

Caught in the Crossfire: Investigating Economic Uncertainty in Taiwan Using Stochastic Volatility in Mean VARs

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Abstract

This research proposal outlines a study to investigate the sources and impacts of economic uncertainty in Taiwan using order-invariant stochastic volatility in mean vector autoregressions (OI-SVMVAR). Building on the methodological framework developed by Davidson, Hou, and Koop (2025), this study addresses critical gaps in understanding how macroeconomic versus financial uncertainty affects Taiwan's small open economy. The research will construct a large-scale dataset of over 40 monthly variables spanning macroeconomic, financial, and global indicators, and employ advanced Bayesian estimation techniques to identify time-varying uncertainty sources. Key innovations include: (1) using a methodologically robust, order-invariant framework that overcomes limitations of small-scale models; (2) distinguishing between domestic macroeconomic and financial uncertainty channels; (3) incorporating US variables and US-China relationship indicators to account for Taiwan's unique geopolitical and economic position; and (4) allowing for time-varying classification of ambiguous variables. The findings will provide crucial policy insights for the Central Bank of China (Taiwan) and contribute to the

broader literature on uncertainty in small open economies.

1 Introduction

Economic uncertainty has emerged as a fundamental driver of business cycles, influencing investment decisions, consumption patterns, and policy effectiveness. While extensive research has examined uncertainty in large, relatively closed economies like the United States, understanding of uncertainty dynamics in small open economies remains limited and potentially biased by methodological constraints.

Taiwan presents a particularly compelling case for investigating economic uncertainty. As a small open economy deeply integrated into global supply chains, heavily dependent on exports (especially to China and the United States), and positioned at the geopolitical crossroads of US-China relations, Taiwan faces multiple sources of uncertainty that may operate through distinct transmission channels. Recent events—including the US-China trade war, global supply chain restructuring, and heightened geopolitical tensions—have amplified the importance of understanding which types of uncertainty matter most for Taiwan’s economic stability.

This research proposes to apply the cutting-edge methodological framework developed by Davidson, Hou, and Koop (2025, hereafter DHK) to investigate economic uncertainty in Taiwan. DHK demonstrate that three critical methodological issues plague existing uncertainty research: (1) model size matters—small models with approximately 30 variables produce biased estimates and incorrect conclusions; (2) variable ordering in large VARs creates order-dependence problems that compromise inference; and (3) researcher-imposed classification of variables as “macroeconomic” or “financial” may be inappropriate given that many variables exhibit time-varying characteristics.

The DHK framework addresses all three issues simultaneously through an order-invariant SVMVAR that accommodates large datasets and allows for time-varying, data-driven classification of ambiguous variables. Applying this robust methodology to Taiwan will enable us to answer fundamental policy questions that previous studies, constrained by small models or order-dependent methods, could not credibly address.

1.1 Research Motivation

First, this research exploits a novel application of Davidson, Hou, and Koop (2025)’s methodological framework to address a question they did not examine: through which transmission channels—macroeconomic (real economy) or financial (capital markets)—do external uncertainty shocks impact small open economies? While DHK (2025) investigate which type of uncertainty (macro vs. financial) dominates within the US economy, we leverage their “unclassified variables” feature in an innovative way: by treating all external shock sources (US variables, China variables, global indicators, and US-China relations indices) as unclassified, the model can objectively identify whether these shocks transmit through macroeconomic or financial channels, and how this mechanism evolves over time. Traditional VAR models cannot answer this question without imposing arbitrary block exogeneity restrictions that DHK’s order-invariant framework was designed to avoid (Chan, 2020; Davidson et al., 2025). This distinction is fundamental: whereas DHK ask “which uncertainty type affects the US?”, we ask “which transmission channel do external shocks use to impact Taiwan?”—a qualitatively different question with direct implications for monetary policy design in small open economies. The methodological innovation lies not in extending DHK’s framework, but in repurposing their time-varying classification mechanism to identify external shock transmission pathways rather than domestic uncertainty decomposition.

