Spatially indexed distribution analysis as applied to house prices in the UK

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Background

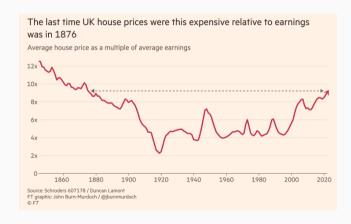
House Prices in the UK

- Ipsos 2022:
 - 69% of Britons believed there was a national housing crisis in the UK.
 - 50% disagree vs 22% agree that there is enough affordable housing locally.
- YouGov August 2025:
 - housing as important an issue as the environment, defence and security, or tax.
 - among 18–24-year-olds it is the second most important issue behind 'the economy'.
- Labour Party Manifesto, 2024:
 - Housing is 'one of the country's biggest barriers to growth'.
 - They aim build 1.5 million homes, including cheaper/social housing, over the parliament.

Existing Work

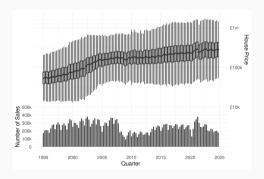
1. Main points

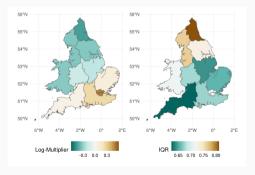
- In 2024, the median average home in England, at £290,000, cost 7.7 times the median average earnings of a full-time employee (£37,600); in Wales the average home (£201,000) was 5.9 times annual earnings (£34,300).
- Affordability in England and Wales in 2024 has returned to its precoronavirus (COVID-19) pandemic levels after a sharp increase between 2020 and 2021 (worsening affordability); median house sales prices have increased by 1% since 2021, while average earnings have increased by 20%.
- In 2024, 9% of local authorities (LAs) (27) had homes bought for less than five times workers' earnings on average and were therefore deemed affordable; this is the highest proportion since 2015, but well below that at the start of the series in 1997 (88% of areas).
- Housing affordability improved in 289 of the 318 LAs in England and Wales (91%) and worsened in 28 (9%) since 2023.
- The most affordable LAs in 2024 were Blaenau Gwent (with a ratio of 3.8), Burnley (3.9) and Blackpool (3.9); the least affordable was Kensington and Chelsea (27.1, which was about seven times less affordable).
- Between 2019 and 2024, 4 of the 10 largest increases in affordability ratios (worsening affordability) were in the East Midlands, while the 10 largest decreases in affordability ratios have all occurred in LAs in London.



House Price Data

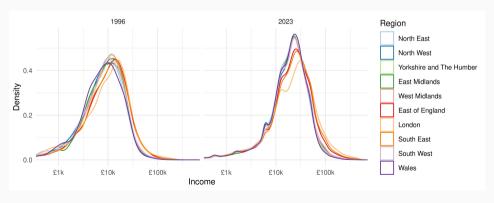
- HM Land Registry Price Paid Data Set.
- 1 Jan 1995—: over 30 million single residential property sales for value, referenced by address and date, and marked by price recorded.





Income Data

- HM Revenue and Customs' Survey of Personal Incomes.
- 6 Apr 1995—: a weighted sample of those who may be liable for income tax, referenced by region and tax year, and marked by total income.



Overview

• Prices paid:

$$p_i \sim P(s_i, t_i), \quad i \in [n].$$
 (1)

• Incomes earnt:

$$y_i \sim Y(r_i, \tau_i), \quad i \in [m],$$
 (2)

with weightings w_i .

• Aim is to estimate these distribution fields P and Y, then to interpret them through a regression of the quantile functions:

$$P^{-1}(s,t;q) = \theta(s,t;q)Y^{-1}(s,t;q).$$
(3)

• The parameter θ can be interpreted for each $q \in [0,1]$, or by taking functionals.

Methodology

House Price Distributions

- We estimate P on a spatial grid $s \in S$ and for each tax year $\tau \in T$.
- We apply a Nadaraya–Watson type smoother to the observations:

$$\hat{P}(s,\tau;p) : \propto \sum_{i \in [n]: t_i \in \tau} K_h(d(s,s_i)) \mathbb{I}(p_i \le p), \tag{4}$$

where K_h is a Gaussian kernel of bandwidth h.

• The bandwidth h is chosen locally to ensure the effective sample size is fixed:

$$\sum_{i \in [n]: t_i \in \tau} \left(K_h(d(s_i, s)) \right)^2 / \left(\sum_{i \in [n]: t_i \in \tau} K_h(d(s_i, s)) \right)^2 := \frac{1}{n^*} = \frac{1}{100}.$$
 (5)

Income Distributions

- We estimate Y on the same grid $(s, \tau) \in S \times T$.
- \bullet For each region r we estimate a regional income distribution:

$$\hat{Y}(r,\tau;y) : \propto \sum_{i \in I(r,\tau)} w_i \mathbb{I}(y_i \le y), \tag{6}$$

where $I(r, \tau) := \{i \in [m] : \tau_i = \tau, r_i = r\}.$

• We suppose there is a population density ρ , and then smooth by a Nadaraya–Watson type smoother to these estimates:

$$\hat{Y}(s,\tau;y) : \propto \sum_{r} \left(\int_{r} K_{h}(d(s,\tilde{s})) \rho(\tilde{s}) d\tilde{s} \right) \hat{Y}(r,\tau;y). \tag{7}$$

• The bandwidth is chosen again to ensure a fixed effective sample size:

$$\sum_{r} \left(\int_{r} \left(K_{h}(d(s,\tilde{s})) \right)^{2} \left(\rho(\tilde{s}) \right)^{2} d\tilde{s} \right) \left(\sum_{i \in I(r,\tau)} w_{i}^{2} \middle/ \left(\sum_{i \in I(r,\tau)} w_{i} \right)^{2} \right) \left/ \left(\sum_{r} \int_{r} K_{h}(d(s,\tilde{s})) \rho(\tilde{s}) ds \right)^{2} := \frac{1}{m^{*}}$$

$$(8)$$

Regression Fitting

- Smoothing has occured in the estimation of distributions, so all that remains is a regression analysis of the distributions in time and space.
- Spacetime-varying concurrent regression of house price quantiles on income quantiles can be rearranged as a functional additive model:

$$\log \frac{P^{-1}(s,t;q)}{Y^{-1}(s,t;q)} = \alpha(q) + \beta(s;q) + \gamma(t;q) + \delta(s,t;q).$$
 (9)

• This can be fitted by the backfitting algorithm to obtain functional parameter estimates $\hat{\alpha}$, $\hat{\beta}$, $\hat{\gamma}$ and $\hat{\delta}$.

