

# Capstone Project: Complete Bayesian Forecasting System for Commodities

**Course:** Bayesian Regression and Time Series Forecasting for Commodities Trading

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## Project Overview

In this capstone project, you will build a **complete end-to-end Bayesian forecasting and trading system** for a portfolio of commodities. This project integrates all concepts learned throughout the course.

- Bayesian inference and prior selection
  - Time series analysis and decomposition
  - Bayesian regression and structural time series
  - Hierarchical models for multiple assets
  - Volatility modeling and risk management
  - Proper backtesting and evaluation
  - Trading strategy implementation
- 

## Project Requirements

### Deliverables

- 1 . **Data Pipeline:** Download and preprocess at least 3 - 5 related commodities
- 2 . **Exploratory Analysis:** Time series characteristics, seasonality, correlations
- 3 . **Bayesian Models:** At least 2 different model types with proper diagnostics
- 4 . **Forecasting System:** Multi-step probabilistic forecasts
- 5 . **Trading Strategy:** Uncertainty-aware position sizing
- 6 . **Backtesting:** Walk-forward validation with proper metrics
- 7 . **Risk Analysis:** VaR, CVaR, drawdown analysis
- 8 . **Written Report:** 5 -page summary with key findings

### Grading Rubric

Component	Weight	Criteria
Technical Implementation	40 %	Code quality, model sophistication, diagnostics
Performance	30 %	Forecast accuracy (CRPS), trading metrics (Sharpe)
Risk Management	20 %	Uncertainty quantification, position sizing
Documentation	10 %	Clear explanation, reproducibility

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# Part 1 : Data Pipeline and Exploratory Analysis

## 1 . 1 Select Your Commodity Portfolio

Choose ONE of these commodity complexes:

### Option A: Energy Complex

- WTI Crude Oil (CL=F)
- Brent Crude Oil (BZ=F)
- Natural Gas (NG=F)
- Gasoline (RB=F)
- Heating Oil (HO=F)

### Option B: Agricultural Complex

- Corn (ZC=F)
- Wheat (ZW=F)
- Soybeans (ZS=F)
- Soybean Oil (ZL=F)
- Soybean Meal (ZM=F)

### Option C: Precious Metals

- Gold (GC=F)
- Silver (SI=F)
- Platinum (PL=F)
- Copper (HG=F)

### Option D: Custom (subject to approval)

- Choose 3 - 5 related commodities with economic rationale

In [ ]:

```
# Setup
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from scipy import stats
import pymc as pm
import arviz as az
import warnings
warnings.filterwarnings('ignore')

# Course utilities
import sys
sys.path.append('..')
from datasets.download_data import (
    get_commodity_data,
    get_multiple_commodities,
    prepare_commodity_dataset,
    create_train_test_split
)
from utils.plotting import (
```

```

    plot_time_series,
    plot_forecast,
    plot_posterior,
    plot_backtest_results
)
from utils.metrics import (
    crps, crps_gaussian,
    forecast_summary,
    backtest_summary
)
from utils.backtesting import (
    WalkForwardValidator,
    Backtester,
    create_folds
)

np.random.seed(42)
plt.style.use('seaborn-v0_8-whitegrid')

print("Setup complete!")

```

```

In [ ]: # TODO: Define your commodity portfolio
# Example for Energy Complex:

COMMODITIES = [
    'crude_oil',
    'natural_gas',
    'gasoline',
    # Add more commodities...
]

START_DATE = '2015-01-01'
END_DATE = None # Today

# Download data
# YOUR CODE HERE
data = get_multiple_commodities(COMMODITIES, start=START_DATE, end=END_DATE)
print(f"Downloaded {len(data.columns)} commodities")
print(f"Date range: {data.index[0]} to {data.index[-1]}")
print(f"Total observations: {len(data)}")
data.head()

```

```

In [ ]: # TODO: Exploratory Data Analysis
# 1. Plot all commodity prices (normalized)
# 2. Calculate and visualize correlations
# 3. Test for stationarity (ADF, KPSS)
# 4. Analyze seasonality patterns
# 5. Identify any structural breaks or regime changes

# YOUR CODE HERE

```

## 1.2 Document Your Findings

### Questions to Answer:

- 1 . What are the key characteristics of your chosen commodities?
- 2 . Are there clear seasonal patterns? Which commodities show strongest seasonality?
- 3 . What is the correlation structure? Are there natural pairs for spread trading?

- 4 . Are the series stationary? What transformations are needed?
- 5 . Are there any obvious regime changes in the data?

*Your EDA Summary Here:*

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- 
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## Part 2 : Bayesian Forecasting Models

Implement at least TWO of the following model types:

- 1 . **Bayesian Linear Regression** with economic factors
- 2 . **Bayesian Structural Time Series (BSTS)**
- 3 . **Hierarchical Model** for multiple commodities
- 4 . **Gaussian Process** regression
- 5 . **Stochastic Volatility** model

For each model:

- Justify your prior choices
- Run prior predictive checks
- Fit the model with PyMC
- Check convergence (R-hat, ESS, trace plots)
- Run posterior predictive checks
- Generate forecasts with uncertainty

### 2 . 1 Model 1 : [Your Choice]

**Model Description:** [Explain your model choice and why it's appropriate]

```
In [ ]: # Model 1 Implementation
#
# TODO:
# 1. Define priors with justification
# 2. Build PyMC model
# 3. Run prior predictive check
# 4. Fit model (sample from posterior)
# 5. Check convergence
# 6. Run posterior predictive check

# YOUR CODE HERE
```

