23(4), 387–403. [http://doi.org/10.1016/S0140-9883\(00\)00078-5](http://doi.org/10.1016/S0140-9883(00)00078-5)
- Levesque, A., Pietzcker, R. C., Baumstark, L., & Luderer, G. (2021). Deep decarbonisation of buildings energy services through demand and supply transformations in a 1.5°C scenario. *Environmental Research Letters*, 16(5), 054071. <http://doi.org/10.1088/1748-9326/abdf07>
- Loulou, R., & Labriet, M. (2008). ETSAP-TIAM: the TIMES integrated assessment model Part I: Model structure. *Computational Management Science*, 5(1-2), 7–40. <http://doi.org/10.1007/s10287-007-0046-z>

- Meier, H., & Rehdanz, K. (2010). Determinants of residential space heating expenditures in Great Britain. *Energy Economics*, 32(5), 949–959. <http://doi.org/10.1016/j.eneco.2009.11.008>
- Metcalf, G. E., & Hassett, K. A. (1999). Measuring the Energy Savings from Home Improvement Investments: Evidence from Monthly Billing Data. *Review of Economics and Statistics*, 81(3), 516–528. <http://doi.org/10.1162/003465399558274>
- Miller, M., & Alberini, A. (2016). Sensitivity of price elasticity of demand to aggregation, unobserved heterogeneity, price trends, and price endogeneity: Evidence from U.S. Data. *Energy Policy*, 97, 235–249. <http://doi.org/10.1016/j.enpol.2016.07.031>
- Pindyck, R. S., & Rubinfeld, D. L. (2018). *Microeconomics* (Ninth edition). NY NY: Pearson.
- Rehdanz, K. (2007). Determinants of residential space heating expenditures in Germany. *Energy Economics*, 29(2), 167–182. <http://doi.org/10.1016/j.eneco.2006.04.002>
- RWI. (2021). *Anwendungsbilanzen 2020 für den Sektor der privaten Haushalte und den Verkehrssektor in Deutschland*. Retrieved from [https://ag-energiebilanzen.de/wp-content/uploads/2020/10/rwi\\_anwendungsbilanz\\_2020\\_\\_priv.\\_hh\\_und\\_verkehr\\_\\_vorl.\\_eb.pdf](https://ag-energiebilanzen.de/wp-content/uploads/2020/10/rwi_anwendungsbilanz_2020__priv._hh_und_verkehr__vorl._eb.pdf)
- Schmitz, H., & Madlener, R. (2020). Heterogeneity in price responsiveness for residential space heating in Germany. *Empirical Economics*, 59(5), 2255–2281. <http://doi.org/10.1007/s00181-019-01760-y>
- Schulte, I., & Heindl, P. (2017). Price and income elasticities of residential energy demand in Germany. *Energy Policy*, 102, 512–528. <http://doi.org/10.1016/j.enpol.2016.12.055>
- Stede, J., Schütze, F., & Wietschel, J. (2020). Wärmemonitor 2019: Klimaziele bei Wohngebäuden trotz sinkender CO<sub>2</sub>-Emissionen derzeit außer Reichweite (Version 2.0). *DIW Wochenbericht*. [http://doi.org/10.18723/DIW\\_WB:2020-40-1](http://doi.org/10.18723/DIW_WB:2020-40-1)
- Stiglitz, J. (2019). *Addressing Climate Change through Price and Non-Price Interventions* (No. w25939) (p. w25939). Cambridge, MA: National Bureau of Economic Research. Retrieved from <http://www.nber.org/papers/w25939.pdf>
- Techem. (2019). *Energiekennwerte 2019*. Eschborn: Techem. Retrieved from <https://www.techem.com/content/dam/techem/newsroom/studien/Techem-Energiekennwerte-Studie-2019.pdf>

- UBA. (2016). *CO<sub>2</sub>-Emissionsfaktoren für fossile Brennstoffe*. Retrieved from [https://www.umweltbundesamt.de/sites/default/files/medien/1968/publikationen/co2-emissionsfaktoren\\_fur\\_fossile\\_brennstoffe\\_korrektur.pdf](https://www.umweltbundesamt.de/sites/default/files/medien/1968/publikationen/co2-emissionsfaktoren_fur_fossile_brennstoffe_korrektur.pdf)
- Varian, H. R. (2010). *Intermediate microeconomics: a modern approach* (8. ed). New York, NY: Norton.