Regression Analysis

- One can examine the parameters directly as functions of $q \in [0, 1]$ directly, but finite-dimensional summaries of them can also be useful for interpretation.
- Here, a Legendre polynomial expansion is utilised:

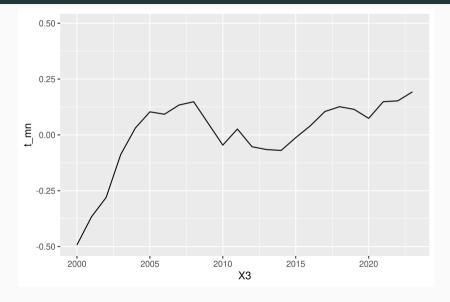
$$\hat{\alpha}(q) = \sum_{k=1}^{\infty} \hat{\alpha}_k \phi_k(q), \quad \text{where} \quad \hat{\alpha}_k = \langle \hat{\alpha}, \phi_k \rangle, \quad \text{etc.},$$
 (10)

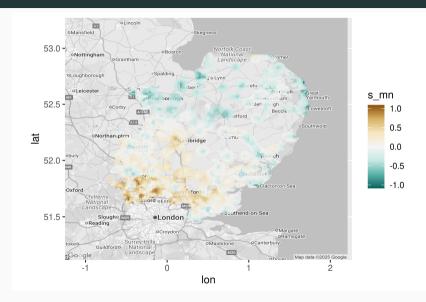
where ϕ_k is the k-th order Legendre polynomial transformed to [0, 1] and normalised.

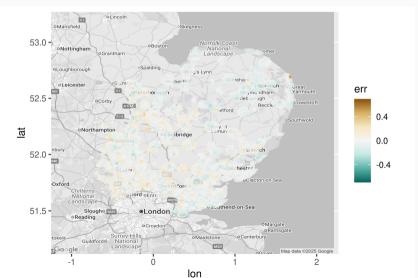
- The first three components are readily interpretable:
 - 1. the relative expense of housing across the income spectrum;
 - 2. how uniform house prices are compared to the income distribution;
 - 3. a skew towards either higher or lower priced homes.