### 2 . 2 Model 2 : [Your Choice]

**Model Description:** [Explain your model choice and why it's appropriate]

```
In [ ]:
```

```
# Model 2 Implementation  
#  
# YOUR CODE HERE
```

## 2 . 3 Model Comparison

Compare your models using:

- WAIC (Widely Applicable Information Criterion)
- LOO (Leave-One-Out Cross-Validation)
- Out-of-sample CRPS

```
In [ ]: # Model Comparison  
#  
# YOUR CODE HERE
```

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## Part 3 : Trading Strategy Development

Implement a trading strategy that:

- 1 . Uses your Bayesian forecasts
- 2 . Accounts for forecast uncertainty in position sizing
- 3 . Manages risk appropriately

Choose from:

- **Mean reversion** with credible interval bands
- **Trend following** with probability-based entries
- **Pairs/spread trading** with hierarchical model
- **Volatility targeting** with stochastic vol forecasts
- **Custom strategy** (describe your approach)

```
In [ ]: # Trading Strategy Implementation  
#  
# TODO:  
# 1. Define signal generation from forecasts  
# 2. Implement position sizing (Kelly criterion or similar)  
# 3. Add risk management rules  
# 4. Generate signals for full history  
  
def generate_trading_signals(forecasts, actuals, model_uncertainty):  
    """  
    Generate trading signals from Bayesian forecasts.  
  
    Parameters:  
    -----  
    forecasts : pd.DataFrame  
        Posterior predictive samples or point forecasts  
    actuals : pd.Series  
        Actual price series  
    model_uncertainty : pd.Series  
        Posterior standard deviation
```

```

Returns:
-----
pd.Series
    Signal series (-1 to 1)
"""

# YOUR CODE HERE
pass

def bayesian_position_size(signal, forecast_mean, forecast_std,
                           max_position=1.0, kelly_fraction=0.5):
"""
Calculate position size using Bayesian Kelly criterion.

Parameters:
-----
signal : float
    Trading signal direction
forecast_mean : float
    Expected return
forecast_std : float
    Standard deviation of forecast
max_position : float
    Maximum position size
kelly_fraction : float
    Fraction of Kelly (0.5 = half-Kelly)

Returns:
-----
float
    Position size
"""

# YOUR CODE HERE
pass

```

## Part 4 : Backtesting and Evaluation

Perform rigorous backtesting with:

- 1 . Walk-forward validation (no look-ahead bias)
- 2 . Proper evaluation metrics (CRPS for forecasts, Sharpe for trading)
- 3 . Comparison to benchmark strategies

```
In [ ]: # Walk-Forward Backtesting
#
# TODO:
# 1. Set up walk-forward validation folds
# 2. For each fold: fit model, generate forecasts, calculate signals
# 3. Compute evaluation metrics per fold
# 4. Aggregate results

# YOUR CODE HERE
```

In [ ]:

```
# Full Backtest  
#  
# YOUR CODE HERE
```

```
In [ ]: # Performance Summary  
#  
# TODO: Create comprehensive performance report  
  
# YOUR CODE HERE
```

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## Part 5 : Risk Analysis

Analyze the risk characteristics of your strategy:

- 1 . Value at Risk (VaR) and Conditional VaR (CVaR)
- 2 . Maximum drawdown analysis
- 3 . Tail risk assessment
- 4 . Scenario analysis (what if volatility doubles?)

```
In [ ]: # Risk Analysis  
#  
# YOUR CODE HERE
```

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## Part 6 : Final Report

Write a 5 -page summary covering:

- 1 . Executive Summary ( 0 . 5 page)**
  - Key findings and performance metrics
- 2 . Data and Methodology ( 1 page)**
  - Commodity selection rationale
  - Model choices and prior justification
- 3 . Model Results ( 1 . 5 pages)**
  - Posterior analysis
  - Forecast accuracy
  - Model comparison
- 4 . Trading Performance ( 1 page)**
  - Backtest results
  - Risk metrics
  - Comparison to benchmarks
- 5 . Conclusions and Limitations ( 1 page)**
  - Key insights

- What worked, what didn't
- Suggestions for improvement

## Your Report Here

(Use markdown formatting for your written report)

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### 1 . Executive Summary

[Your summary here]

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### 2 . Data and Methodology

[Your methodology here]

---

### 3 . Model Results

[Your results here]

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### 4 . Trading Performance

[Your performance analysis here]

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### 5 . Conclusions and Limitations

[Your conclusions here]

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## Submission Checklist

Before submitting, ensure you have:

- Downloaded and preprocessed at least 3 commodities
- Completed exploratory data analysis with visualizations
- Implemented at least 2 Bayesian models with proper diagnostics
- Justified all prior choices
- Generated probabilistic forecasts
-

Implemented a trading strategy with uncertainty-aware position sizing

- - Performed walk-forward backtesting
  - 
  - Calculated proper evaluation metrics (CRPS, Sharpe, etc.)
  - 
  - Completed risk analysis (VaR, drawdown, etc.)
  - 
  - Written 5 -page summary report
  - 
  - All code runs without errors
  - 
  - Results are reproducible (random seed set)
- 

**Good luck! This project is your opportunity to demonstrate mastery of Bayesian forecasting in commodity trading.**