

# E-CREDA 40' Research Update

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The price of clean air: Quantifying air pollution exposure in real estate investment decisions

By PD Dr. Marcelo Cajias – PATRIZIA SE / IREBS

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20<sup>th</sup> of April

15:00 – 15:40 CET

Online seminar

E-CREDA improves the access to and understanding of available CRE data to facilitate higher quality applied research and more frequent interaction between investors, data providers and academics to achieve data parity with other major asset classes, like stocks and bonds



INVESTMENT STRATEGY & RESEARCH

# RESEARCH BRIEF

The price of clean air – Quantifying air pollution exposure in real estate decisions

PD Dr. Marcelo Cajias  
M.Sc. Rebecca Restle

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AIR POLLUTION  
A HIDDEN THREAT IN  
URBAN AREAS



02

AIR POLLUTION AND  
HEDONIC MODELLING

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AIR POLLUTION  
MATTERS AND AFFECTS  
RENT LEVELS



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KEY TAKEAWAYS

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# Air pollution – a hidden threat in Berlin?



Source: PATRIZIA, Google

# Air pollution – a hidden threat in Berlin?

**Best air quality**

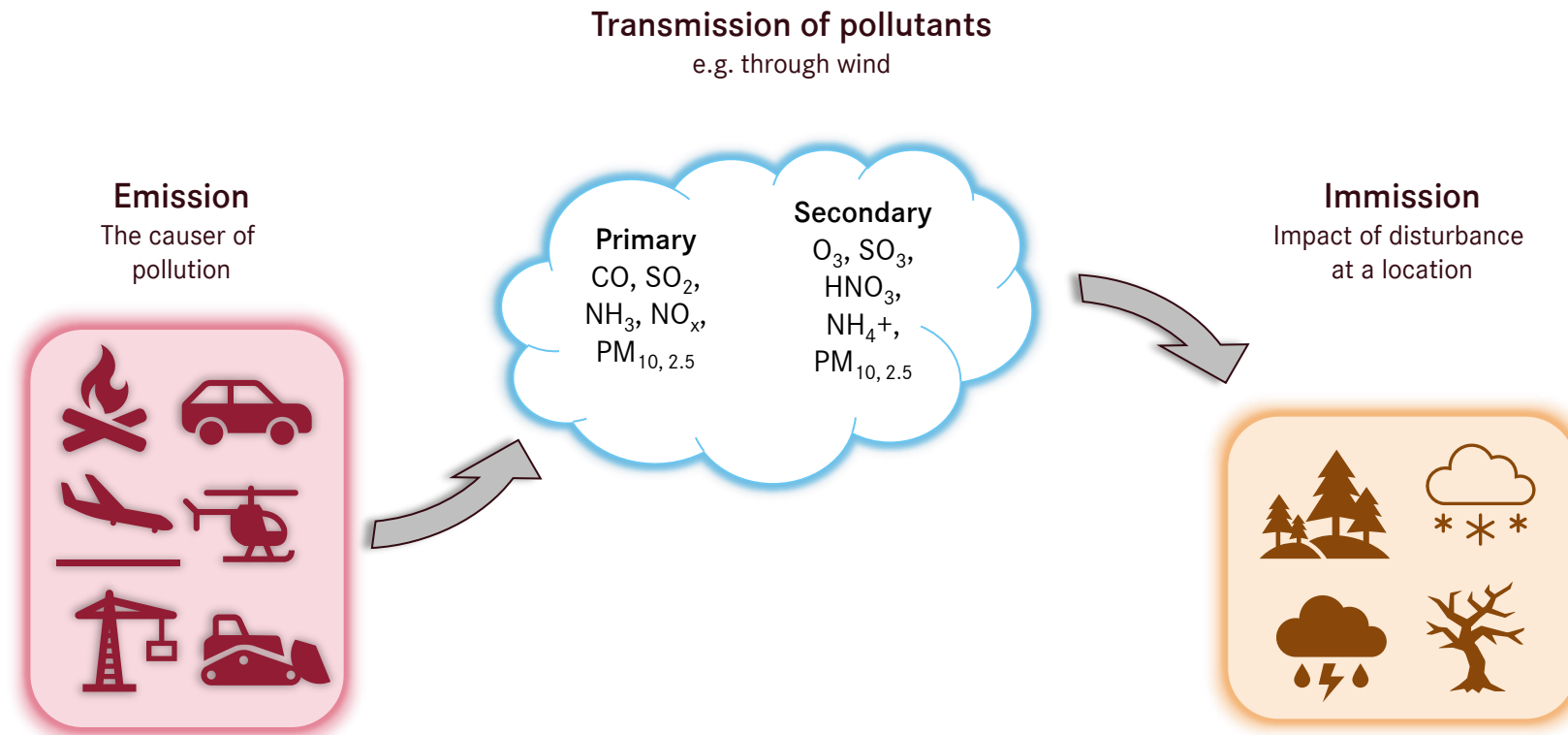


**Worst air quality**



Source: PATRIZIA, Google

# Air pollution: What is it and where it comes from?



Source: PATRIZIA

# Why does air pollution matter and what are the effects on health and environment?

## Risk and size of particulates

- For particulates: health risks can be caused by the dust itself as it can damage the organs depending on the size.
- The bigger the particle, the more likely it is for the dust to be exhaled or stay in the nose.
- The smaller the particle, the more dangerous, as they can penetrate deeper organs, e.g., lungs and bronchial tubes.
- Heavy metals can adhere to the surface of the particles, which would produce toxic air. Short-term studies showed higher risks of hypertension and heart rhythm variability for high concentrations.

**In 2016, it was estimated that  
~37,000 premature deaths in  
Germany were attributable to air  
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## Causality and evidence

- Causality is not always conclusively evident: It is important to note that even if most epidemiological studies found significant relationships between certain diseases and air pollution.
- Harmfulness of each component is not always possible to understand as there is exposure to a combination of substances (Brook et al. 2010).
- Fine particulates PM2.5 have the most severe health effects, and short- and long-term exposure have effects on the human body, European Environment Agency (EEA 2019).

