

Real Estate Price Modeling with GAMs

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Overview

This report presents a Generalized Additive Model (GAM) for real estate price prediction. The model is built on a clean, reproducible pipeline combining dbt-based feature engineering and a Python GAM implementation.

Experiment Configuration

This report corresponds to the baseline experiment:

- Experiment: `baseline_gam.yaml`
 - Model: LinearGAM
 - Target: Price per m²
-

Data Loading

LinearGAM

```
=====
Distribution:           NormalDist Effective DoF:
Link Function:         IdentityLink Log Likelihood:
Number of Samples:     20502 AIC:
                        AICc:
                        GCV:
                        Scale:
                        Pseudo R-Squared:
=====
```

Feature Function	Lambda	Rank	EDoF	P > x
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```

=====
f(0)           [0.6]      131      130.1    1.11e-16
f(1)           [0.6]      3        2.0      1.11e-16
f(2)           [0.6]      3        2.0      1.11e-16
f(3)           [0.6]      2        1.0      1.11e-16
f(4)           [0.6]      9        8.0      1.11e-16
intercept          1        0.0      1.11e-16
=====
Significance codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

WARNING: Fitting splines and a linear function to a feature introduces a model identifiability
         which can cause p-values to appear significant when they are not.

WARNING: p-values calculated in this manner behave correctly for un-penalized models or models
         with known smoothing parameters, but when smoothing parameters have been estimated, the
         p-values are typically lower than they should be, meaning that the tests reject the null too
         often.
None
{'r2': OrderedDict([('explained_deviance', np.float64(0.8677604585353504)), ('McFadden', np.

/home/nuntiusten/dev/real-estate-gam-db/real_estate_gam/run_gam.py:53: UserWarning: KNOWN P

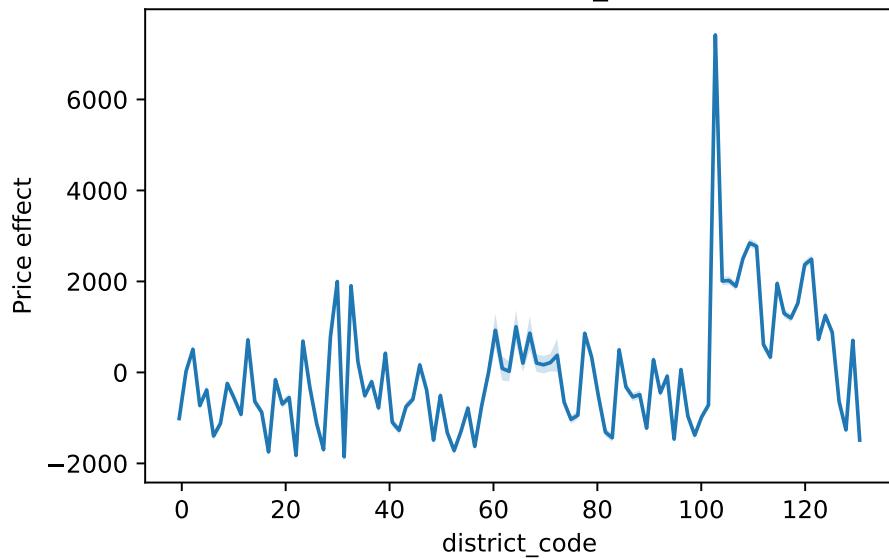
Please do not make inferences based on these values!

Collaborate on a solution, and stay up to date at:
github.com/dswah/pyGAM/issues/163