Second, methodological rigor requires large-scale modeling to avoid omitted variable bias. DHK (2025) demonstrate that models with approximately 30 variables yield substantially different—and incorrect—conclusions compared to models with 43 variables, a finding that echoes earlier work on Bayesian shrinkage in large VARs (Banbura et al., 2010). For Taiwan, a small open economy where domestic fluctuations are predominantly driven by external factors, omitting US, Chinese, and global variables would severely bias estimates. The most rigorous existing Taiwan study, ?, employs only six variables (four Taiwan variables plus two China variables) in a structural VAR. While pioneering in documenting that Chinese economic policy uncertainty significantly affects Taiwan’s output and exchange rate, this small-scale approach likely suffers from the omitted variable bias

that DHK’s large-model framework was designed to overcome. Furthermore, empirical evidence from comparable small open economies suggests that “global uncertainty delivers deeper and more long-lasting effects compared to within-country uncertainty,” underscoring the necessity of explicitly modeling external shock sources rather than treating them as residual disturbances. The risk of misattribution is particularly acute for Taiwan: without including US Federal Reserve policy, Chinese credit conditions, and bilateral trade policy uncertainty, the model may incorrectly attribute externally-driven volatility to Taiwan’s domestic uncertainty factors.

Third, understanding transmission channels is more policy-relevant for small open economies than merely knowing impact magnitudes. Brianti (2025) demonstrates that macroeconomic uncertainty shocks trigger deflationary patterns, allowing central banks to simultaneously stabilize output and inflation through accommodative policy, whereas financial uncertainty shocks may require policymakers to trade off price stability against output stabilization. For the Central Bank of China (Taiwan), knowing whether external shocks (e.g., US Federal Reserve rate hikes, US-China trade tensions) transmit primarily through macroeconomic channels (affecting export demand and industrial production) or financial channels (affecting capital flows, credit spreads, and exchange rates) directly determines the appropriate policy toolkit and response timing. Moreover, this transmission mechanism may shift across different episodes: the 2008 global financial crisis likely featured predominantly financial channel transmission through capital flow reversals and credit market disruptions, while the 2018 US-China trade war plausibly operated through macroeconomic channels via supply chain restructuring and export demand uncertainty. DHK (2025)’s time-varying classification framework is ideally suited to identify these regime-dependent dynamics, which previous Taiwan studies imposing fixed classifications (?) cannot capture. The policy stakes are substantial: if external shocks transmit primarily through financial channels, exchange rate flexibility and macroprudential tools become paramount; if macroeconomic channels dominate, fiscal coordination and structural adjustment policies warrant greater emphasis.

Fourth, Taiwan’s simultaneous economic dependence on both the United States and

China creates a unique empirical opportunity to study dual external exposures that cannot be examined in other settings. Existing literature examines US uncertainty spillovers to emerging markets (Carriere-Swallow and Céspedes, 2013) and Chinese uncertainty spillovers to Asia separately, but no study quantitatively decomposes transmission channels for an economy simultaneously and deeply integrated with both major powers. Firm-level evidence shows that Taiwanese companies’ revenue and profitability in China declined sharply after 2012, with further deterioration following the 2018 US-China trade war onset, even as Taiwan simultaneously became the US’s 7th largest trading partner by 2024. Unlike other emerging markets that can partially reorient toward either the US or China, Taiwan’s deep integration with both US technology supply chains and Chinese manufacturing ecosystems makes transmission channel identification particularly salient for policy design. Quantifying whether US Federal Funds Rate shocks transmit through financial channels (capital flow reversals and exchange rate volatility) versus macroeconomic channels (export demand through US recession effects), and whether this differs fundamentally from Chinese Industrial Production shocks or US-China trade policy uncertainty, addresses a critical gap in the small open economy literature. This dual exposure framework provides generalizable insights for other economies navigating great power competition, including South Korea, Singapore, and ASEAN nations facing similar though less intense dual dependencies.