**Long-term studies revealed a serious risk of organ damage, but also short-term effects on the cardiovascular system can occur**



# Why does air pollution matter and what are the effects on health and environment?

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## We only touch the surface

- Greenhouse gas CO2 is a prominent contributor to climate change. But pollutants like NO2 or SO2 also cause effects on ecosystem through acid rain (US EPA 2016).
- Black carbon is relevant: High warming potential due to its ability to absorb sunlight. It can influence regional cloud formation and rainfall patterns.

Environmental impacts include acid rain and contribution to the greenhouse effect

# The causes of high pollution levels reveal where we need to act

## Mitigation chances

With the rise of the electric-powered car or other alternatives, traffic will have a smaller share in air pollution in the next few years (Eichlseder et al. 2021).

The European Commission (2019) claims that half of particulate matter pollution in the EU comes from heating systems. This points to seasonal variations in concentration levels.

Responsible for decreasing pollution concentrations are vegetation, distance to emission sources, altitude, and wind speed

## Negative effects

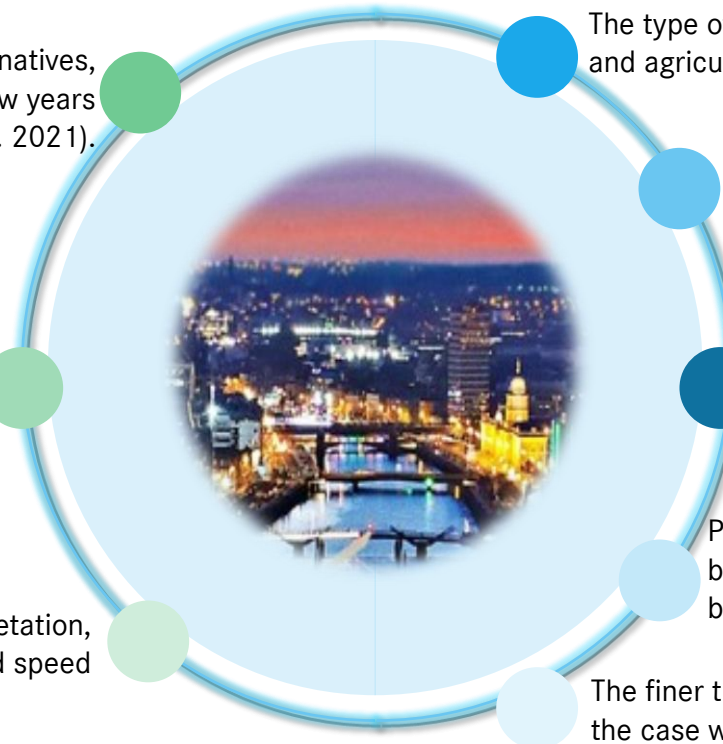
The type of heating has a negative impact, as do construction sites and agriculture.

Wind speed can have a strong influence, followed by air pressure, humidity, and temperature, depending on the season (Yang et al. 2020).

The combustion of fuels mainly causes PM10 and NO<sub>2</sub>, in addition to the chemical reaction of oxygen gas with ambient nitrogen gas.

PM10 is negatively correlated with humidity since the particles become heavier as they absorb water and can therefore no longer be distributed (Rumaling et al. 2021).

The finer the particle size, the less affected by humidity, which is the case with PM2.5 (Munir et al. 2017).



# The willingness to pay for clean air – international empirical evidence



The hedonic regression revealed that in the city of Nantes, noise has a significant negative impact on rent prices, air pollution does not. Pollution values were derived from an *Atmospheric Dispersion Modelling System* (Le Boennec and Salladarré 2017) .

Sun and Yang (2020) analyzed the behavior of  $PM_{2.5}$  on housing prices in several Chinese cities. They conclude that there is a negative relationship between  $PM_{2.5}$  and housing prices and that cities with rapid growth and increased industrial activity experience a real estate boom.

In Quito, Ecuador Borja-Urbano et al. 2021 found a significant negative relationship between housing price and air pollutants like CO,  $PM_{2.5}$ , and  $NO_2$  values. Based on their results an increase of 1% of  $PM_{2.5}$  leads to a reduction of 15 US\$/m<sup>2</sup> ceteris paribus.

# Our statistical approach: everything is related to everything else, but near things are more related than distant things

## Measurement of the dynamics behind the distribution of pollutants is complex

**Approach 1:** Use physical and chemical processes to calculate the distribution starting from the pollution source (e.g., Knote et al. 2015).

**Approach 2:** estimate values by applying geoinformatic and geostatistical interpolation methods on the measurement network through the city (Deligiorgi and Philippopoulos 2011).

## The selected method: Spatial interpolation

"[...] the process of using points with known values to estimate values at other unknown points" (QGIS Project 2022).

## The transformation equation

$$\hat{X}(s_0) = \sum_{i=1}^N \lambda_i X(s_i)$$

The predicted position is indicated by  $s_0$ .  $N$  is the number of measured values and  $X(s_i)$  is the observed value at Position  $i$ , with  $\lambda_i$  as an unknown weight.

More weight  $\lambda_i$  is given to points closer than those located further away; this is known as the inverse distance weighting.

Thus, the weighting is solely dependent on the distance to the predicted location, and the output is a surface raster with weighted averages.

# The hedonic model for isolating price differentials to changes in air quality

## The semiparametric hedonic model

$$\log(\text{Rent}_{i,t}) = \sum_{p=1}^P s_p(\log(X_{p,i,t})) + \sum_{b=1}^B \beta_b \cdot X_{b,i,t} + \sum_{m=1}^M s_m(\log(X_{m,i,t})) + \sum_{d=1}^D s_d(\log(X_{d,i,t})) + \sum_{l=1}^L s_l(X_{l,i,t}) + \text{error}$$

$\sum_{p=1}^P s_p(\log(X_{p,i,t})) +$  Interpolated pollutant variables  
 $\sum_{b=1}^B \beta_b \cdot X_{b,i,t} +$  Binary hedonic characteristics  
 $\sum_{m=1}^M s_m(\log(X_{m,i,t})) +$  Metric hedonic characteristics  
 $\sum_{d=1}^D s_d(\log(X_{d,i,t})) +$  Distance to relevant amenities  
 $\sum_{l=1}^L s_l(X_{l,i,t}) +$  Socioeconomic variables  
 error Estimation error

Variable	Unit	Mean 2018	Mean 2021
Refurbished	Binary	0.18	0.13
Built-in kitchen	Binary	0.51	0.48
Balcony	Binary	0.63	0.63
Parking	Binary	0.22	0.21
Elevator	Binary	0.40	0.49
Terrace	Binary	0.13	0.12

The apartment contains a bathtub, a built-in-kitchen, a balcony, but neither a parking slot nor an elevator or a terrace.