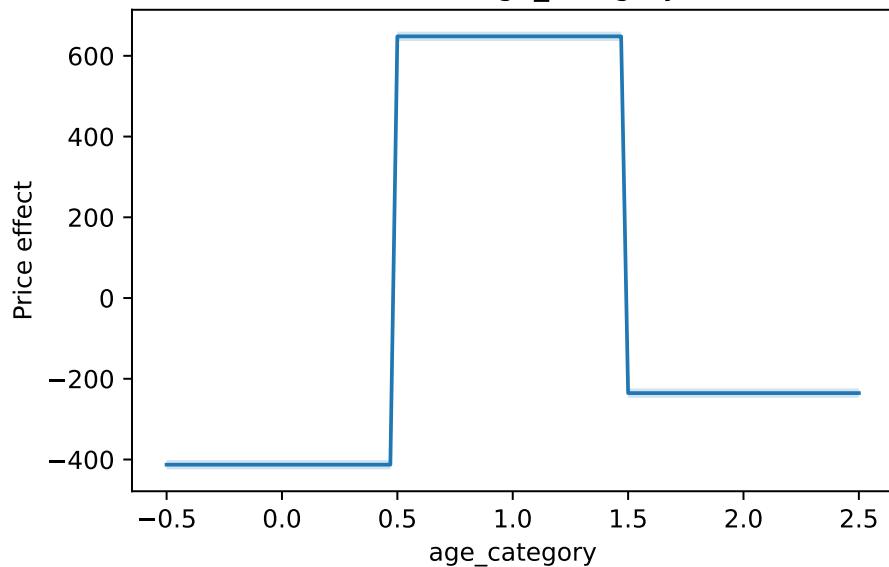
print(gam.summary())

```

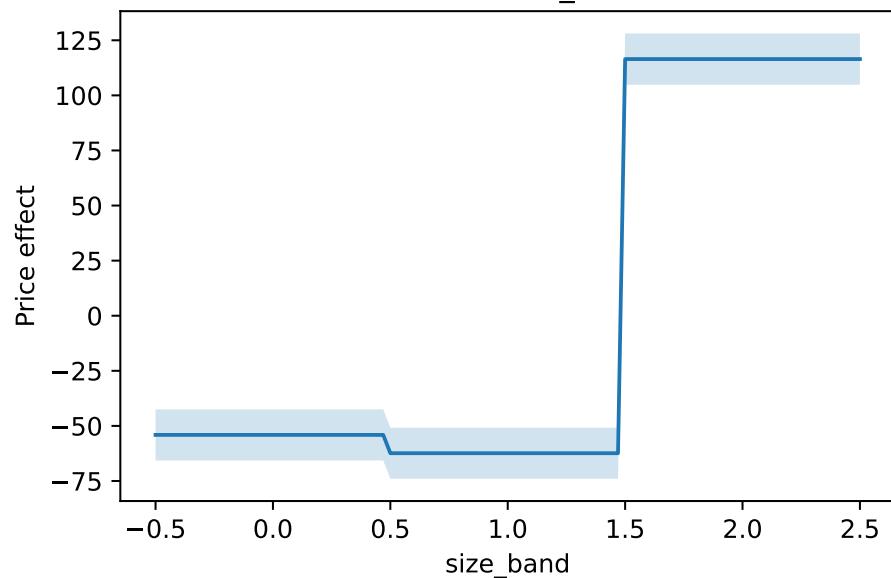
Effect of district_code



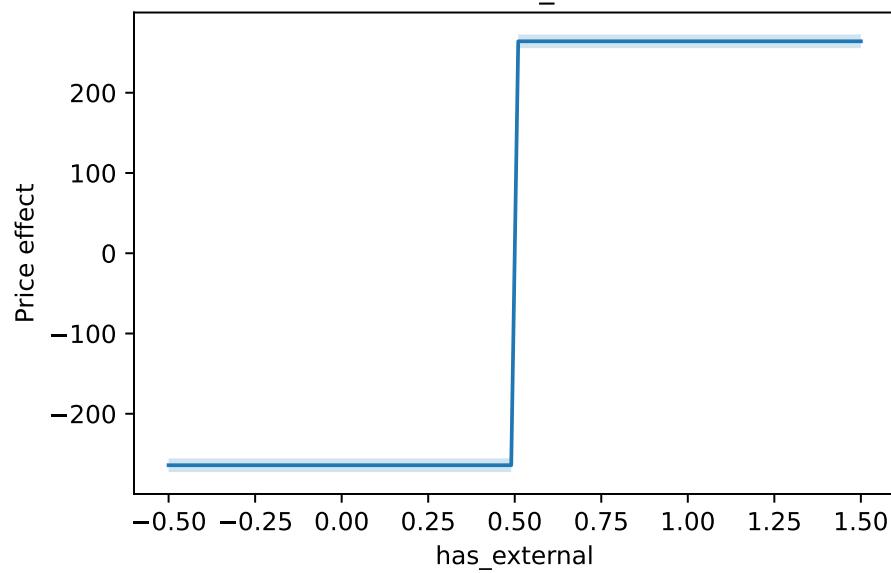
Effect of age_category



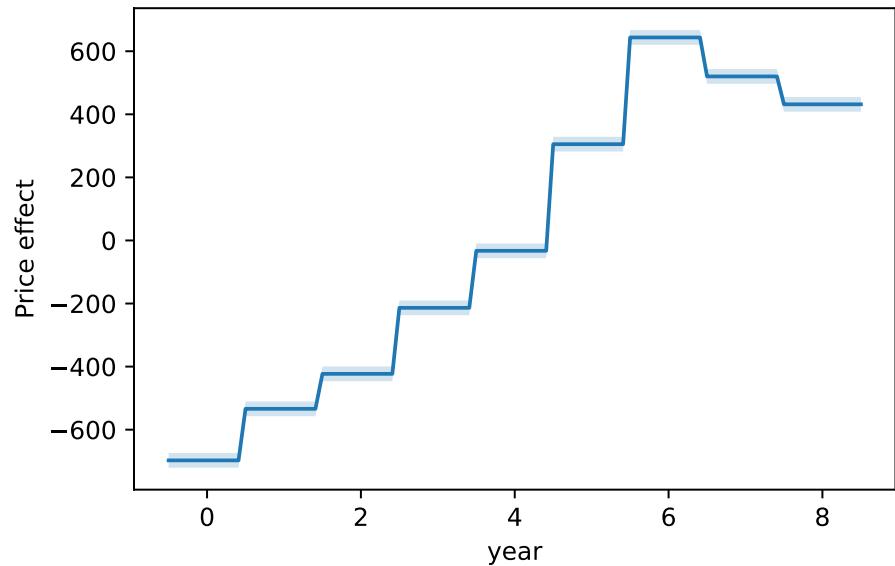
Effect of size_band



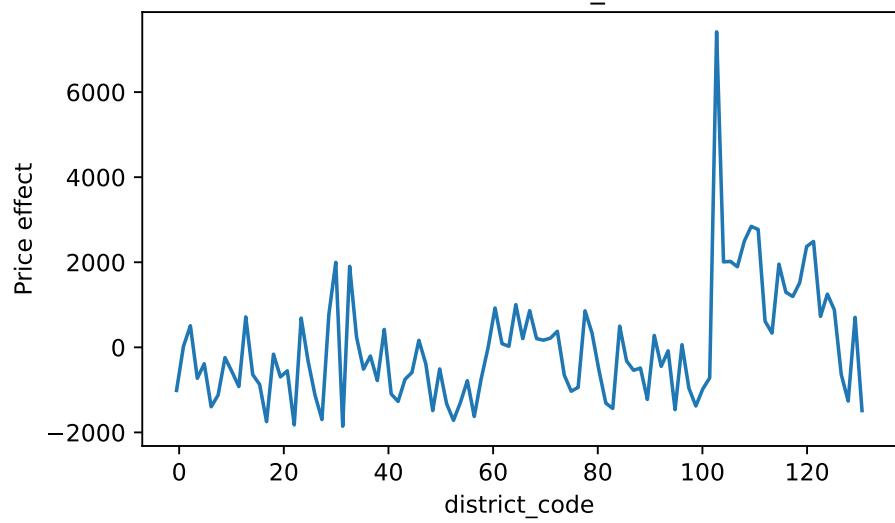