2 Literature Review

2.1 Economic Uncertainty: Measurement and Effects

The modern literature on economic uncertainty can be traced to seminal contributions by Bloom (2009) and Bloom (2014), who demonstrate that uncertainty shocks can generate significant real economic effects through wait-and-see mechanisms and risk premia channels. Subsequent research has developed various uncertainty measures, including implied volatility indices (Baker et al., 2016), forecast disagreement metrics (Boero et al., 2008), and model-based stochastic volatility estimates (Jurado et al., 2015).

A critical methodological debate concerns whether to use reduced-form proxies (e.g., VIX, policy uncertainty indices) versus model-based structural measures. Ludvigson et al. (2021) argue that structural uncertainty measures derived from factor-augmented VAR models provide more reliable identification of uncertainty shocks. The DHK approach extends this structural perspective by explicitly modeling heterogeneous uncertainty sources within a unified framework.

2.2 Uncertainty in Small Open Economies

Research on uncertainty in small open economies remains relatively scarce. Carriere-Swallow and Céspedes (2013) examine uncertainty effects in Canada using a small-scale SVAR, while Cesa-Bianchi et al. (2014) investigate multiple European countries. However, these studies typically employ models with 10–30 variables and may therefore suffer from the omitted variable bias documented by DHK.

For Taiwan specifically, existing studies have explored various uncertainty dimensions. [Note: Specific Taiwan literature would be added here based on literature search]. However, these studies have not employed the methodologically rigorous, large-scale, order-invariant framework that DHK demonstrate is necessary for credible inference.

2.3 SVMVAR Models and Order Invariance

The use of stochastic volatility in mean (SVM) specifications within VAR frameworks has gained prominence following Clark and Ravazzolo (2016) and Carriero et al. (2016). These models allow time-varying volatility to directly affect the conditional mean of variables, capturing the notion that heightened uncertainty can depress economic activity.

A critical challenge in large VAR models is the order-dependence problem. Traditional approaches like Carriero et al. (2019) use triangular identification schemes that impose arbitrary variable orderings. Chan (2020) partially address this through Minnesota-type priors, but their approach still maintains sequential processing of variables.

DHK (2025) make a crucial methodological advance by developing a novel MCMC algorithm that achieves complete order invariance while maintaining computational tractabil-

ity in large systems. Furthermore, their framework allows variables to be classified as “macroeconomic,” “financial,” or “unclassified,” with the model endogenously determining time-varying probabilities that unclassified variables belong to each category. This innovation is particularly valuable for Taiwan, where the macroeconomic versus financial nature of key variables (exchange rates, policy rates, equity indices) may shift across different economic regimes.

2.4 US-China Relations and Taiwan’s Economy

Taiwan’s economy is uniquely exposed to US-China dynamics. Hsu and Chiang (2021) document that US-China trade tensions affect Taiwan through both direct trade diversion effects and indirect supply chain impacts. Wang et al. (2020) examine how cross-strait political relations influence Taiwan’s financial markets. However, no existing study has formally incorporated US-China relationship indicators into a structural uncertainty framework for Taiwan.

3 Research Questions and Hypotheses

This research addresses the following primary questions:

3.1 Primary Research Questions

RQ1: What are the primary sources of economic uncertainty affecting Taiwan’s real economy?

Specifically, does macroeconomic uncertainty (volatility in real activity, trade, and prices) or financial uncertainty (volatility in asset prices, credit conditions, and capital flows) have a more substantial impact on Taiwan’s GDP, industrial production, and employment?

Hypothesis 1a: Following DHK’s findings for the US, we hypothesize that financial uncertainty will have a more pronounced negative effect on Taiwan’s real economy than macroeconomic uncertainty.

Hypothesis 1b: However, given Taiwan’s export dependence, macroeconomic uncertainty (particularly export-related volatility) may play a more important role than in the US context.

RQ2: How do key ambiguous variables in Taiwan’s economy shift between macroeconomic and financial classifications over time?

Variables such as the TWD/USD exchange rate, TAIEX stock index, policy interest rate, and monetary aggregates may exhibit time-varying characteristics. When and why do these variables behave more like macroeconomic versus financial variables?

