

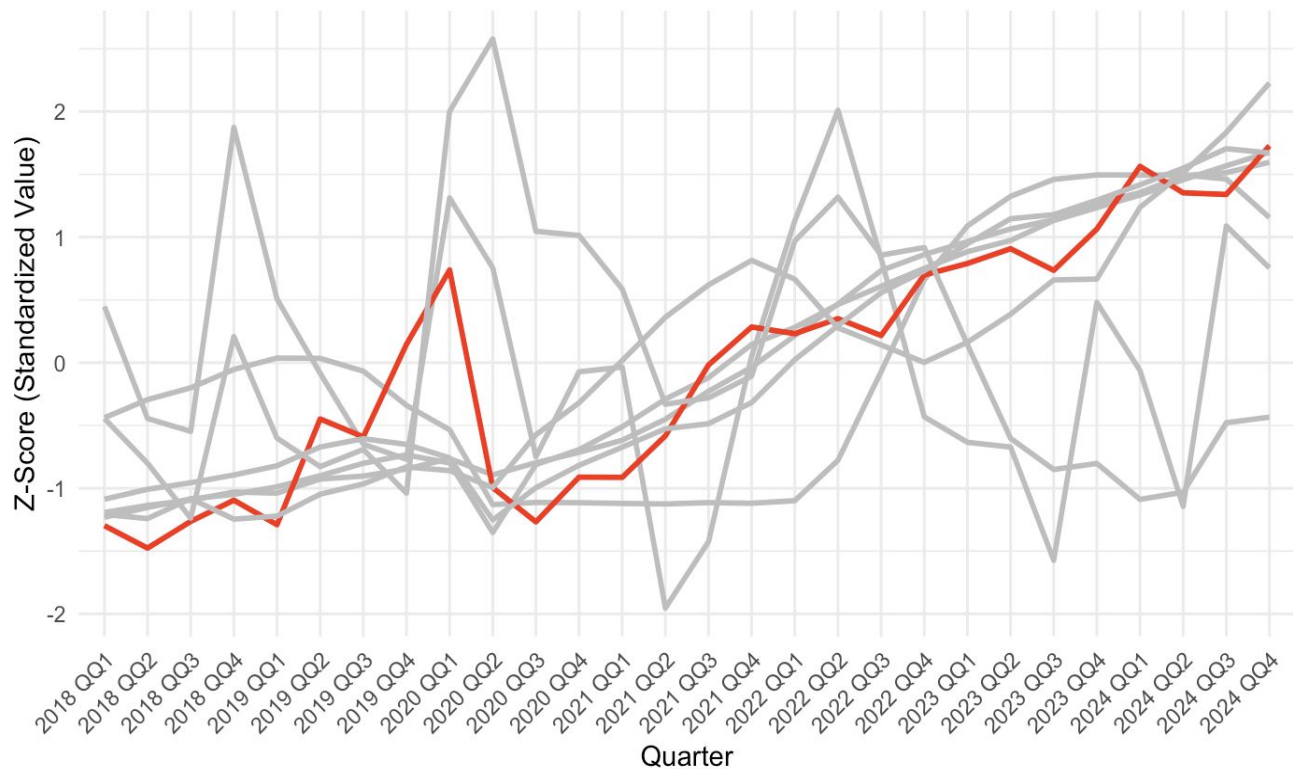
Rent, Risk, and Recovery: Strategic Leasing Insights from Macroeconomic Models

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Predicting National Average Rent with Macro Factors

Standardized Macro Trends with Highlight on Rent



tech quality activity policies investment availability
tenant corporate work index consumer capital
hybrid formation growth market infrastructure bond industry
local interest trends office costs risk unemployment
rates earnings business remote level lease tax exchange
rate space venture building price yields
prices stock population policy producer prevalence

Avg National Rent
(standardized)

Macro Data (GDP,
S&P 500, CPI, Federal
Funds, VIX, PNFI)

Panel Regression for City-Specific Sensitivities - Fixed-Effects Model (Common Slopes)

$$price_{i,t} = \alpha_i + \beta_1 GDP_pct_change_t + \beta_2 SP_pctchange_t + \beta_3 fed_rate_t + \beta_4 vix_pctchange_t + \beta_5 CPI_pct_change_t + \beta_6 PNFI_pct_change_t + \varepsilon_{i,t}$$

OLS Regression Results

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Dep. Variable:      price      R-squared:      0.973
Model:              OLS        Adj. R-squared:  0.971
Method:              Least Squares      F-statistic:    607.7
Date:                Sun, 06 Apr 2025    Prob (F-statistic): 2.59e-189
Time:                06:48:25          Log-Likelihood:  -670.03
No. Observations:    270              AIC:            1372.
Df Residuals:        254              BIC:            1430.
Df Model:            15
Covariance Type:     nonrobust
    