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





Variable	Unit	Mean	Mean
		2018	2021
Rent	EUR/month	788.41	788.31
Size	sqm	70.88	66.31
Rent sqm	EUR/sqm	10.94	11.68
Age (relative to 2017)	Integer	63.6	57.77

An average apartment in 2018 has an asking rent of 10.94 EUR/sqm/p.m. and is 64 years old relative to 2017.

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Variable	Unit	Mean	Mean
		2018	2021
City centre distance	Km.	7.90	7.60
Bus stop	Km.	0.19	0.19
Post box	Km.	0.21	0.19
Bar	Km.	0.48	0.53
School	Km.	0.32	0.33
Fast food	Km.	0.31	0.31
Park	Km.	0.27	0.27
Restaurant	Km.	0.23	0.25
Café	Km.	0.37	0.39
Playground	Km.	0.19	0.18
Supermarket	Km.	0.28	0.29
Atm	Km.	0.54	0.55

On average, it is 7.9 km away from the CBD, 370 meters to the nearest café and 280 meters to the closest supermarket. The bus station is 190 m away, whereas the nearest school is located 320 meters nearby.

# The hedonic model for isolating price differentials to changes in air quality

## The semiparametric hedonic model

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Variable	Unit	Mean	Mean
		2018	2021
Purchasing Power	EUR/HH/ZIP	36.2k	38.3k

Source: PATRIZIA, Valuemarktdaten



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Variable	Unit	Mean	Mean
		2018	2021
PM10	$\mu\text{m}/\text{m}^3$	17.35	14.32
NO2	$\mu\text{m}/\text{m}^3$	25.00	16.37

Source: PATRIZIA, Valuemarktdaten

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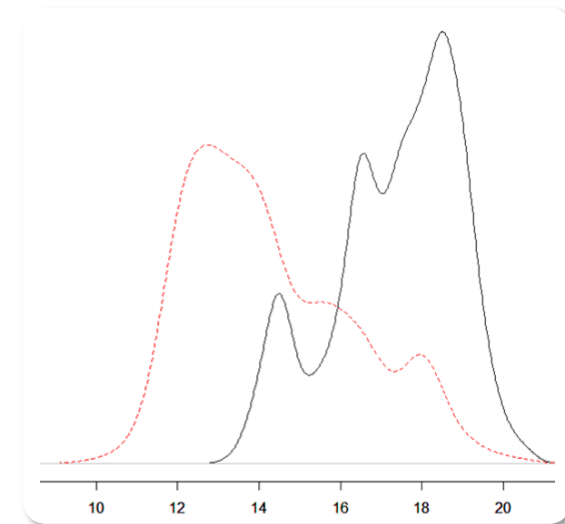
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PM10 concentration  
2018 and 2021



Source: PATRIZIA, Valuemarktdaten

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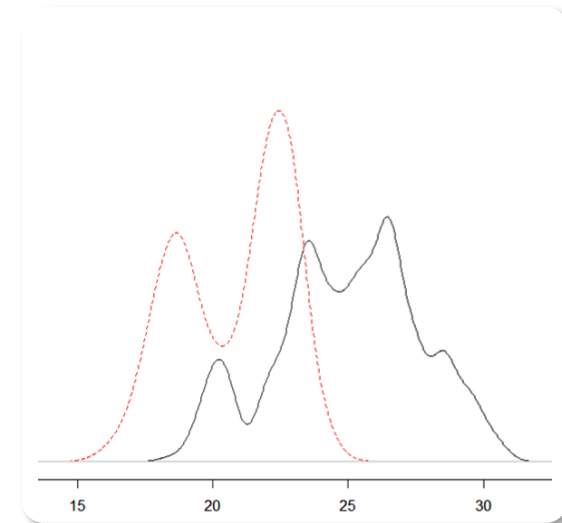
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NO2 concentration  
2018 and 2021



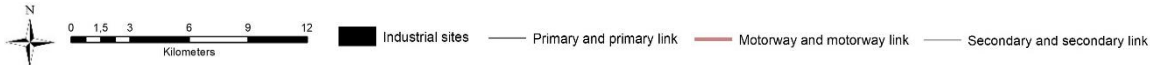
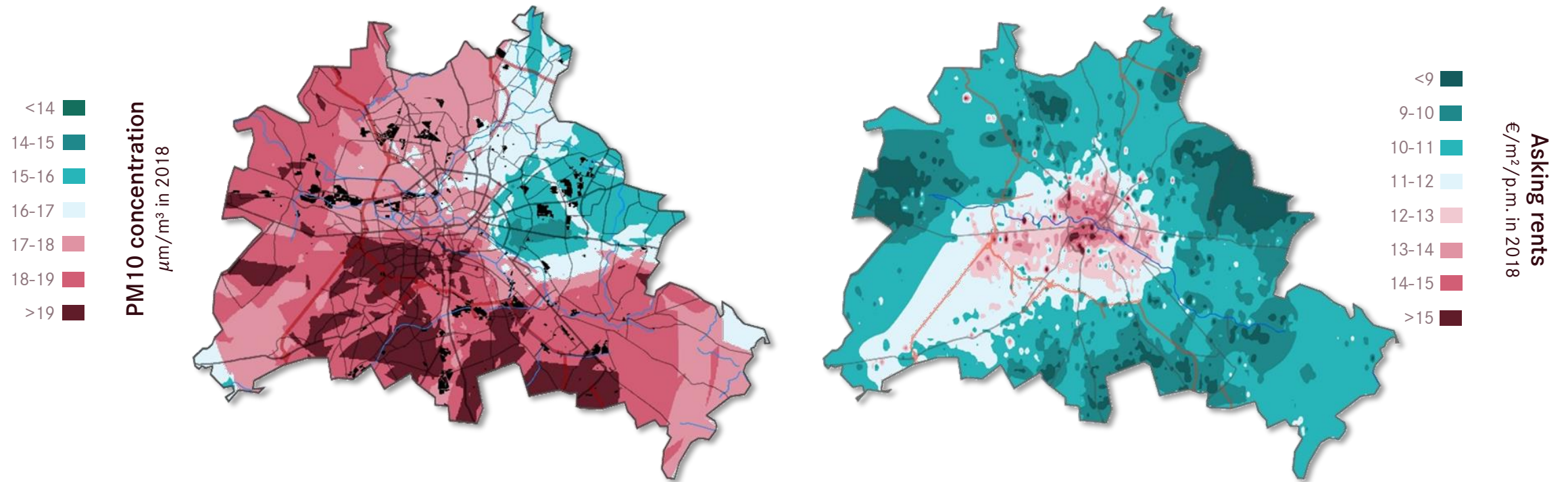
Source: PATRIZIA, Valuemarktdaten