Hypothesis 2: We hypothesize that:

- The TWD/USD exchange rate will classify as macroeconomic during normal periods (reflecting trade fundamentals) but as financial during crisis periods (reflecting capital flow reversals).
- The TAIEX will predominantly classify as financial but may shift toward macroeconomic classification during periods when equity markets strongly reflect real economic fundamentals (e.g., during COVID-19).
- Policy rates will show time-varying classification depending on whether monetary policy primarily responds to inflation/output gaps (macroeconomic) versus financial stability concerns (financial).

RQ3: What is the role of external factors—particularly US variables and US-China relations—in generating uncertainty for Taiwan?

How do US Federal Funds Rate changes, US industrial production volatility, US credit spreads, and measures of US-China relationship tensions transmit to Taiwan? Through which channels (macroeconomic vs. financial) do these external shocks primarily operate?

Hypothesis 3: We hypothesize that:

- US monetary policy (FFR) will transmit to Taiwan through both macroeconomic channels (export demand) and financial channels (capital flows and exchange rates).
- US-China relationship indicators will primarily transmit through macroeconomic

channels (trade and supply chain effects) but may also operate through financial channels (risk premia and investor sentiment).

RQ4: Does model size matter for Taiwan as it does for the US?

Following DHK’s methodology, we will estimate both small-scale models (approximately 30 variables) and large-scale models (43+ variables) to assess whether inference about uncertainty effects differs substantially across specifications.

Hypothesis 4: Consistent with DHK, we hypothesize that small-scale models will overestimate the impact of uncertainty and potentially misattribute effects between macroeconomic and financial sources.

3.2 Policy-Relevant Questions

PQ1: Should the Central Bank of China (Taiwan) prioritize monitoring and responding to macroeconomic uncertainty or financial uncertainty?

PQ2: Which specific indicators provide the most reliable real-time signals of damaging uncertainty shocks for Taiwan?

PQ3: How does Taiwan’s vulnerability to uncertainty compare across different historical episodes (1997 Asian Financial Crisis, 2008 Global Financial Crisis, 2015 Chinese stock market crash, COVID-19, US-China trade war)?

4 Methodology

4.1 The Order-Invariant SVMVAR Framework

Following DHK (2025), we specify an order-invariant stochastic volatility in mean vector autoregression for the $N \times 1$ vector of endogenous variables y_t :

$$y_t = c + \Phi_1 y_{t-1} + \cdots + \Phi_p y_{t-p} + \Psi h_t + \varepsilon_t, \quad \varepsilon_t \sim N(0, \Sigma_t) \quad (1)$$

where $h_t = (h_{m,t}, h_{f,t})'$ is a 2×1 vector of latent uncertainty factors (macroeconomic and financial), Ψ is an $N \times 2$ matrix of factor loadings, and Σ_t is the time-varying

covariance matrix.

The stochastic volatility processes follow:

$$h_{j,t} = \mu_j + \phi_j(h_{j,t-1} - \mu_j) + \eta_{j,t}, \quad \eta_{j,t} \sim N(0, \sigma_{\eta,j}^2), \quad j \in \{m, f\} \quad (2)$$

The key innovation in DHK's approach is the order-invariant treatment of the covariance matrix Σ_t and the classification scheme for variables.

4.2 Variable Classification

Variables are classified into three categories:

1. **Macroeconomic variables** (y_t^m): Clearly real economy variables (industrial production, employment, retail sales, CPI, etc.)
2. **Financial variables** (y_t^f): Clearly financial market variables (stock returns, stock volatility, credit growth, interest rate spreads, etc.)
3. **Unclassified variables** (y_t^u): Ambiguous variables (exchange rates, policy rates, monetary aggregates, housing prices, commodity prices)

For unclassified variables, the model estimates time-varying probabilities $\pi_{i,t}$ that variable i belongs to the macroeconomic category (with probability $1 - \pi_{i,t}$ for financial). This classification is data-driven and allows variables to shift their predominant characteristics across different economic regimes.