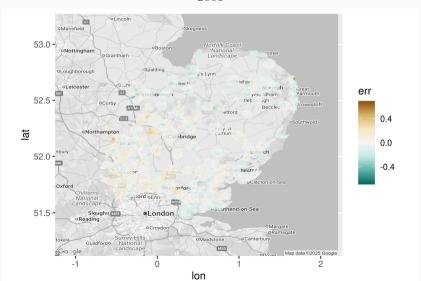
Application

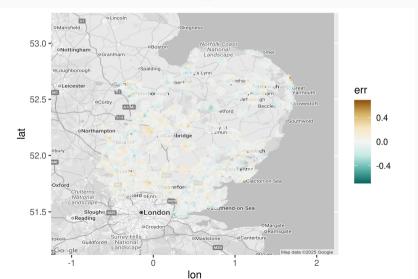
East of England: $\hat{\gamma}_1$



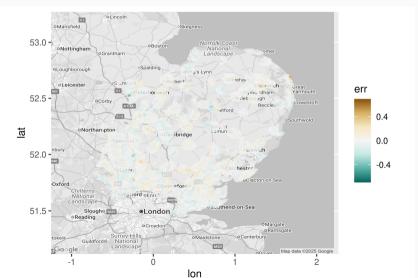




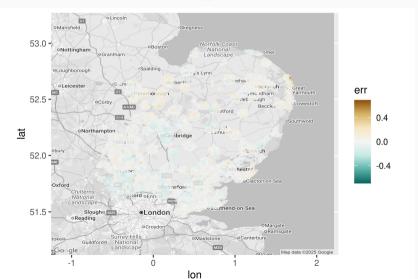




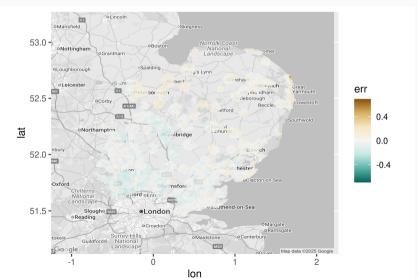




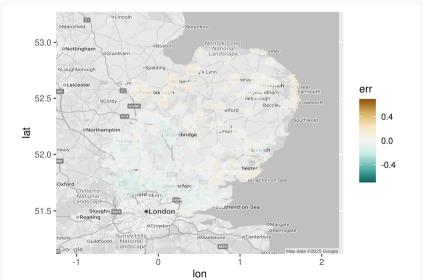




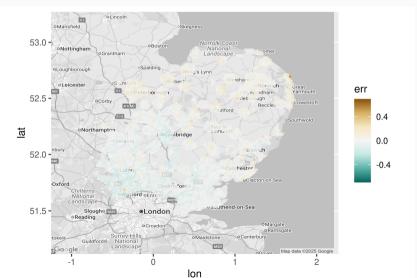




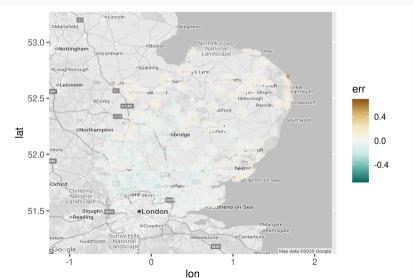


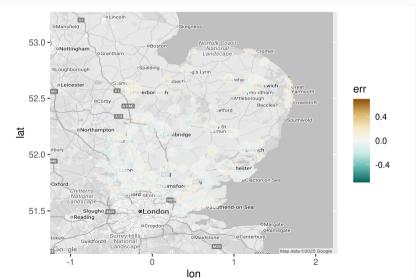


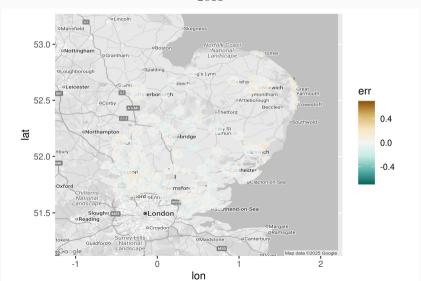




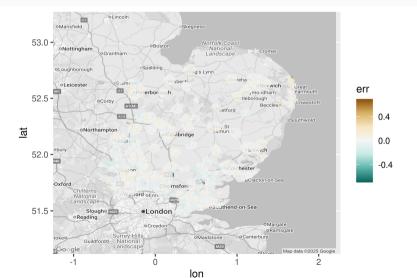




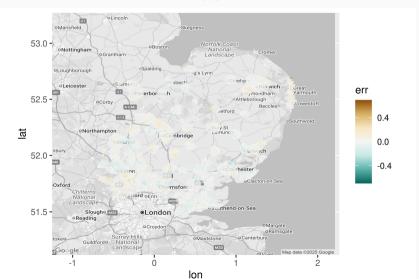




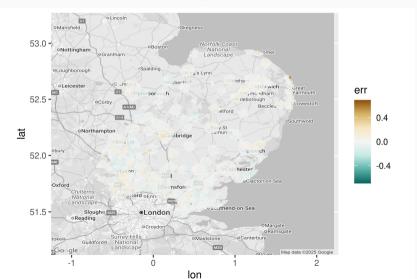




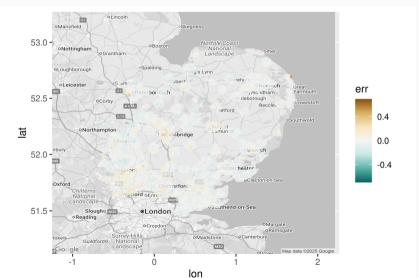




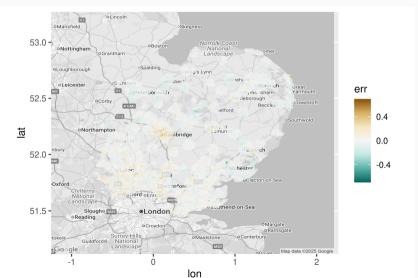




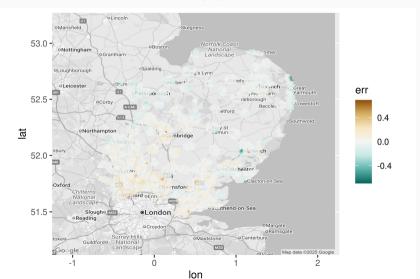




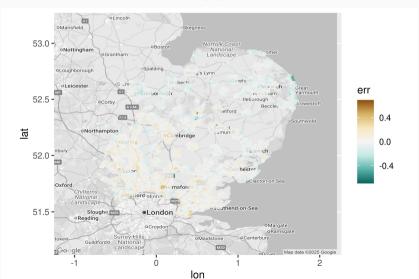






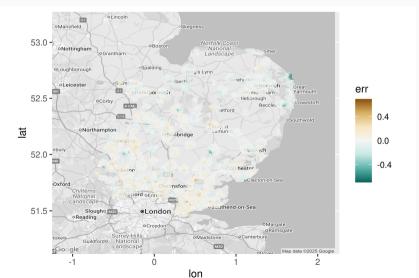


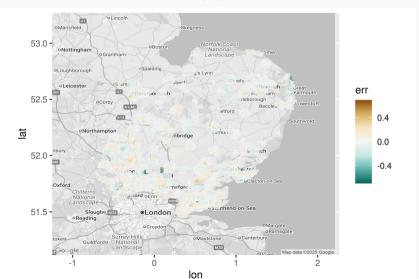


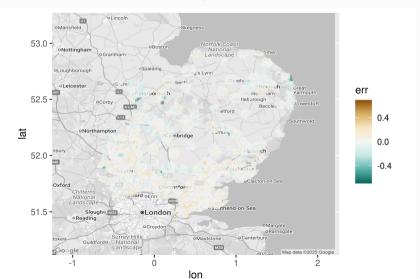


East of England: $\hat{\delta}_1^{\perp}$



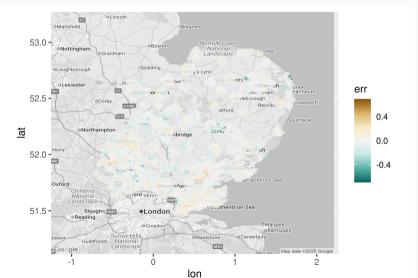




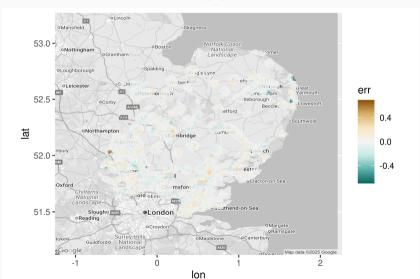


East of England: $\hat{\delta}_1$

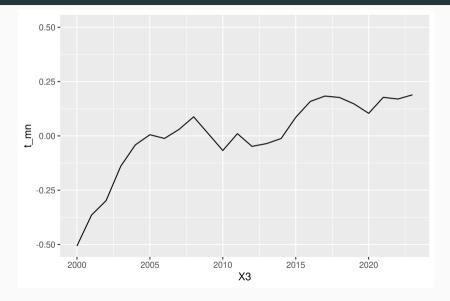


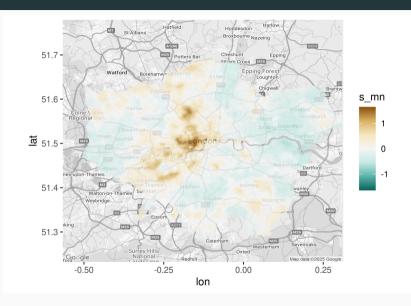


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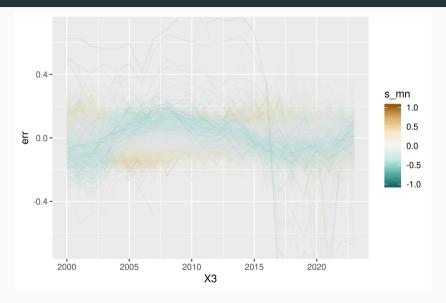