# The hedonic model explains the willingness to pay for clean air

## The hedonic model

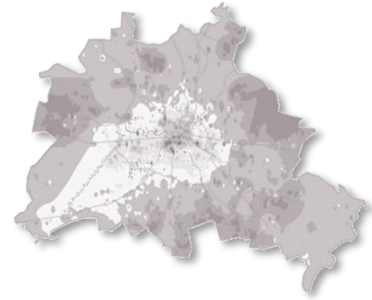
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# Particulate matter and rents 2018

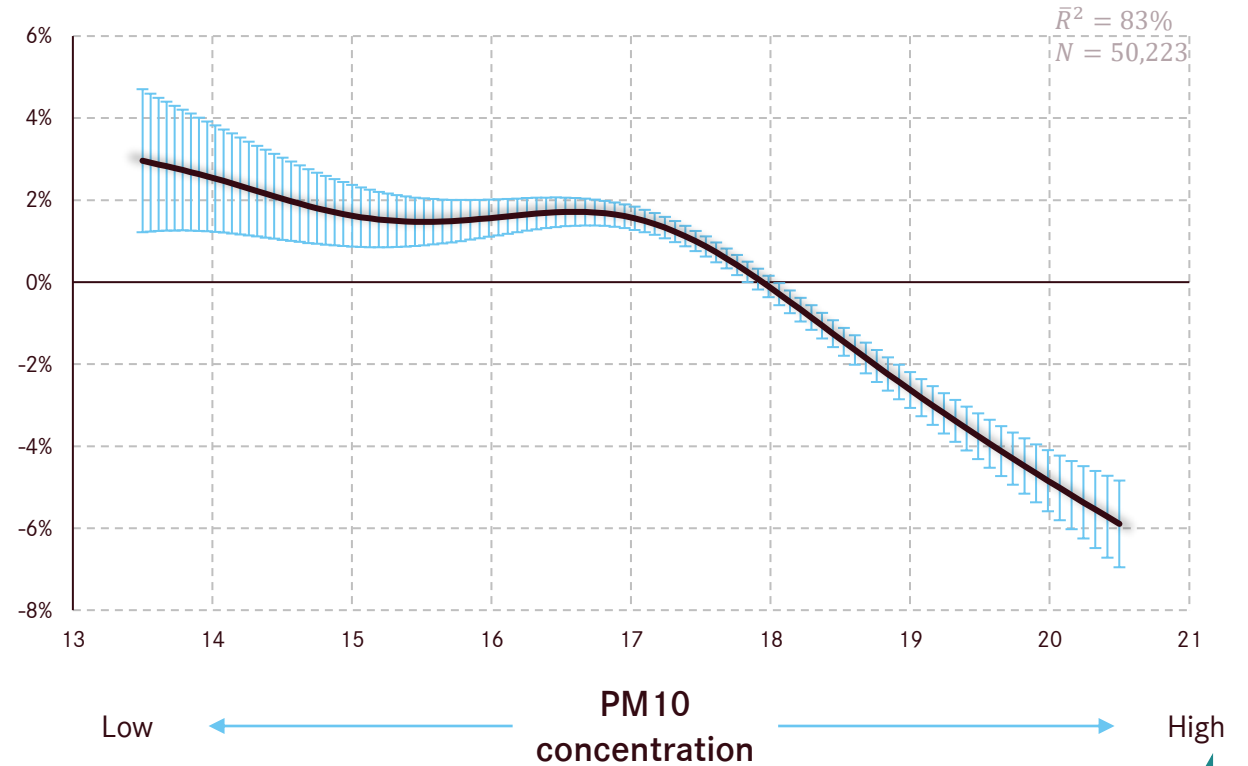


Source: PATRIZIA, Valuemarktdaten

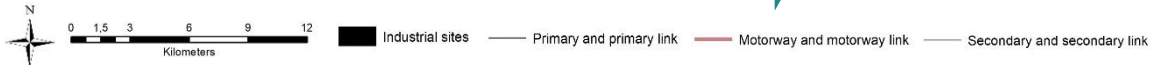
# Particulate matter and rents 2018



Premium  
↑  
Rent deviation from market average in % from €/p.m.  
↓  
Discount

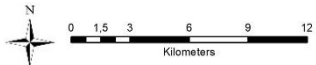
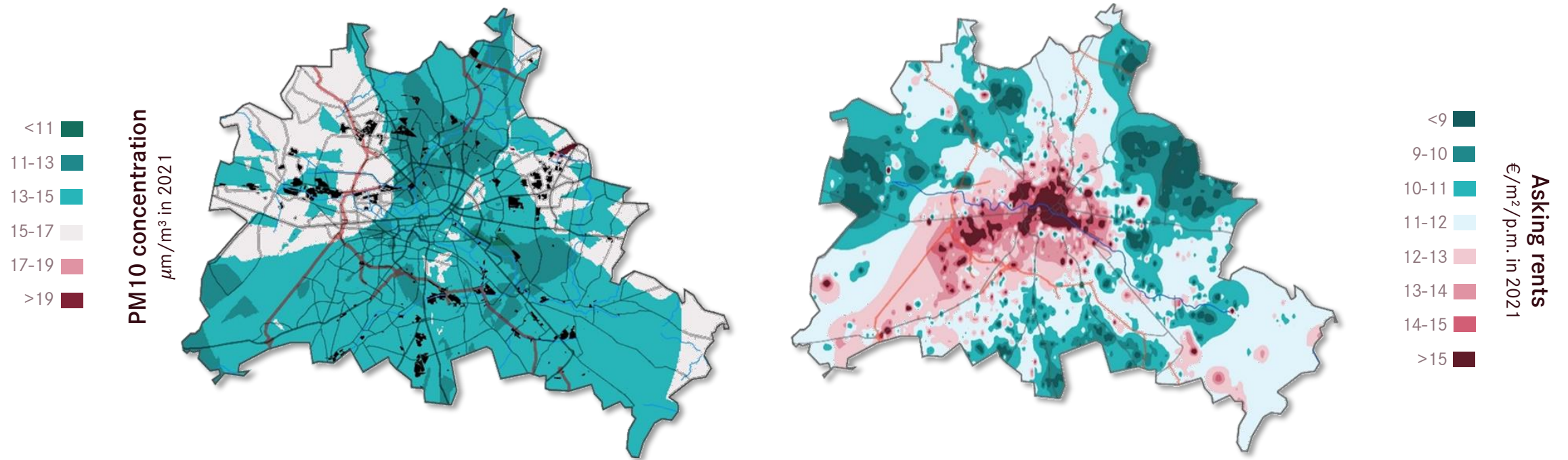


The rental income of apartments located in areas with low pollution was 2% higher than market average. Assets in areas with high pollution were offered at a discount between -2% and 6% in comparison with average assets.