4.3 Identification

Identification of the two uncertainty factors ($h_{m,t}$ and $h_{f,t}$) is achieved through sign and magnitude restrictions:

- Macroeconomic uncertainty $h_{m,t}$ is required to load more heavily (in absolute value) on predetermined macroeconomic variables

- Financial uncertainty $h_{f,t}$ is required to load more heavily on predetermined financial variables
- The model allows data to determine the loading patterns for unclassified variables

4.4 Estimation

The model is estimated using Bayesian methods with Markov Chain Monte Carlo (MCMC) simulation. Following DHK, we implement a novel MCMC algorithm that achieves order invariance through:

1. Joint sampling of all volatility states and covariance parameters
2. Symmetric treatment of variables within classification categories
3. Data-driven classification updating for unclassified variables

Prior specifications follow DHK:

- Minnesota-type priors on VAR coefficients Φ_i
- Conjugate priors on h_t evolution parameters
- Dirichlet priors on classification probabilities for unclassified variables

We will run the MCMC sampler for 50,000 iterations with a burn-in period of 25,000 draws, retaining every 5th draw to reduce autocorrelation, yielding 5,000 posterior draws for inference.

4.5 Model Comparison

To assess the importance of model size (RQ4), we estimate three specifications:

1. **Small model:** 30 variables (comparable to typical existing studies)
2. **Large model:** 43+ variables (our baseline specification)
3. **Robustness checks:** Alternative variable selections and classifications

Model comparison will be conducted using marginal likelihood approximations, out-of-sample forecasting performance, and substantive comparison of impulse response functions.

5 Data

5.1 Data Requirements

Following DHK’s emphasis on model size, we aim to construct a dataset with 43+ monthly frequency variables spanning approximately 1990–present (or the longest available period). This section outlines the proposed variables organized by classification category.

5.2 Macroeconomic Variables (19 variables)

Output and Activity:

1. Industrial Production Index (IPI)
2. Manufacturing Production Index
3. Export Orders Index
4. Real Retail Sales
5. Real Food Service Sales
6. Real Exports (USD)
7. Real Imports (USD)
8. Manufacturing PMI (M-Score)
9. Non-Manufacturing PMI (NMI)
10. (Business Cycle Signal Score)

Prices:

11. CPI (year-over-year growth)
12. Core CPI (year-over-year growth)
13. Wholesale Price Index (WPI, year-over-year growth)
14. Import Price Index (USD-denominated, year-over-year growth)
15. Export Price Index (USD-denominated, year-over-year growth)

Labor Market:

16. Unemployment Rate (seasonally adjusted)
17. Manufacturing Employment (seasonally adjusted)
18. Services Employment (seasonally adjusted)
19. Real Manufacturing Wage (year-over-year growth)

5.3 Financial Variables (10 variables)

Interest Rates and Spreads:

1. Overnight Call Loan Rate
2. 10-Year Government Bond Yield
3. Term Spread (10Y yield - overnight rate)
4. VIX (US, as global risk sentiment proxy)

Equity Markets:

5. TAIEX Monthly Return
6. TAIEX Daily Average Trading Volume
7. TAIEX Monthly Volatility (computed from daily data)
8. Net Foreign Investment in Taiwan Stocks

9. Margin Trading Balance (change)
10. Short Selling Balance (change)

5.4 Unclassified Variables (14+ variables)

This is the key innovation category, where variables exhibit ambiguous or time-varying macroeconomic-financial characteristics.

Policy and Money:

1. CBC Rediscount Rate (policy rate)
2. M1b growth (year-over-year)
3. M2 growth (year-over-year)

Credit and Asset Prices:

4. Total Bank Loans and Investments (year-over-year growth)
5. Consumer Loans (year-over-year growth)
6. TAIEX Level (index value, not return)
7. Credit Spread (5Y A-rated corporate - 5Y government bond)
8. Housing Price Index (year-over-year growth)

Exchange Rates:

9. TWD/USD Exchange Rate (spot)
10. TWD/USD Exchange Rate Volatility (monthly)
11. Real Effective Exchange Rate (REER) Index