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	coef	std err	t	P> t	[0.025	0.975]
C(city)[Atlanta]	26.5549	0.801	33.161	0.000	24.978	28.132
C(city)[Austin]	39.0612	0.801	48.778	0.000	37.484	40.638
C(city)[Boston]	37.2109	0.801	46.468	0.000	35.634	38.788
C(city)[Manhattan]	74.5817	0.801	93.135	0.000	73.005	76.159
C(city)[Northern New Jersey]	26.6540	0.801	33.285	0.000	25.077	28.231
C(city)[Northern Virginia]	31.5887	0.801	39.447	0.000	30.012	33.166
C(city)[Philadelphia]	26.6458	0.801	33.274	0.000	25.069	28.223
C(city)[Salt Lake City]	22.3190	0.801	27.871	0.000	20.742	23.896
C(city)[San Francisco]	69.4495	0.801	86.726	0.000	67.873	71.027
C(city)[Seattle]	38.0139	0.801	47.471	0.000	36.437	39.591
GDP_pct_change	-0.0278	0.133	-0.209	0.834	-0.289	0.233
SP_pctchange	0.0920	0.042	2.205	0.028	0.010	0.174
fed_rate	0.6175	0.098	6.280	0.000	0.424	0.811
vix_pctchange	0.0084	0.005	1.654	0.099	-0.002	0.018
CPI_pct_change	1.9914	0.432	4.608	0.000	1.140	2.842
PNFI_pct_change	-0.2668	0.184	-1.451	0.148	-0.629	0.095
Omnibus:	11.882	Durbin-Watson:	0.366			
Prob(Omnibus):	0.003	Jarque-Bera (JB):	23.420			
Skew:	0.163	Prob(JB):	8.21e-06			
Kurtosis:	4.405	Cond. No.	749.			

- City fixed effects α capture large rent level differences (e.g., Manhattan \approx \$75/SF vs. Phoenix \approx \$25/SF)
- Common slope estimates:
 - Positive** for GDP and S&P 500 (though not all are statistically significant)
 - Negative** for VIX
 - Slightly negative** for the federal funds rate (suggesting higher rates may suppress rents)
- The model's R^2 is **relevant** \rightarrow macro variables explain some rent variation
- City-specific dynamics are crucial**

Bayesian Hierarchical Model (Partial Pooling)

Let θ represent all unknown parameters:

$$\theta = \{\alpha_j, \beta_j^{GDP}, \beta_j^{SP}, \mu_\alpha, \mu_{\beta, GDP}, \mu_{\beta, SP}, \tau_\alpha, \tau_{\beta, GDP}, \tau_{\beta, SP}, \sigma^2\}$$

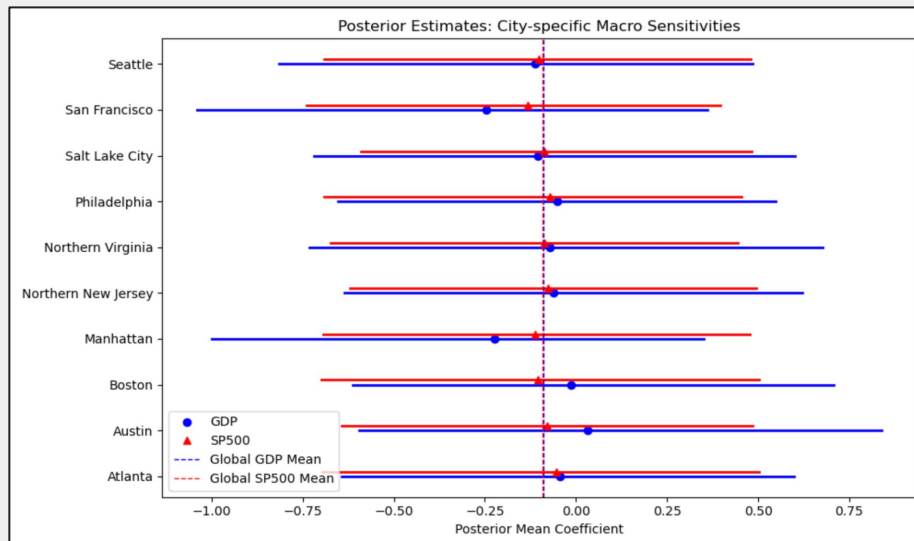
Given the data $D = \{y_i, x_i\}$, the posterior is:

$$p(\theta | D) \propto \underbrace{\prod_{i=1}^N \mathcal{N}(y_i | \alpha_{j[i]} + \beta_{j[i]}^{GDP} x_i^{GDP} + \beta_{j[i]}^{SP} x_i^{SP}, \sigma^2)}_{\text{Likelihood}} \times$$

$$\underbrace{\prod_{j=1}^J \mathcal{N}(\alpha_j | \mu_\alpha, \tau_\alpha^2) \cdot \mathcal{N}(\beta_j^{GDP} | \mu_{\beta, GDP}, \tau_{\beta, GDP}^2) \cdot \mathcal{N}(\beta_j^{SP} | \mu_{\beta, SP}, \tau_{\beta, SP}^2)}_{\text{City-level priors}} \times$$

$$\underbrace{\mathcal{N}(\mu_\alpha | 0, 10^2) \cdot \mathcal{N}(\mu_{\beta, GDP} | 0, 5^2) \cdot \mathcal{N}(\mu_{\beta, SP} | 0, 5^2)}_{\text{Hyperpriors on means}} \cdot \underbrace{\text{HalfNormal}(\tau_\alpha | 5) \cdot \text{HalfNormal}(\tau_{\beta, GDP} | 2) \cdot \text{HalfNormal}(\tau_{\beta, SP} | 2)}_{\text{Hyperpriors on std devs}} \cdot \underbrace{\text{HalfNormal}(\sigma | 1)}_{\text{Likelihood noise prior}}$$

- **Modeling Insight – Shrinkage:**
 - Extreme values (ex: **Manhattan's GDP effect**) are pulled toward the average
 - **Partial pooling** reduces overconfidence and better reflects uncertainty
- **Macroeconomic Impacts:**
 - **GDP** has a weak but positive effect on rents
 - **Stock market** gains have a positive influence



- **City-Specific Dynamics:**
 - **San Francisco and Manhattan:** Rent trends are less tied to GDP (possibly inverse)
 - **Phoenix, Atlanta, Houston:** More responsive to economic growth
 - **Stock market effects** are broadly positive and more consistent across cities