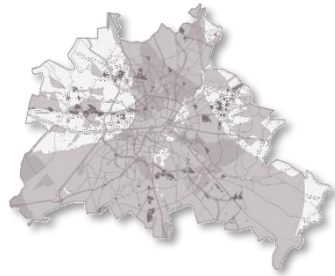
Source: PATRIZIA, Valuemarktdaten

# Particulate matter and rents 2021

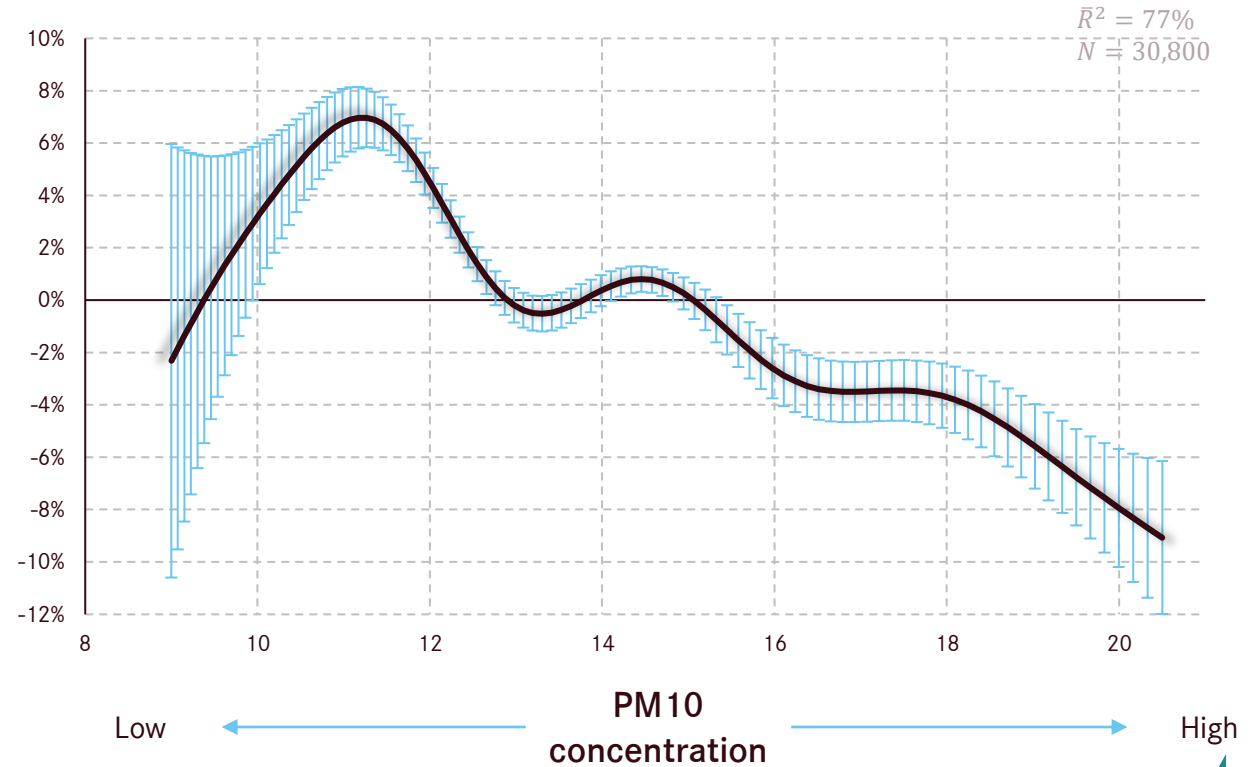


Source: PATRIZIA, Valuemarktdaten

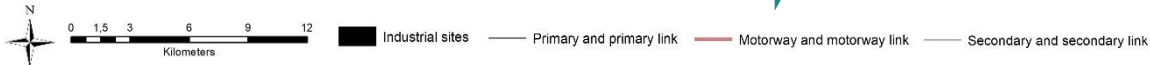
# Particulate matter and rents 2021



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Discount



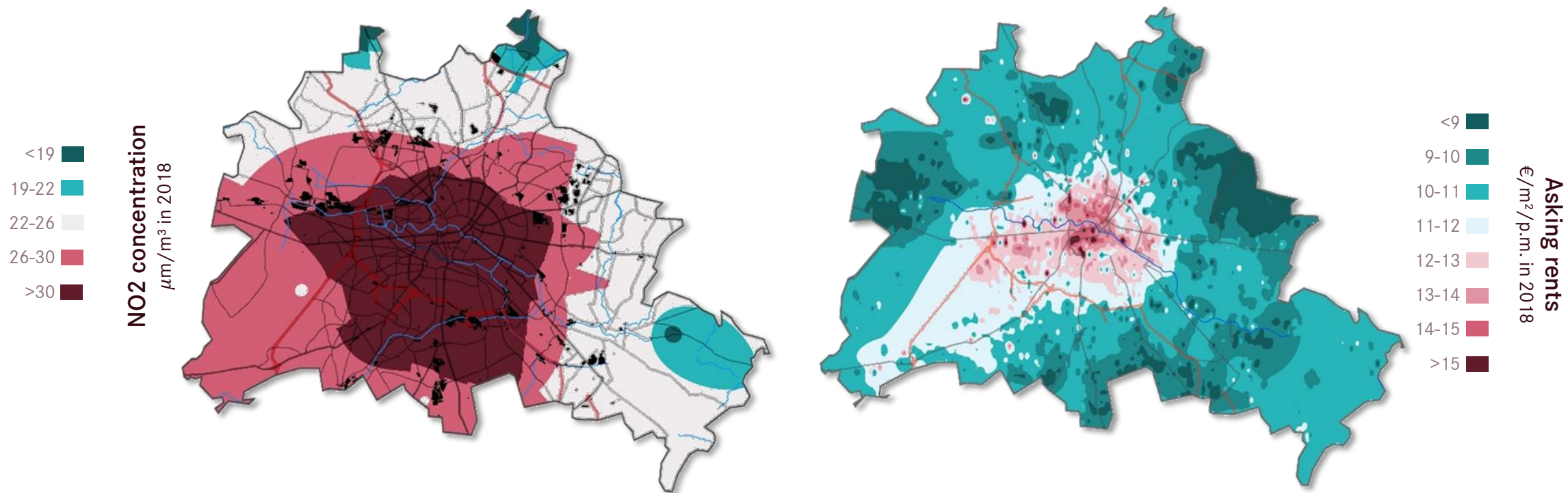
Assets located in areas with low pollution were offered at either market average or a premium of 6% relative to the average market. Assets in high polluted areas were traded in 2021 at a discount between -4% until -8%