Effect of has_external

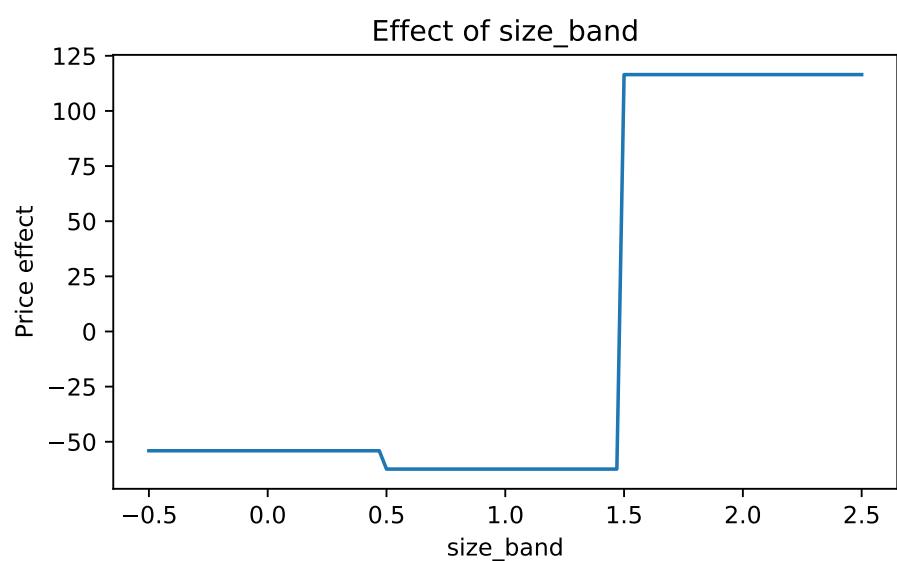
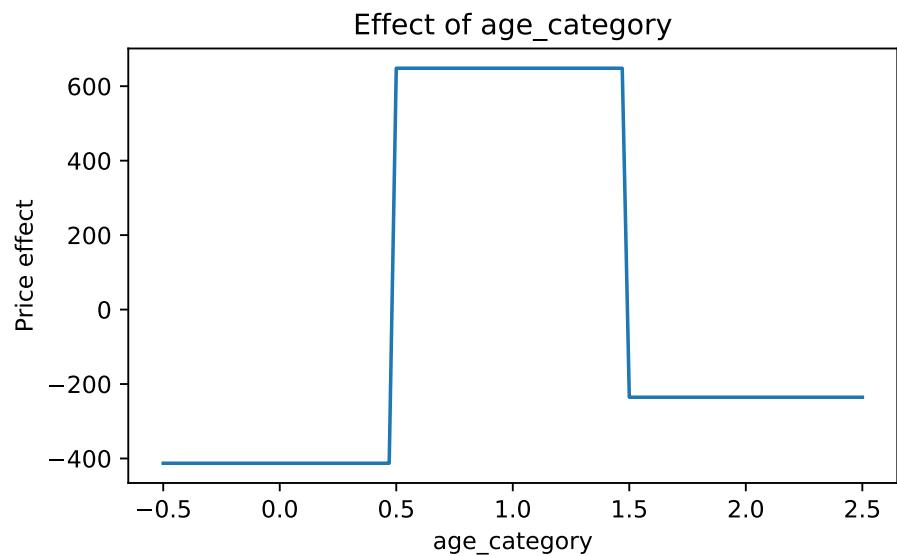


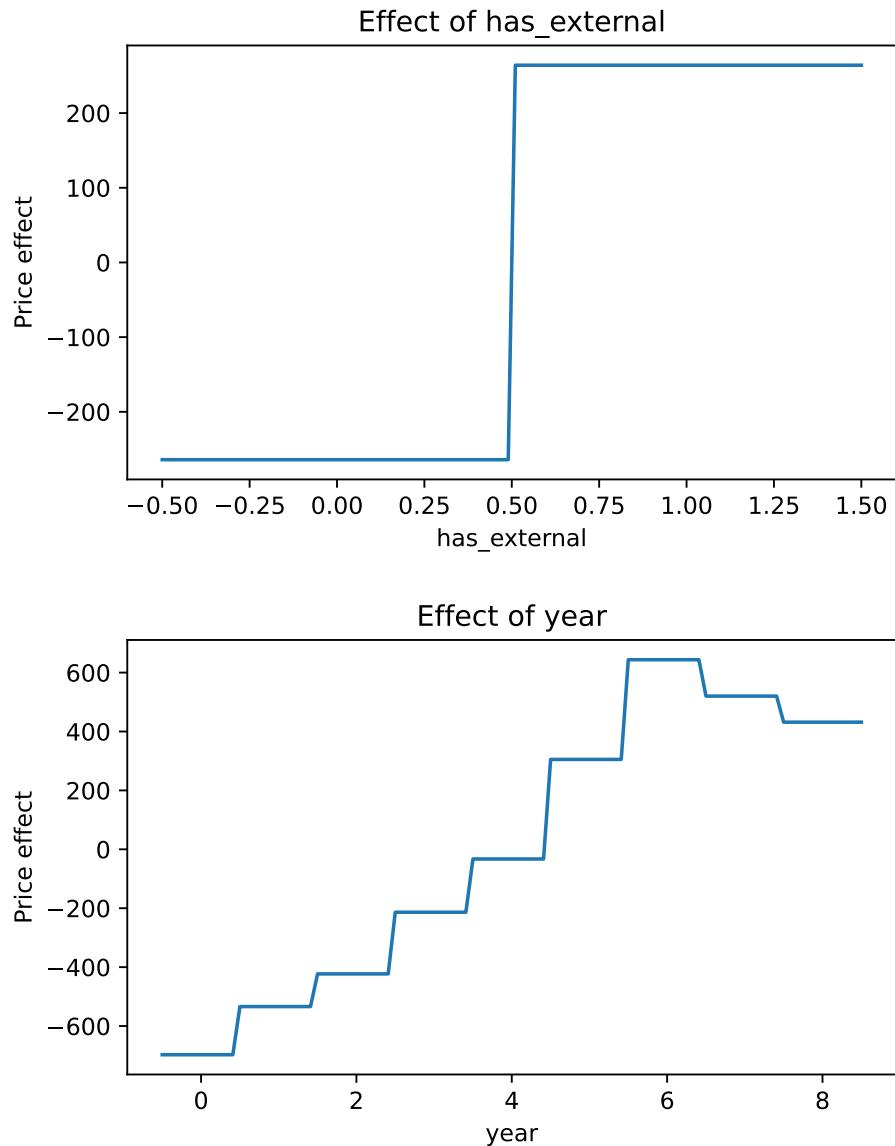
Effect of year



Effect of district_code







Key Takeaways

- The GAM captures strong non-linear effects.
- Categorical variables show clear price differentials.
- The model achieves solid explanatory power while remaining interpretable.

Reproducibility

This report was generated from a fully reproducible pipeline. To reproduce:
`python3 -m src.run_gam quarto render reports/gam_report.qmd`

Source code and configuration are available in the public repository.