



#### Dengue Cases Forecasting Based on eXtreme Gradient Boosting Ensemble with Coyote Optimization

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# AGENDA

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- Objective
- Dataset description
- Methodology
- Results
- Conclusion
- References



# INTRODUCTION

- Dengue is considered a public health problem in tropical regions. The development of efficient forecasting models is important to supporting health care officials to optimally disseminate available resources in the dengue-prone areas.
- In **2020** in **Brazil**, there were **979,764** probable cases reported (incidence rate of **466.2** cases per **100** thousand inhabitants) (Brazil, 2020).
- There are a **limited number of studies** devoted to obtaining **efficient forecasting models** worldwide, especially for the **Brazil** case (Hoyos et al., 2021).



### INTRODUCTION

- Climatic, demographic, and social factors makes the development of forecasting models a challenging task.
- Cooperative Ensembles models can handle with that the time series.
- Hybridize eXtreme gradient boosting (XGBoost) with metaheuristics for hyperparameters tuning can be a suitable approach for that.





# OBJECTIVE

 The objective of this study is to explore the viability of using the Coyote Optimization Algorithm (COA) (Pierezan and Coelho, 2018), to tuning the XGBoost hyperparameters for the task of forecasting multi-step-ahead dengue incidence in Parana, Brazil.



# DATASET DESCRIPTION

- The dataset refers to the Dengue cases number in PR available on DATASUS.
- Dengue cases is the **output** variable.
- Lagged values of dengue and exogenous variables such as precipitation (mm), maximal and minimum temperature (°C), thermal amplitude, and humidity are considered as features.



Figure 1: Dengue cases over the time for PR state.



# METHODOLOGY

- **XGBoost** with linear booster is used.
- Five different optimization methods c adopted for comparison:
  - **CS** (Cuckoo Search Optimization).
  - **DE** (Differential Evolution).
  - **FFA** (Firefly Algorithm).
  - **GWO** (Grey Wolf Optimzer).
  - **GA** (Genetic Algorithm).
- Performance measures:
  - MAPE (Mean absolute percentage error).
  - **RMSE** (Root mean squared error).



Figure 2: Flowchart of proposed workflow.



### RESULTS

RESULTS OF THE OPTIMIZED XGBOOST MODEL IN TERMS OF RMSE AND MAPE (30 RUNS) TO FORECAST DENGUE CASES ONE UP TO THREE-MONTHS-AHEAD.

Forecasting	Statistical	COA		CS		DE		FFA		GA		GWO	
Horizon	Indicator	RMSE	MAPE										
One-Month-Ahead	Minimum	1713.02	0.84	1731.29	0.89	1731.15	0.86	1729.77	0.90	1714.62	0.78	1729.11	0.90
	Median	1766.03	0.95	1789.70	1.05	1804.11	1.10	1794.90	1.09	1775.90	0.98	1764.10	1.04
	Arithmetic Average	1781.11	1.04	1822.97	1.07	1867.30	1.10	1838.31	1.01	1782.03	1.03	1798.53	1.05
	Maximum	2324.51	1.37	2367.71	1.32	2469.16	1.33	2365.05	1.33	1952.43	1.31	2027.33	1.32
	Std	106.51	0.16	125.53	0.14	169.64	0.15	158.45	0.16	48.44	0.15	88.34	0.12
	Trimmead Average	1771.63	1.03	1806.79	1.07	1850.67	1.10	1823.37	1.10	1778.35	1.03	1792.84	1.05
Two-Months-Ahead	Minimum	2137.04	1.66	2134.07	1.76	2119.51	1.77	2149.11	1.67	1972.87	1.47	2148.96	1.67
	Median	2255.32	2.01	2308.71	2.08	2359.21	2.11	2342.11	2.05	2267.86	2.03	2285.98	2.06
	Arithmetic Average	2304.89	2.01	2414.14	2.12	2452.36	2.16	2482.09	2.09	2325.46	2.05	2470.95	2.12
	Maximum	2579.12	2.35	3459.93	2.83	3485.67	2.89	3524.81	2.90	2578.88	2.36	3521.92	2.90
	Std	122.48	0.18	314.00	0.23	363.39	0.28	356.59	0.27	145.87	0.18	424.53	0.36
	Trimmead Average	2301.09	2.01	2386.79	2.10	2427.34	2.15	2456.74	2.08	2329.00	2.06	2444.91	2.11
Three-Months-Ahead	Minimum	4648.69	3.94	4359.29	3.12	4481.53	3.44	4329.44	2.88	4098.81	3.90	4712.15	3.92
	Median	4716.16	6.19	4807.40	5.09	4944.75	5.34	4925.31	5.27	4794.93	6.14	4912.82	5.79
	Arithmetic Average	4837.34	5.93	4886.85	5.45	4968.31	5.62	4965.94	5.50	4841.11	5.87	5001.54	5.86
	Maximum	5128.22	7.56	5744.22	7.55	5781.80	7.96	5828.92	7.55	5198.96	8.01	5787.45	7.62
	Std	119.14	1.32	268.26	1.45	286.06	1.37	306.57	1.36	201.01	1.42	319.13	1.27
	Trimmead Average	3577.82	3.90	3630.58	3.68	3687.54	3.80	3706.88	3.71	3579.39	3.88	3711.34	3.92



#### RESULTS







Figure 5: CD plot for RMSE.







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Figure 6: CD plot for MAPE.

# CONCLUSION

- This study proposed a novel combination of COA and XGBoost to forecast dengue cases multi-step-ahead for PR state.
- The performance of the proposed XGBoost model using COA was compared with XGBoost coupled with DE, CS, COA, DE, FFA, GA, and GWO for hyperparameters tuning.
- **Regarding accuracy** (average of RMSE and MAPE, over 30 runs), the proposed **COA-XGBoost** outperforms compared approaches.
- The **ranking** of **feature importance** is dengue cases lags, maximal temperature, precipitation, minimal temperature, humidity, and thermal amplitude.

#### • For future works

- Performing spatio-temporal analysis.
- Comparing COA-XGBoost with artificial neural networks and state-of-art methods.

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# Thank you!

# Any questions?