Source: PATRIZIA, Valuemarktdaten

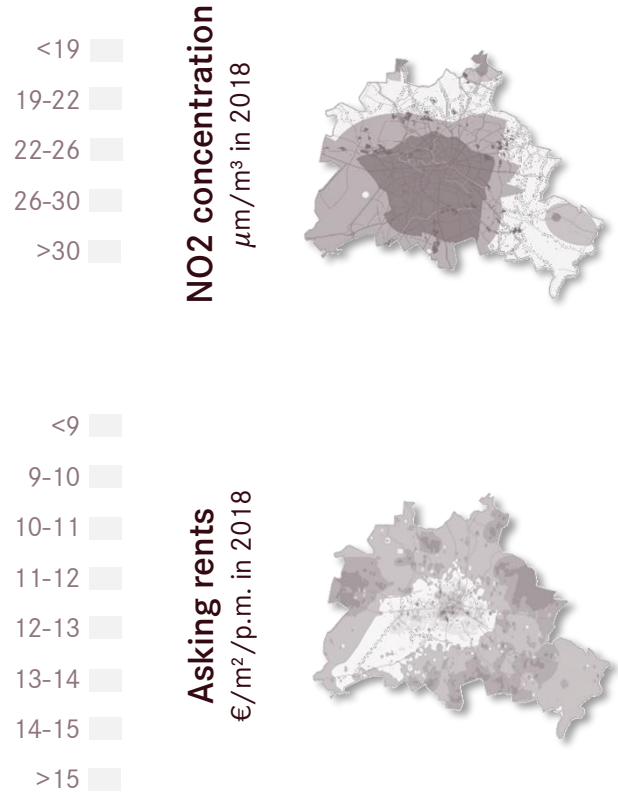


# Nitrogen dioxide and rents 2018

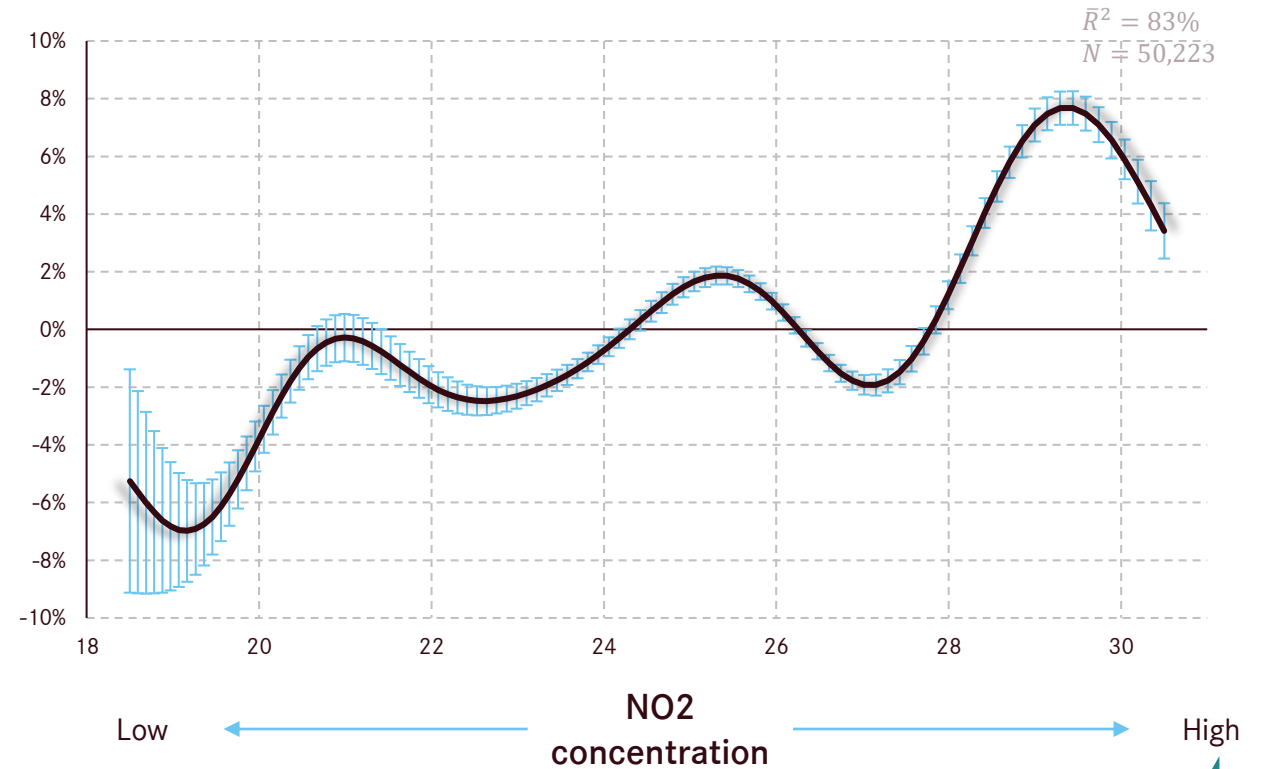


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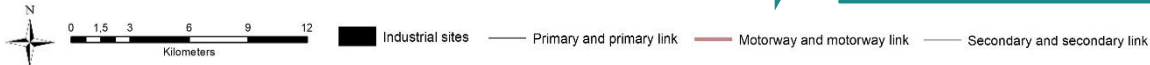
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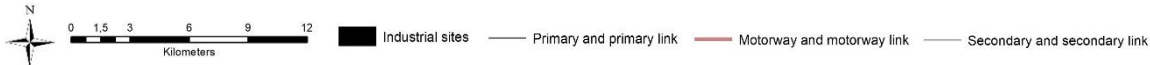