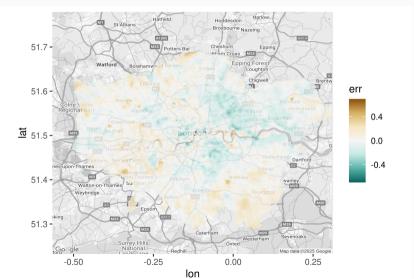
London: $\hat{\gamma}_1$

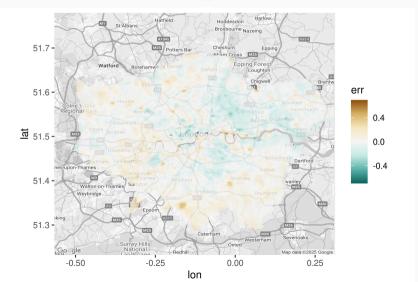




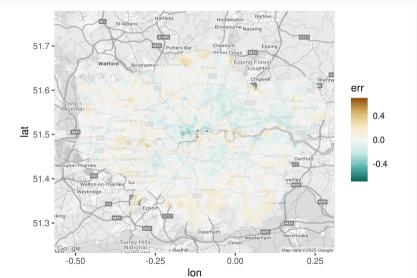
East of England: Gentrification

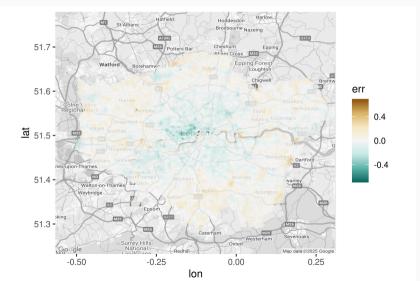




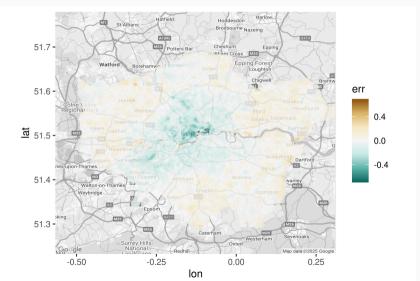


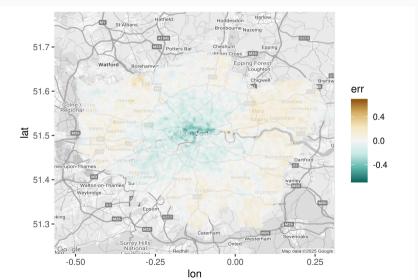




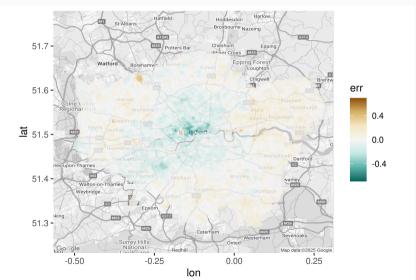




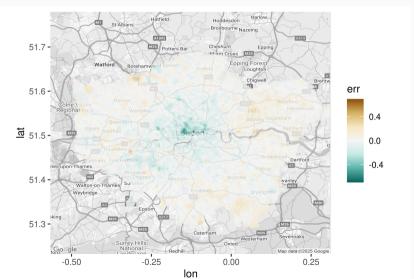




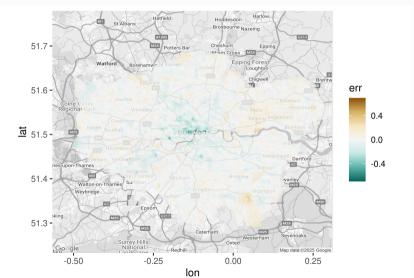




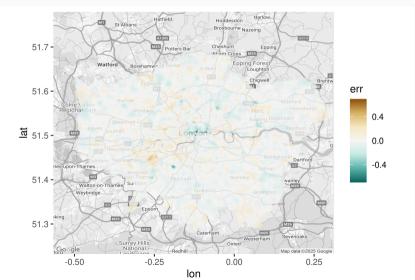




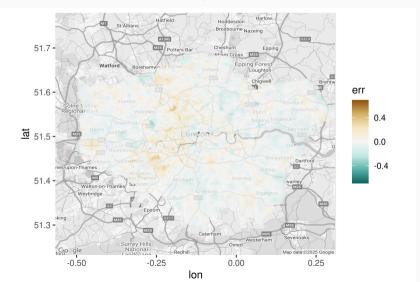




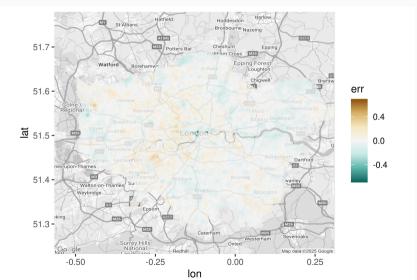
$\textbf{London: } \hat{\delta}_1$



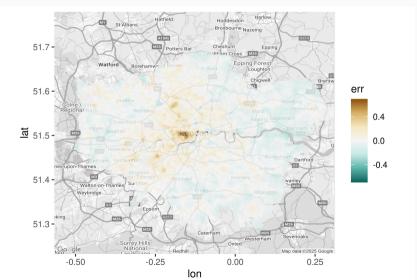




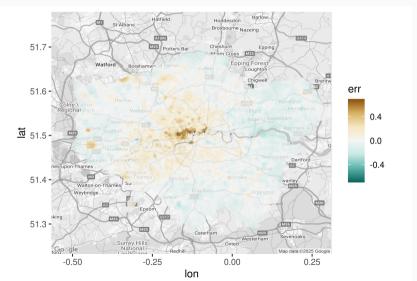




