

Relative Predictors and Logistic Model for Future Power

1 Construct relative predictors (US vs. UK)

Let t index years. For any variable X , define the *log ratio*

$$x_t \equiv \ln \frac{X_t^{\text{US}}}{X_t^{\text{UK}}},$$

and for shares in $(0, 1)$ use *log-odds*.

Affordability (wheat). Given $A_t^{\text{US}} = p_t^{\text{US}}/y_t^{pc, \text{US}}$ and A_t^{UK} ,

$$a_t \equiv \ln \frac{A_t^{\text{US}}}{A_t^{\text{UK}}} \quad (\text{higher} \Rightarrow \text{wheat less affordable in US vs UK}).$$

Often flip the sign so “cheaper in US” is positive: $\tilde{a}_t \equiv -a_t$.

GDP (level, not per capita).

$$g_t \equiv \ln \frac{\text{GDP}_t^{\text{US}}}{\text{GDP}_t^{\text{UK}}}.$$

Military/industrial capacity (relative global shares). Let S_t^{US} and S_t^{UK} denote the US and UK shares of global military capacity:

$$S_t^{\text{US}} = \frac{\text{MilEx}_t^{\text{US}}}{\sum_i \text{MilEx}_t^i}, \quad S_t^{\text{UK}} = \frac{\text{MilEx}_t^{\text{UK}}}{\sum_i \text{MilEx}_t^i},$$

where MilEx_t^i is military expenditure for country i at time t . Then:

$$m_t \equiv \ln \frac{S_t^{\text{US}}}{S_t^{\text{UK}}} \quad (\text{log ratio of global military shares}).$$

This captures relative position in the global military hierarchy rather than simple bilateral comparison. Alternative: use military personnel shares MilPer_t^i instead of expenditures.

Education (e.g., avg. years, enrollment, literacy).

$$e_t \equiv \ln \frac{\text{Edu}_t^{\text{US}}}{\text{Edu}_t^{\text{UK}}}.$$

Urbanization share. For $s_t^{\text{US}}, s_t^{\text{UK}} \in (0, 1)$ use a log-odds differential:

$$u_t \equiv \underbrace{\ln \frac{s_t^{\text{US}}}{1 - s_t^{\text{US}}}}_{\text{logit}(s_t^{\text{US}})} - \underbrace{\ln \frac{s_t^{\text{UK}}}{1 - s_t^{\text{UK}}}}_{\text{logit}(s_t^{\text{UK}})}.$$

(A plain log ratio $\ln(s_t^{\text{US}}/s_t^{\text{UK}})$ also works, but log-odds behaves better near 0/1.)

Add lags to reduce simultaneity: use $x_{t-\ell}$ with $\ell \in \{1, 2\}$.

2 Define a binary “future power” target

Pick a horizon h (e.g., 5 or 10 years). Let the *relative power score* be R_t :

- **GDP leader (PTT-lite):** $R_t \equiv \ln(\text{GDP}_t^{\text{US}}/\text{GDP}_t^{\text{UK}})$.
- **Composite (optional):** $R_t = w_g g_t + w_m m_t + w_u u_t$ (standardize components, set w by PCA or equal weights).

Two binary targets:

(A) Leader at $t + h$: $Y_t^{(h)} = \mathbf{1}\{R_{t+h} \geq 0\}$ (= 1 if US leads at $t + h$).

(B) Overtake in $(t, t + h]$: $Y_t^{(h)} = \mathbf{1}\{R_t < 0 \wedge R_{t+h} \geq 0\}$.

(A) predicts who leads at horizon h ; (B) predicts overtake risk within the window. With two countries and ~ 110 years, (A) is simpler and more stable.

3 Logistic specification

Let $X_t = [\tilde{a}_{t-\ell}, g_{t-\ell}, m_{t-\ell}, e_{t-\ell}, u_{t-\ell}]$. Include a smooth time trend $f(t)$ (e.g., cubic spline) to absorb global drift:

$$\Pr(Y_t^{(h)} = 1 \mid X_t) = \Lambda\left(\beta_0 + \beta_a \tilde{a}_{t-\ell} + \beta_g g_{t-\ell} + \beta_m m_{t-\ell} + \beta_e e_{t-\ell} + \beta_u u_{t-\ell} + f(t)\right),$$

where $\Lambda(z) = \frac{1}{1+e^{-z}}$.

Interpretation. $\beta_g > 0$: a higher US/UK GDP ratio today raises the probability the US leads at $t + h$. $\beta_a > 0$: cheaper wheat in the US (relative affordability positive) raises that probability (structural-change channel). Expect $\beta_m > 0$, $\beta_u > 0$, $\beta_e > 0$.

Trend. $f(t)$: natural cubic spline with 3–5 df. With only two countries, do not include country fixed effects (collinear).

Estimation. No weights needed; use robust SEs (HC1).

4 Output: a predictive variable of future relative power

The model returns

$$\hat{q}_t^{(h)} \equiv \widehat{\Pr}(Y_t^{(h)} = 1 \mid X_t) \in [0, 1],$$

interpretable as the probability the US is the leader at $t + h$ (or the overtake risk in the hazard version).

If a smooth *power score* is preferred, map the same index via a rescaled logistic:

$$S_t^{(h)} \equiv 2\hat{q}_t^{(h)} - 1 \in [-1, 1],$$

so > 0 favors US, < 0 favors UK.

5 Practical fitting details (tiny panel)

- **Horizon/lag grid:** try $h \in \{5, 10\}$, $\ell \in \{1, 2\}$.
- **Regularization:** with ~ 100 obs, add ridge penalty:

$$\max_{\beta} \sum_t \ell(\beta; Y_t, X_t) - \lambda \|\beta\|_2^2.$$

- **Validation:** time-based CV (rolling origin) to avoid look-ahead.
- **Robustness:** swap m_t across military expenditure shares (MilEx) and military personnel shares (MilPer); test composite measures combining both.

6 Minimal data pipeline (pseudocode)

1. **Build relative predictors:**

$$\tilde{a}_t = -\ln \frac{A_t^{\text{US}}}{A_t^{\text{UK}}}, \quad g_t = \ln \frac{\text{GDP}_t^{\text{US}}}{\text{GDP}_t^{\text{UK}}}, \quad m_t = \ln \frac{S_t^{\text{US}}}{S_t^{\text{UK}}},$$

where $S_t^{\text{US}} = \text{MilEx}_t^{\text{US}} / \sum_i \text{MilEx}_t^i$ and $S_t^{\text{UK}} = \text{MilEx}_t^{\text{UK}} / \sum_i \text{MilEx}_t^i$,

$$e_t = \ln \frac{\text{Edu}_t^{\text{US}}}{\text{Edu}_t^{\text{UK}}}, \quad u_t = \text{logit}(s_t^{\text{US}}) - \text{logit}(s_t^{\text{UK}}).$$

2. **Create target $Y_t^{(h)}$:**

$$R_{t+h} = g_{t+h} \quad (\text{or composite}), \quad Y_t^{(h)} = \mathbf{1}\{R_{t+h} \geq 0\},$$

then drop the last h years.

3. **Assemble design** with lag ℓ , add spline $f(t)$.
4. **Fit** logistic (optionally ridge), **predict** $\hat{q}_t^{(h)}$.
5. **Use** $\hat{q}_t^{(h)}$ as the future power predictor; date “inflections” where $\hat{q}_t^{(h)}$ crosses 0.5 or where $\Delta \hat{q}_t^{(h)}$ peaks.