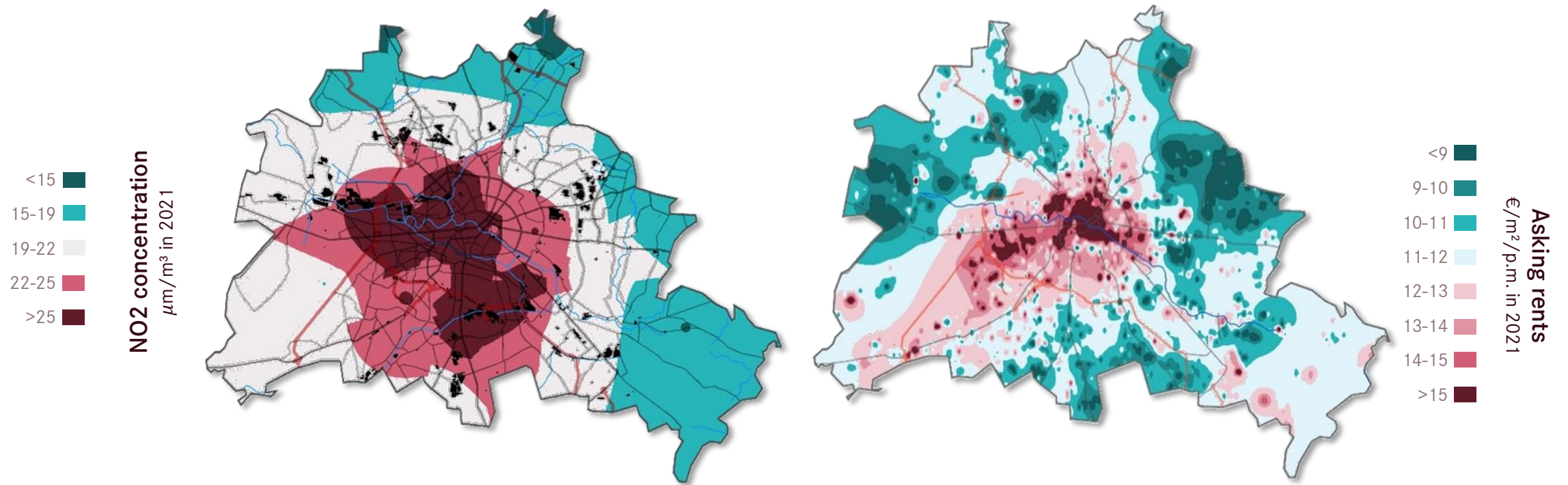


**Market failure:** In 2018, apartments in areas with low traffic and combustion from motor vehicles were offered at a discount of -2% until -6%, whilst assets in areas with high nitrogen dioxide levels showed a premium of less than 8% relative to the market average levels.