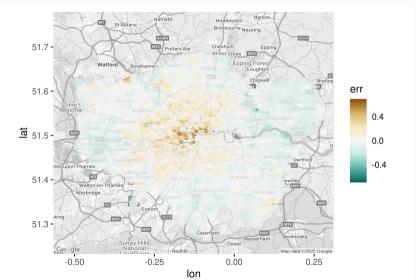




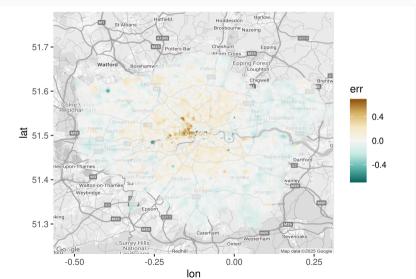




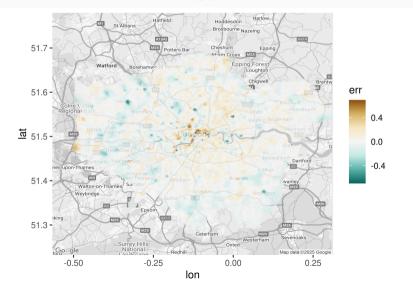




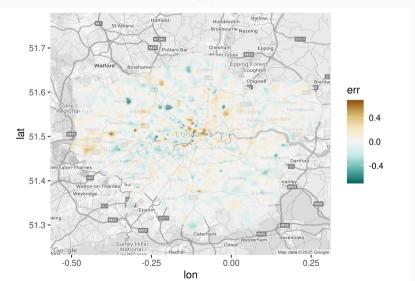




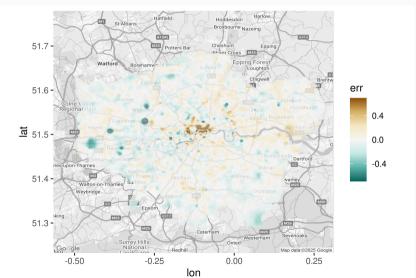


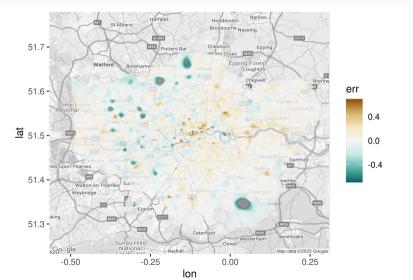




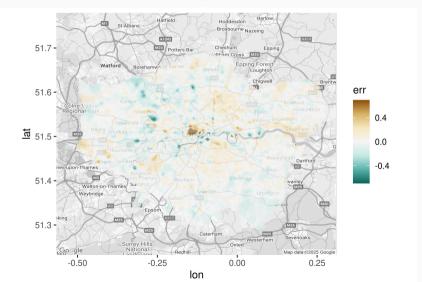


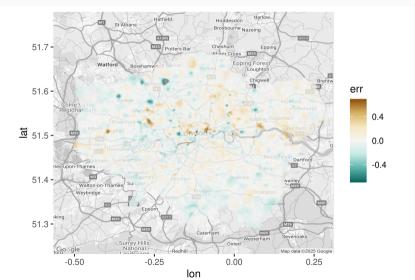


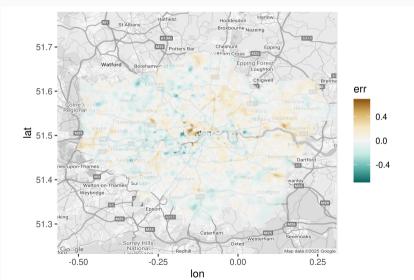




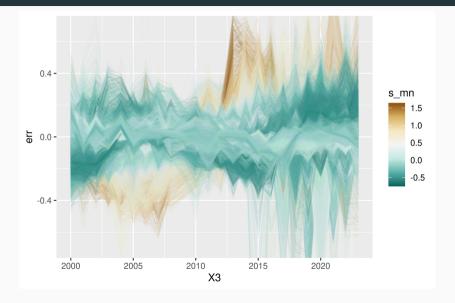




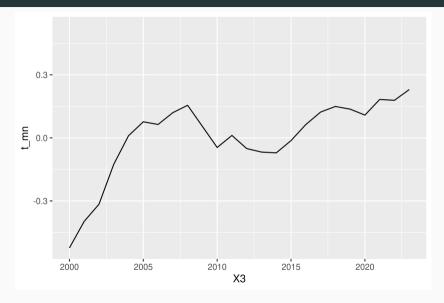




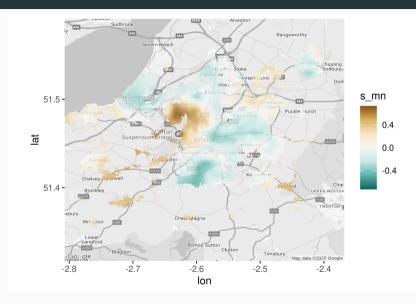
London: Gentrification



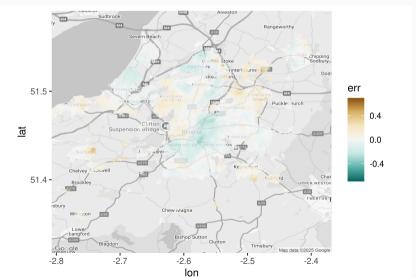
Bristol and Surroundings: $\hat{\gamma}_1$



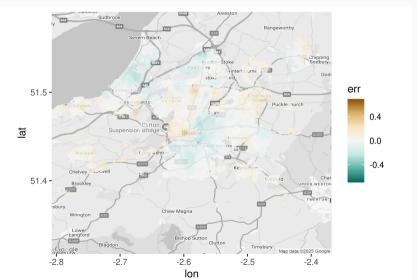
Bristol and Surroundings: \hat{eta}_1^{-1}



