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

$$x_t \equiv \ln \frac{X_t^{\rm US}}{X_t^{\rm UK}}.$$

Based on available data, we construct the following relative predictors:

Military power ratio. Let S_t^{US} and S_t^{UK} denote the US and UK shares of global military capacity:

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

where $MilCap_t^i$ represents military capacity for country i at time t. Then:

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

This captures relative position in the global military hierarchy rather than simple bilateral comparison.

Education ratio.

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

Food price volatility ratio. Let σ_t^{US} and σ_t^{UK} denote the volatility of food prices in the US and UK respectively:

$$v_t \equiv \ln \frac{\sigma_t^{\rm US}}{\sigma_t^{\rm UK}}$$
 (higher \Rightarrow more volatile food prices in US vs UK).

Exchange rate volatility. Let σ_t^{FX} denote the volatility of the US/UK exchange rate.

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): $R_t = w_m m_t + w_e e_t + w_v v_t + w_\sigma \sigma_t^{\text{FX}}$ (standardize components, set w by PCA or equal weights).

Two binary targets:

- (A) Leader at t + h: $Y_t^{(h)} = \mathbf{1}\{R_{t+h} \ge 0\}$ (= 1 if US leads at t + h).
- **(B) Overtake in** $(t, t+h]: Y_t^{(h)} = \mathbf{1}\{R_t < 0 \land R_{t+h} \ge 0\}.$

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

3 Logistic specification

Let $X_t = [m_{t-\ell}, e_{t-\ell}, v_{t-\ell}, \sigma_{t-\ell}^{\text{FX}}]$. Include a smooth time trend f(t) (e.g., cubic spline) to absorb global drift:

$$\Pr(Y_t^{(h)} = 1 \mid X_t) = \Lambda(\beta_0 + \beta_m \, m_{t-\ell} + \beta_e \, e_{t-\ell} + \beta_v \, v_{t-\ell} + \beta_\sigma \, \sigma_{t-\ell}^{\text{FX}} + f(t)),$$

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

Interpretation. $\beta_m > 0$: higher US military share relative to UK raises the probability the US GDP advantage increases. $\beta_e > 0$: higher US education levels relative to UK raise that probability. β_v : higher food price volatility in US vs UK may indicate structural instability (sign ambiguous). β_σ : exchange rate volatility effects on future GDP advantage (sign ambiguous).

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:

$$m_t = \ln \frac{S_t^{\text{US}}}{S_t^{\text{UK}}},$$

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

$$e_t = \ln \frac{\mathrm{Edu}_t^{\mathrm{US}}}{\mathrm{Edu}_t^{\mathrm{UK}}}, \quad v_t = \ln \frac{\sigma_t^{\mathrm{US}}}{\sigma_t^{\mathrm{UK}}}, \quad \sigma_t^{\mathrm{FX}} = \text{exchange rate volatility}.$$

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

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

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.