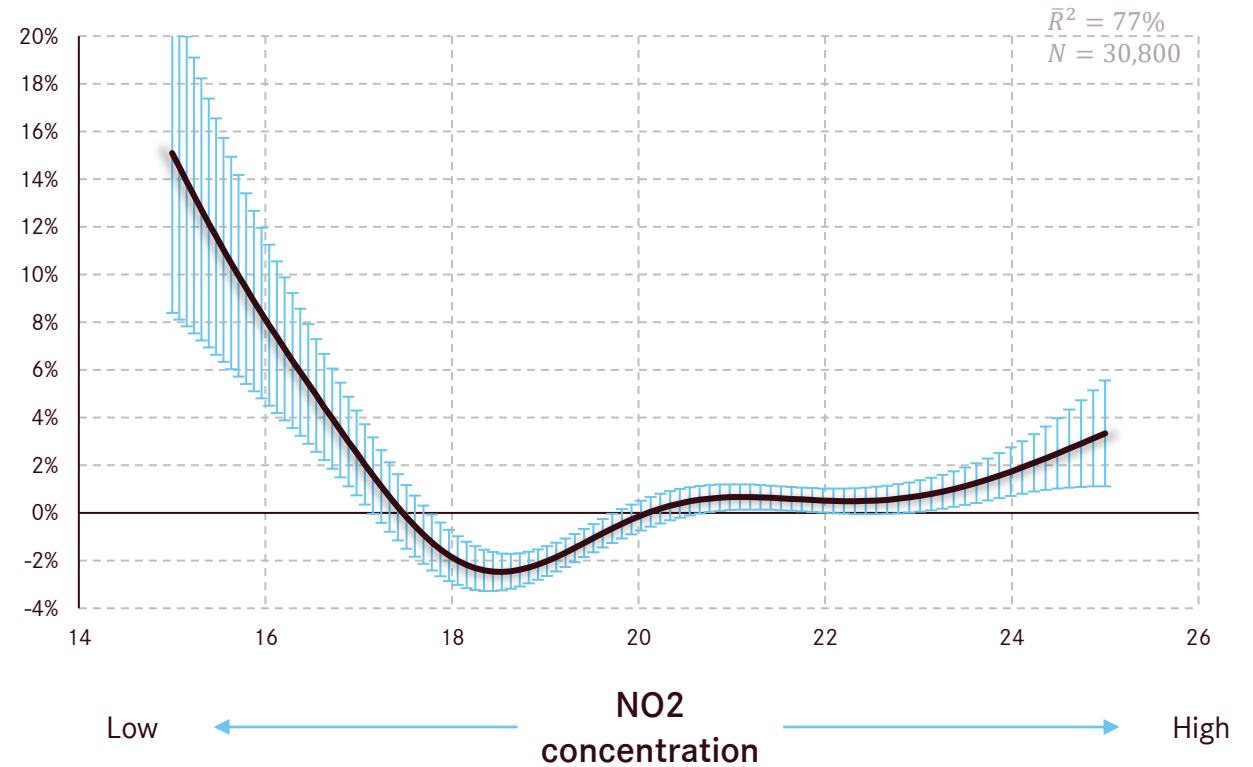
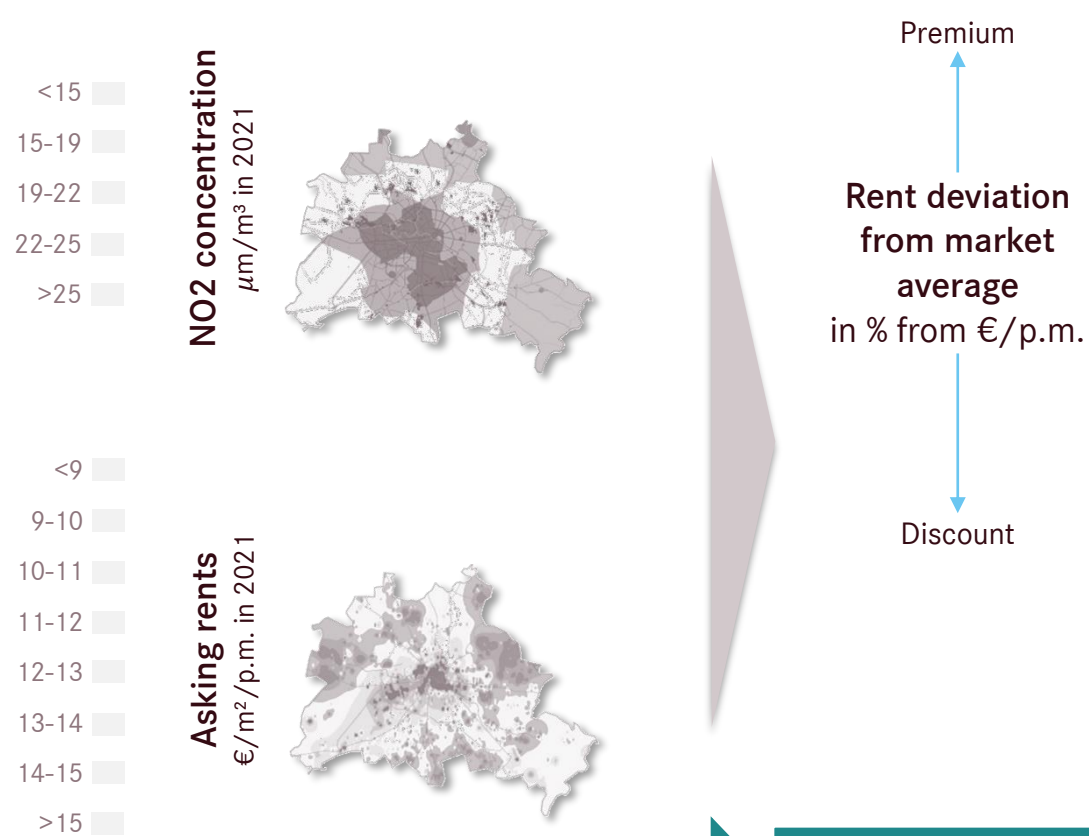
Source: PATRIZIA, Valuemarktdaten

# Nitrogen dioxide and rents 2021

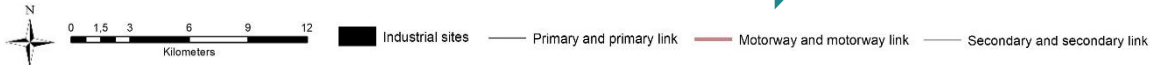


Source: PATRIZIA, Valuemarktdaten

# Nitrogen dioxide and rents 2021



**Market correction:** In 2021, areas with low traffic and combustion from motor vehicles were offered at a premium of up to 14% relative to the market average.



Source: PATRIZIA, Valuemarktdaten

# Key takeaways

1

Short- and long-term exposure to air pollutants poses a serious threat to human health, wellbeing, and the environment

2

Air quality matters when optimizing rental income in a residential portfolio

3

In Berlin, air quality has improved since 2018 pointing to a change in human behaviour and mobility

4

The econometric models explain more than 80% of the variation of rents and identify air quality as an essential influencing factor

5

Tenants are willing to pay up to 5% higher rents for assets located in areas with low pollution levels



**PD Dr. Marcelo Cajias**

Head of Data Intelligence



**M.Sc. Rebecca Restle**

GeoData Scientist

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2023/04/20 PATRIZIA SE

# E-CREDA 40' Research Update

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The price of clean air: Quantifying air pollution exposure in real estate investment decisions

By PD Dr. Marcelo Cajias – PATRIZIA SE / IREBS

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20<sup>th</sup> of April

15:00 – 15:40 CET

Online seminar

E-CREDA improves the access to and understanding of available CRE data to facilitate higher quality applied research and more frequent interaction between investors, data providers and academics to achieve data parity with other major asset classes, like stocks and bonds