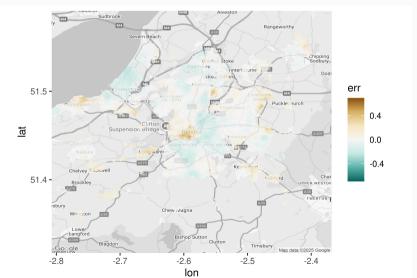




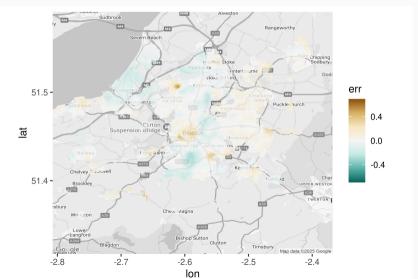




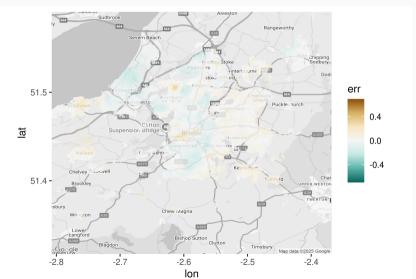




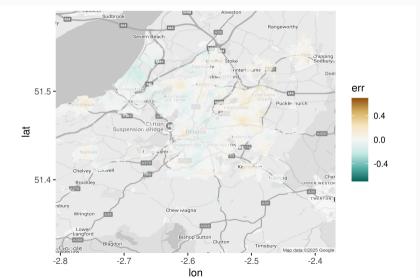




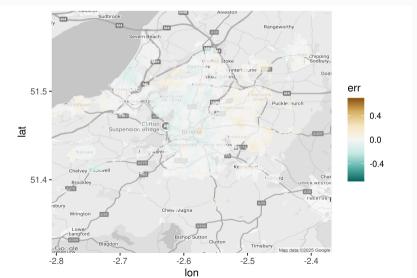




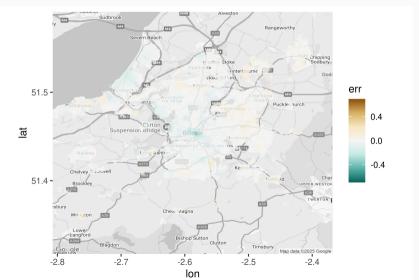




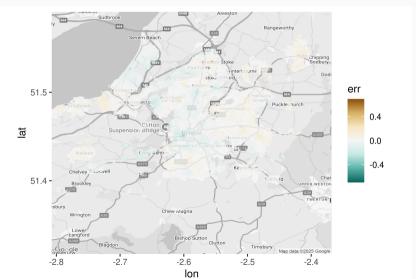




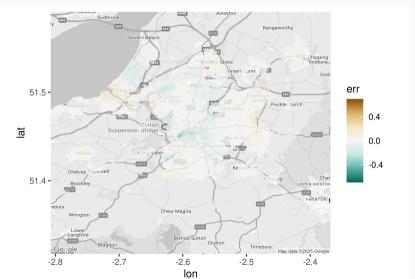




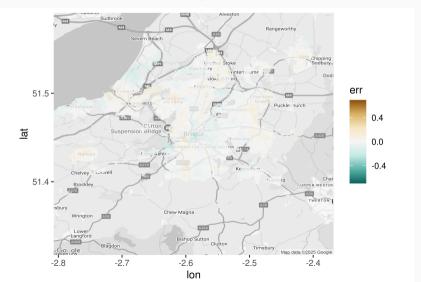




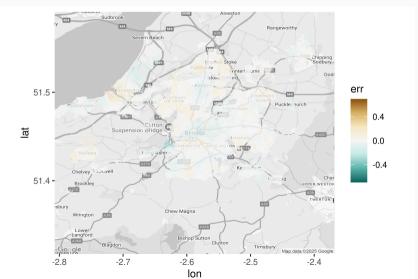




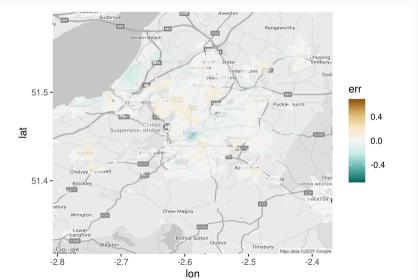




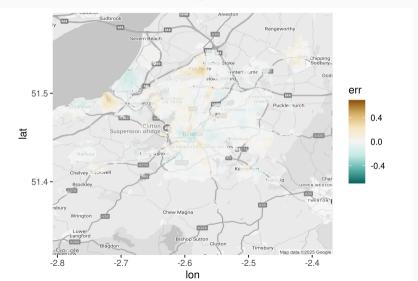




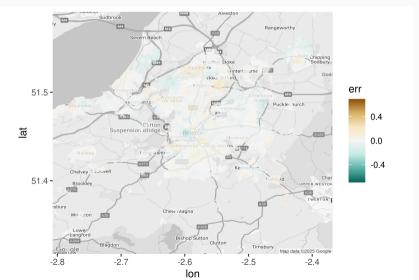




