

# Coverage Testing in Non-Stationary Time Series Forecasts

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This repository accompanies the following **submitted manuscript**:

**Title:**

***Testing Marginal and Conditional Coverage in Non-Stationary Time Series Forecasts through Value-at-Risk Backtesting***

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## Project Summary

This project introduces a **statistical testing framework** for evaluating **Conformal Prediction (CP)** under **non-stationary time series**. Inspired by Value-at-Risk (VaR) backtesting, we assess CP methods using **formal hypothesis tests** for marginal validity, temporal independence, and conditional coverage.

We evaluate CP methods on:

- **Synthetic data** with changepoints and drift (*Barber et al., 2023*)
  - **Electricity price forecasts** (ELEC2 dataset from Kaggle)
  - **S&P 100 stock forecasts** (retrieved from Yahoo Finance API)
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## Folder Structure

```
data/
├── ELEC2/
│   ├── elec_simulation_CP.csv
│   ├── elec_simulation_nexCP_LS.csv
│   └── elec_simulation_nexCP_WLS.csv
├── Financial Timeseries/
│   ├── pred_calib.csv
│   ├── pred_test.csv
│   ├── quantile_calib.csv
│   ├── quantile_test.csv
│   └── Wide_S_P_100_forecasts.csv
└── Synthetic Data/
    ├── simulation_CP_LS.csv
    ├── simulation_nexCP_LS.csv
    └── simulation_nexCP_WLS.csv
```

## Repository Files

File	Description
<code>ConformalPredictor.py</code>	Implements Split CP, CQR, and ACI
<code>TestingFramework.py</code>	Contains all 8 formal backtests
<code>Synthetic_data.ipynb</code>	Runs CP+LS, NexCP+LS, NexCP+WLS on synthetic data (Barber et al.)
<code>Elec_data.ipynb</code>	Applies all 3 Barber methods to the ELEC2 dataset
<code>Modeltrainer.ipynb</code>	Trains LightGBM model on S&P 100 stock return data
<code>Calib_Analysis.ipynb</code>	Applies Split CP, CQR, ACI and all tests to financial data
<code>requirements.txt</code>	Python packages needed to reproduce results

## CP Methods Used

Synthetic & ELEC2 (Barber et al., 2023)

- **CP+LS**: Full Conformal Prediction with Least Squares
- **NexCP+LS**: Non-exchangeable CP with exponential weights
- **NexCP+WLS**: Non-exchangeable CP with Weighted Least Squares

Financial Time Series (S&P 100)

- **Split CP**
- **Conformalized Quantile Regression (CQR)**
- **Adaptive Conformal Inference (ACI)**

## Statistical Tests (8 total, grouped in 4 categories)

Category	Purpose	Included Tests
<b>1. Marginal Coverage</b>	Is average coverage close to target rate $\alpha$ ?	<ul style="list-style-type: none"><li>• Binomial Test (Lower)</li><li>• Binomial Test (Upper)</li><li>• Binomial Test (Two-sided)</li><li>• First Geometric Test (for parameter (a))</li></ul>
<b>2. Violation Independence</b>	Are violations temporally independent?	<ul style="list-style-type: none"><li>• Second Geometric Test (for parameter (b))</li></ul>
<b>3. Conditional Coverage</b>	Are violations predictable from inputs?	<ul style="list-style-type: none"><li>• Joint Geometric Test (for (a) and (b))</li><li>• Dynamic Binary Test</li></ul>
<b>4. Interval Score Comparison</b>	Are intervals both valid and efficient?	<ul style="list-style-type: none"><li>• Comparative Interval-Score Test (Diebold-Mariano)</li></ul>

## Installation

```
git clone https://github.com/KonradRtz/Coverage-Testing-in-Non-stationary-Time-Series.git
cd Coverage-Testing-in-Non-stationary-Time-Series
pip install -r requirements.txt
```

Dependencies include:

`numpy`, `pandas`, `scipy`, `matplotlib`, `scikit-learn`, `lightgbm`

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## How to Run

### Synthetic Data

Run `Synthetic_data.ipynb` to evaluate CP+LS, NexCP+LS, and NexCP+WLS.

Default predictions are included in `data/Synthetic Data/`.

**Alternative:** You may regenerate synthetic predictions using the official implementation by Barber et al.:

[https://rinafb.github.io/code/nonexchangeable\\_conformal.zip](https://rinafb.github.io/code/nonexchangeable_conformal.zip)

Convert the output to:

```
method, time, true_y, lower_bound, upper_bound, violation
```

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### Electricity Forecasts (ELEC2)

Run `Elec_data.ipynb` using files in `data/ELEC2/`.

You may optionally regenerate predictions using the same external code and formatting as above.

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### Financial Forecasts

1. Train forecasting model in `Modeltrainer.ipynb`
  2. Evaluate with Split CP, CQR, and ACI in `Calib_Analysis.ipynb` using Yahoo Finance-based data
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## Output

Results include full statistical test reports, p-values, and interval score comparisons per method.

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## References

- **Barber, Candès, Ramdas, Tibshirani (2023)**  
*Conformal Prediction Beyond Exchangeability*

[arXiv:2202.13415](#)

- **Romano, Patterson, Candès (2019)**

*Conformalized Quantile Regression*

[arXiv:1905.03222](#)

- **Gibbs & Candès (2021)**

*Adaptive Conformal Inference*

[arXiv:2106.00170](#)

- **ELEC2 Dataset**

<https://www.kaggle.com/datasets/yashsharan/the-elec2-dataset>

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