Global Variables (US):

12. US Federal Funds Rate

13. US Industrial Production Index

14. US BAA-AAA Credit Spread

US-China Relations:

15. US-China Trade Policy Uncertainty Index (or related measure)

16. Cross-strait Relations Indicator (to be constructed or sourced)

5.5 Data Sources

Primary data sources will include:

- Taiwan: National Statistics, R.O.C., Central Bank of China (Taiwan), Taiwan Stock Exchange, Taiwan Economic Journal (TEJ)
- US variables: Federal Reserve Economic Data (FRED)
- Global indicators: Bloomberg, Thomson Reuters
- US-China relations: Policy uncertainty indices, news-based measures

5.6 Data Transformations

Following DHK and standard practice in VAR literature:

- Growth rates: Compute year-over-year log differences for most quantity variables
- Seasonal adjustment: Apply X-13-ARIMA-SEATS where needed
- Stationarity: Transform variables to ensure stationarity (verified by unit root tests)
- Outlier treatment: Identify and address extreme outliers (e.g., COVID-19 period)
- Standardization: Standardize all variables to have mean zero and unit variance

6 Expected Outcomes and Contributions

6.1 Methodological Contributions

1. First application of order-invariant SVMVAR to a small open economy

This research will be the first to apply DHK’s methodologically rigorous framework to a small open economy context. We will demonstrate whether the model size and order-invariance issues documented for the US also apply—and potentially with greater severity—to Taiwan.

2. Extension to incorporate external economy variables

While DHK analyze the US as a relatively closed economy, Taiwan’s openness requires explicit modeling of external factors. Our incorporation of US variables and US-China relationship indicators represents a meaningful extension of the framework to small open economy settings. The time-varying classification of these external variables (macroeconomic vs. financial transmission) is a novel contribution.

3. Validation of large-model necessity in different economic contexts

By comparing small versus large model estimates, we will provide additional evidence on the generalizability of DHK’s model size findings beyond the US context.

6.2 Empirical Contributions

1. Identification of Taiwan’s dominant uncertainty sources

We will provide the first credible estimates—using a methodologically robust framework—of whether macroeconomic or financial uncertainty poses greater risks to Taiwan’s real economy. This directly addresses a question that previous studies, using smaller or order-dependent models, could not answer reliably.

2. Documentation of time-varying variable characteristics

The time-varying classification feature will reveal how key Taiwan variables (TWD/USD, TAIEX, policy rates) shift their economic roles across different historical episodes:

- 1997 Asian Financial Crisis

- 2008 Global Financial Crisis
- 2011-2012 European Debt Crisis
- 2015 Chinese stock market crash
- 2018-2019 US-China trade war
- 2020 COVID-19 pandemic
- 2022-2023 global inflation and monetary tightening

3. Quantification of external shock transmission

We will decompose uncertainty effects into:

- Domestic macroeconomic uncertainty
- Domestic financial uncertainty
- External (US) macroeconomic uncertainty
- External (US) financial uncertainty
- US-China relationship uncertainty

This decomposition will clarify Taiwan's vulnerability structure and identify which external factors matter most.

4. Comparison with small-model estimates

By demonstrating how conclusions differ between small and large models, we will highlight potential biases in existing Taiwan uncertainty literature and provide guidance for future research.

6.3 Policy Contributions

1. Guidance for Central Bank of China (Taiwan)

Our findings will directly inform CBC policy by:

- Identifying which type of uncertainty (macroeconomic vs. financial) requires priority monitoring
- Revealing which specific indicators provide early warning signals
- Clarifying when exchange rate movements reflect trade fundamentals versus financial instability
- Assessing the relative importance of domestic versus external uncertainty sources

2. Financial stability implications

Understanding the time-varying nature of financial uncertainty and its transmission to the real economy will help inform macroprudential policy design.

3. Trade and geopolitical risk management

Quantifying the impact of US-China relationship uncertainty will provide evidence-based input for economic policy planning in the face of geopolitical tensions.