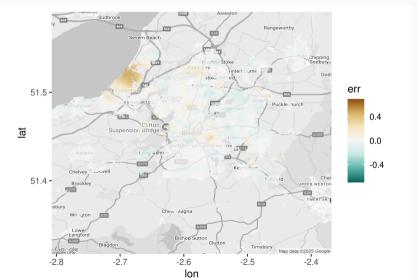




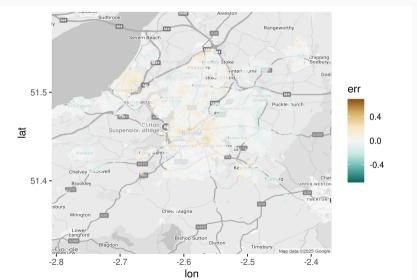




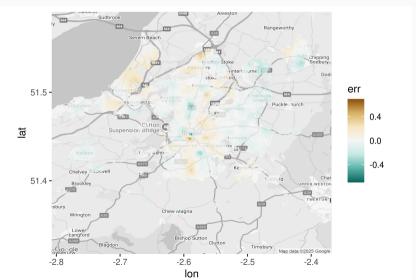




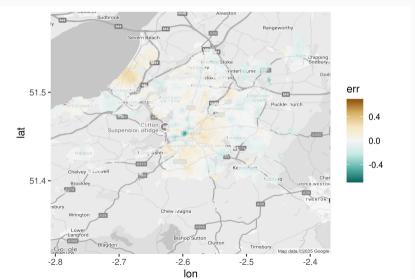




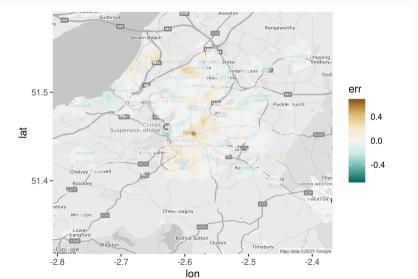




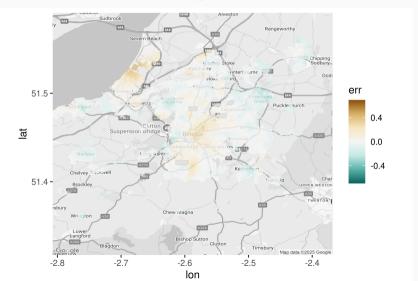




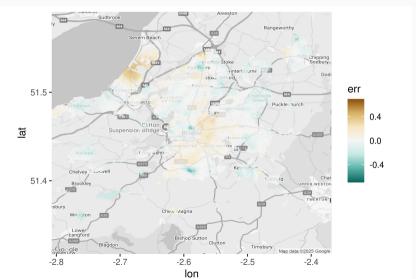




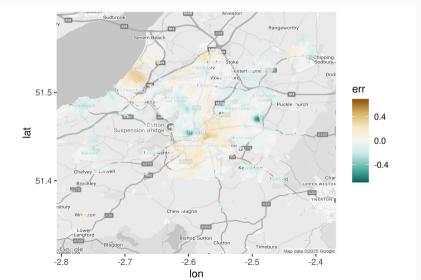




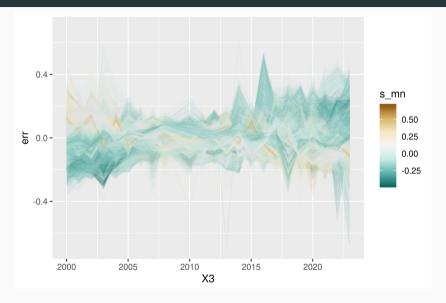


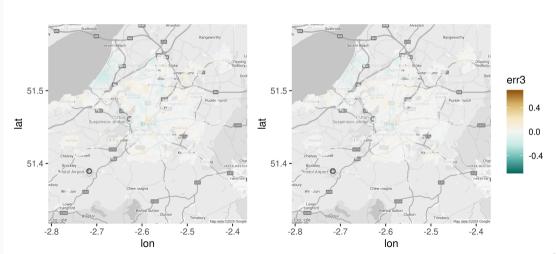


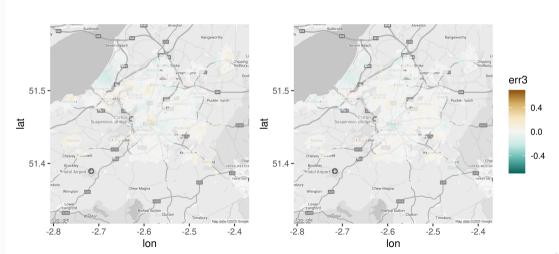


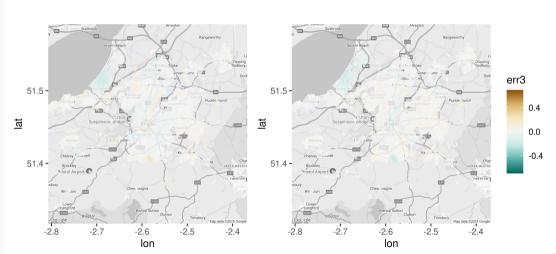


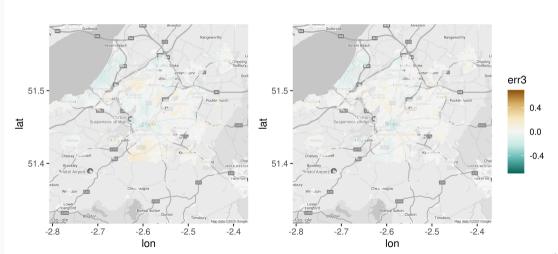
Bristol and Surroundings: Gentrification

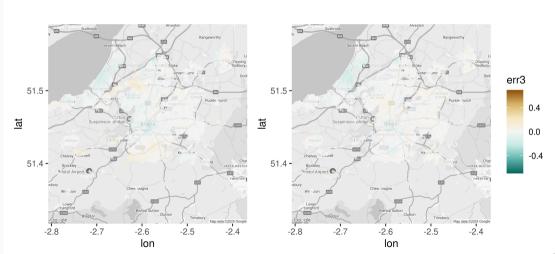


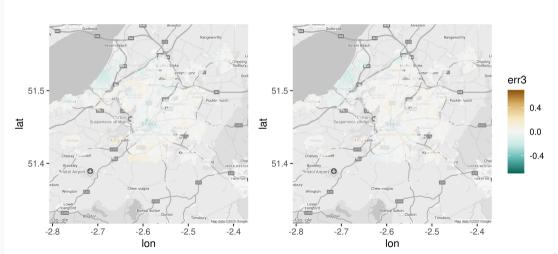


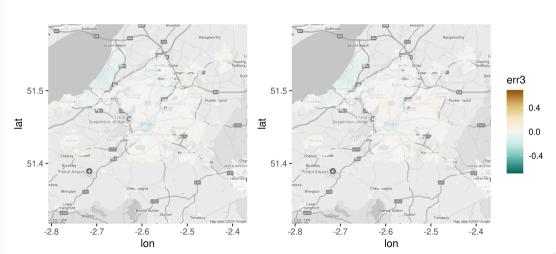


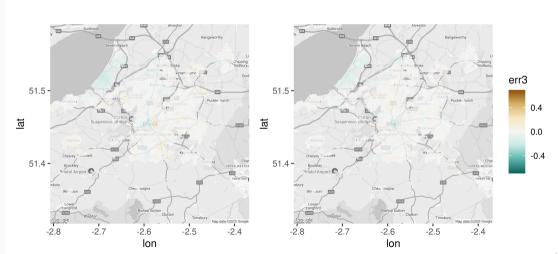


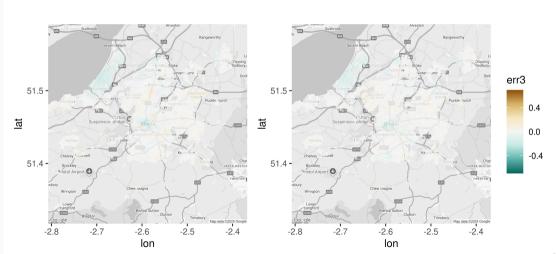


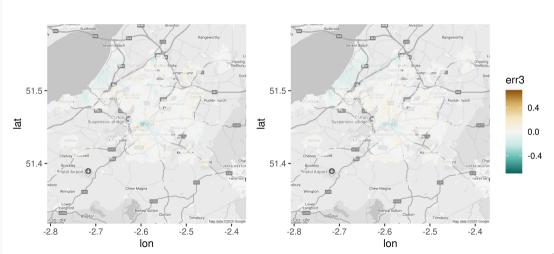


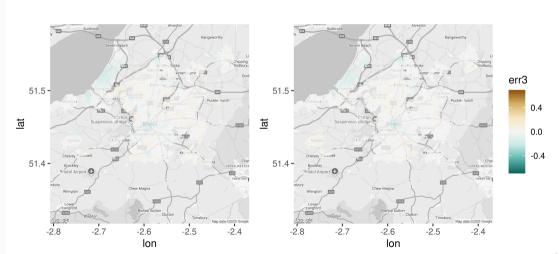


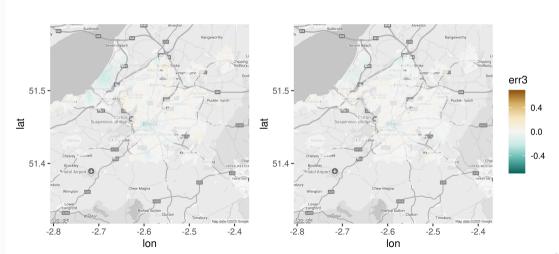


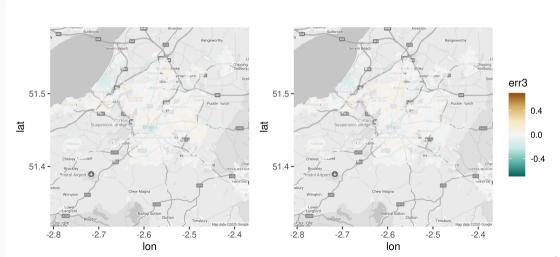


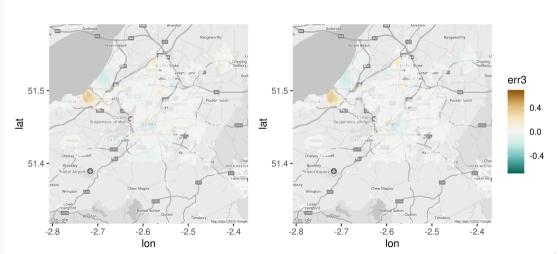


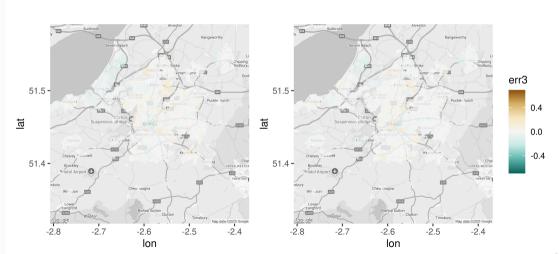


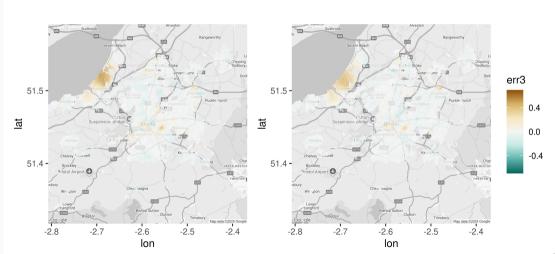


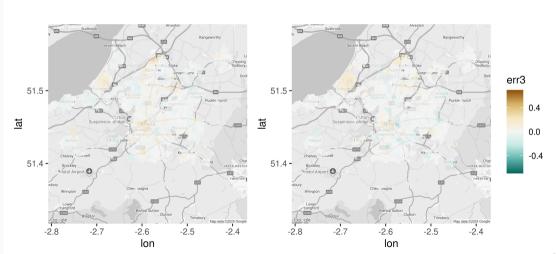


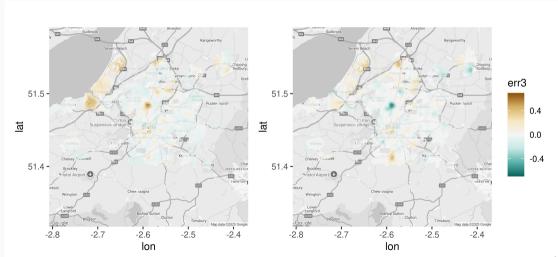


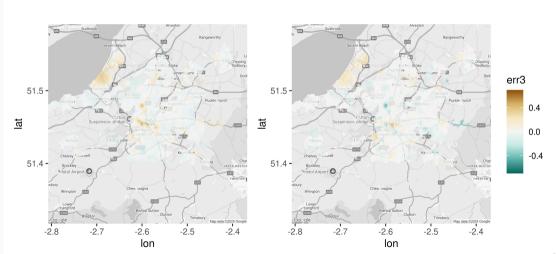


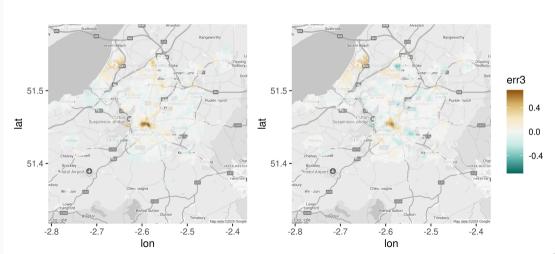


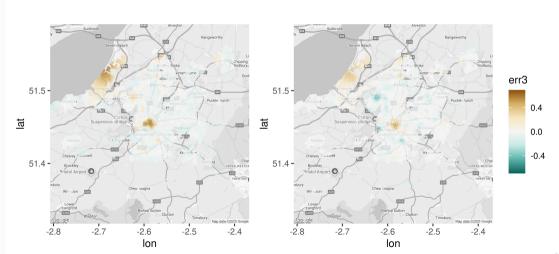


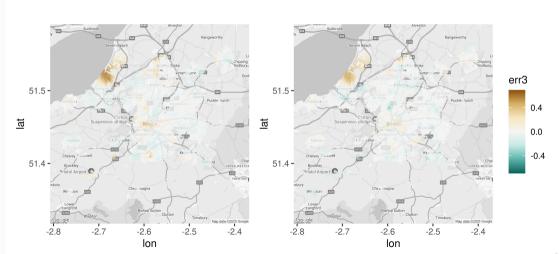


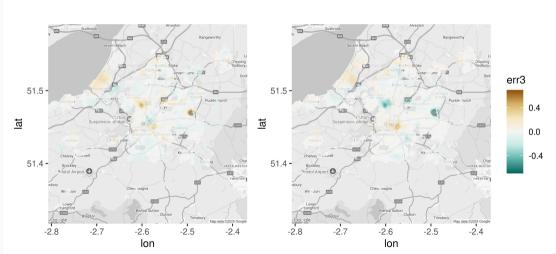












Discussion

Open Questions

- Has the smoothing been chosen/implemented appropriately?
- What's the best way to incorporate uncertainty quantification into the methodology?
- How can we implement an inferential framework to identify regions of interesting change over time?
- How can you factor in rental prices?
- Can this output be used to measure impact of public policy?

Other Work

- 'Exploring Covid-19 Spatiotemporal Dynamics: Non-Euclidean Spatially Aware Functional Registration.' Luke A. Barratt & John A. D. Aston. July 2025. Annals of Applied Statistics.
- 'A Nonparametric and Functional Wombling Methodology.' Luke A. Barratt & John A. D. Aston. February 2025. arXiv:2502.07740.
- 'Spatially Indexed Distribution Analysis as Applied to House Prices in the UK.' Luke A. Barratt & John A. D. Aston. August 2025. arXiv: coming soon.
- 'School Inspections and House Prices: Distribution-on-Functio Regression for Spatial Data.' Luke A. Barratt & John A. D. Aston. Working paper—email lab85@cam.ac.uk for details.