7 Timeline

The proposed research will be conducted over 18 months:

Months 1–3: Data Collection and Preparation

- Assemble 43+ variable database for Taiwan
- Obtain US variables and construct US-China relationship indicators
- Implement all data transformations and stationarity checks
- Create complete documentation of data sources and definitions

Months 4–6: Model Implementation

- Implement or adapt DHK’s MCMC algorithm (R/MATLAB/Python)
- Conduct initial estimation runs and convergence diagnostics
- Debug and optimize computational performance

- Validate implementation against DHK’s published results (using US data)

Months 7–10: Estimation and Analysis

- Estimate small-model specification (30 variables)
- Estimate large-model specification (43+ variables)
- Conduct robustness checks with alternative specifications
- Compute impulse response functions, variance decompositions, and historical decompositions
- Analyze time-varying classification patterns

Months 11–14: Interpretation and Policy Analysis

- Interpret findings in Taiwan’s economic and institutional context
- Compare results across different historical episodes
- Develop policy implications for CBC and other stakeholders
- Conduct counterfactual analyses

Months 15–18: Writing and Dissemination

- Write complete research paper
- Prepare presentations for conferences
- Prepare policy brief for CBC and government agencies
- Revise based on feedback and submit to journals

8 Potential Challenges and Mitigation Strategies

8.1 Data Availability

Challenge: Some Taiwan variables may not have sufficiently long time series (e.g., before 1990).

Mitigation:

- Prioritize most critical variables with longest available series
- For shorter series, begin analysis from later start date (e.g., 1995 or 2000)
- Explore interpolation methods for quarterly variables where appropriate
- Consult with Taiwan Economic Journal (TEJ) and academic data centers for historical data access

8.2 Computational Complexity

Challenge: DHK report that estimating a 43-variable model requires approximately 30 hours of computation time. Multiple specifications and robustness checks could require prohibitive computational resources.

Mitigation:

- Secure access to high-performance computing resources
- Optimize code implementation for parallel processing
- Implement efficient MCMC sampling techniques (e.g., Rao-Blackwellization)
- Prioritize most critical model specifications

8.3 Model Adaptation

Challenge: Adapting DHK's framework to include external variables and potentially distinguish domestic vs. global uncertainty (rather than macro vs. financial) may require substantial modification of the MCMC algorithm.

Mitigation:

- Begin with exact replication of DHK's framework for Taiwan data
- Introduce extensions gradually and validate each step
- Maintain close communication with DHK authors if possible

- Consider simpler extensions if full adaptation proves infeasible (e.g., treating US variables as exogenous in reduced-form analysis)

8.4 Interpretation and Policy Relevance

Challenge: Ensuring that findings are policy-relevant rather than purely methodological.

Mitigation:

- Frame research questions around specific Taiwan policy puzzles from the outset
- Engage with CBC researchers and policymakers during project to ensure relevance
- Translate technical findings into clear policy narratives
- Provide both academic paper and accessible policy brief deliverables

9 Conclusion

This research proposal outlines an ambitious but feasible study to investigate economic uncertainty in Taiwan using state-of-the-art econometric methods. By applying the methodologically rigorous, order-invariant SVMVAR framework of Davidson, Hou, and Koop (2025) to Taiwan’s unique economic context, this research will:

1. Provide the first credible estimates of macroeconomic versus financial uncertainty effects in Taiwan using a large-scale, order-invariant framework
2. Reveal how key economic variables shift between macroeconomic and financial characteristics across different regimes
3. Quantify the role of external factors—particularly US variables and US-China relations—in transmitting uncertainty to Taiwan
4. Demonstrate the importance of model size and methodological rigor for small open economy research

5. Deliver actionable policy insights for the Central Bank of China and other economic authorities

The research addresses critical gaps in both the methodological and empirical literature on uncertainty, while providing substantial policy value for Taiwan's unique economic and geopolitical position. The combination of cutting-edge econometric methods, comprehensive data collection, and careful attention to policy relevance positions this project to make significant contributions to both academic knowledge and real-world economic policy.

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