Forecasting The Cumulative Cases of COVID-19 In Four Large Brazilian Cities Using Machine Learning Approaches

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Overview

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- 2 Objective
- 3 Dataset
- 4 Proposed forecasting framework
- 5 Findings
- 6 Conclusion

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Covid: Brazil's coronavirus cases pass five million

③ 8 October



Country Exceeds Covid-19 157,000 Deaths

Daily average of deaths in the last seven days, however, was 468, according to the consortium



Brazil reaches 150,000 deaths from COVID-19 milestone

Brazil's count of COVID-19 deaths has surpassed 150,000, despite signs the pandemic is slowly retreating in Latin America's largest nation

By MARCELO DE SOUSA and TATIANA POLASTRI 10 October 2020, 22:29 • 4 min read



New coronavirus identified, currently named Severe Acute Respiratory Syndrome CoronaVirus 2 (SARS-CoV-2), causing the coronavirus disease 2019 (COVID-19).

Brazil has more than 5 million cases and 150 thousand deaths.

Problems:

- Overcrowding of hospital beds;
- Lack of medical supplies to treat patients affected more severely;

Short-term forecast allows to:

- Reallocate patients;
- Reapportion resources and staff;
- Alert residents of a determined region in case of a spike;

Models:

- Autoregressive Integrated Moving Average (ARIMA)
- k-Nearest Neighbor (kNN)
- Support Vector Regression (SVR)
- Cubist Regression (CUBIST)
- Stacked Generalization (STACK)

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Objective

Present a time series forecasting framework based on the use of machine learning to predict cumulative cases of COVID-19 achieving accurate predictions, being possibly useful for decision making authorities.

Use forecasting models (CUBIST, kNN, and SVR) cooperatively in the form of a **Stacked Generalization** to bring the accuracy level up.

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Dataset

Aggregated data from Brazilian State Health Offices from March 13th until August 14th, 2020.

The data is gathered in a collaborative project with various contributors and made available with an API or direct download.

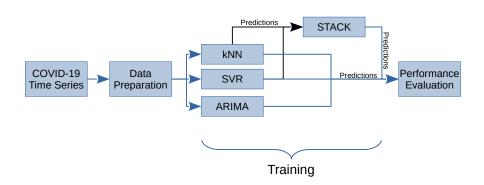
Brasil.IO

Cities: Brasília (DF); Rio de Janeiro (RJ); Salvador (BA); and São Paulo (SP).

Dataset



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Step 1

- Train-test split
- Normalization
- Past 5 days as features
- Recursive one-day-ahead model

Step 2

- Base-learners (kNN and SVR) are trained and its predictions used as input for CUBIST meta-learner.
- Leave-one-out cross-validation is used for training.
- Machine learning models use caret R package.
- ARIMA uses auto.arima function from forecast R package.
- All computations are made in R software environment.

Step 3

- Out-of-sample metrics are computed:
 - Improved percentage index (IP)
 - Mean absolute error (MAE)
 - Root-mean-square error (RMSE)
 - Symmetric mean absolute percentage error (sMAPE)

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Findings - Brasília

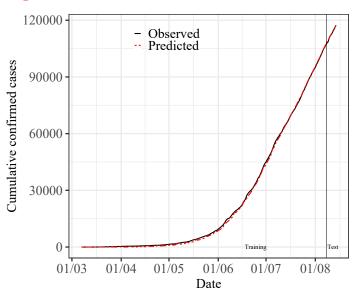
STACK approach could be considered to forecast COVID-19 cases, once the model outperformed others in all performance criteria.

Regarding MAE the performance improved in the ranges of 38.59% – 96.58%.

Ranking

- STACK
- ARIMA
- SVR
- kNN

Findings - Brasília



Findings - Rio de Janeiro

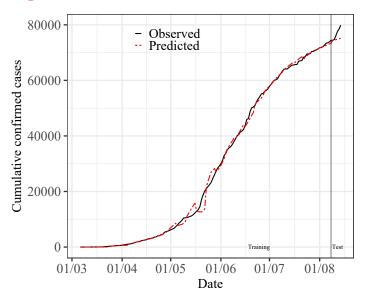
For Rio de Janeiro's cases, the SVR model achieved better accuracy than other models

The improvement in MAE ranged in 9.14% – 61.98%.

Ranking

- SVR
- ARIMA
- STACK
- kNN

Findings - Rio de Janeiro



Findings - Salvador

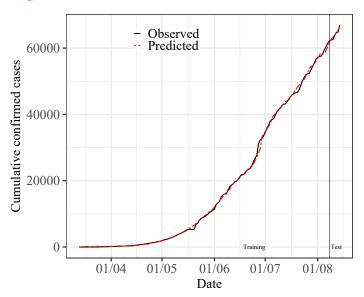
STACK approach outperformed the other models in Salvador's cases.

MAE improvement ranged between 36.78% – 88.71%.

Ranking

- STACK
- ARIMA
- SVR
- kNN

Findings - Salvador



Findings - São Paulo

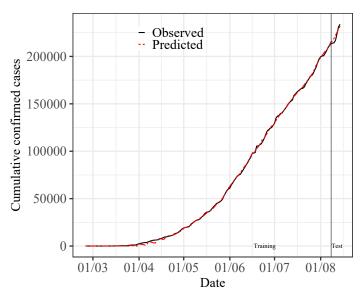
STACK approach achieved better accuracy than other models.

The improvement in MAE ranged in 20.30% – 78.01%.

Ranking

- STACK
- ARIMA
- SVR
- kNN

Findings - São Paulo



Findings - Overview

Table: Performance measures of the models

City	Criteria	ARIMA	kNN	STACK	SVR
Brasília	MAE	501.83	9020.17	308.17	3982.00
	RMSE	528.16	9471.26	394.21	4196.15
	sMAPE	0.44%	8.27%	0.28%	3.56%
Rio de Janeiro	MAE	1374.83	3285.83	2120.83	1249.17
	RRMSE	1700.76	3838.18	2734.95	1502.23
	sMAPE	1.78%	4.33%	2.76%	1.62%
São Paulo	MAE	673.33	3769.67	425.67	2346.83
	RMSE	766.89	4064.64	500.17	2429.35
	sMAPE	1.05%	6.02%	0.66%	3.57%
Salvador	MAE	3784.83	13716.50	3016.50	5277.33
	RRMSE	4242.16	15833.98	3324.96	7355.84
	sMAPE	1.69%	6.30%	1.36%	2.35%

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Conclusion

Stacking-ensemble learning and **SVR** models are suitable tools to forecast COVID-19 cases for most of the adopted cities.

These approaches were able to learn the **nonlinearities** inherent to the evaluated **epidemiological time series**.

ARIMA also showed good consistent results for a six-days-ahead time window.

Conclusion

The ranking of models in Rio de Janeiro's scenarios is:

- SVR
- 2 ARIMA
- **3** STACK
- 4 KNN

For all the other cities:

- **STACK**
- 2 ARIMA
- 3 SVR
- 4 KNN

Conclusion

Remark

- Even though the models discussed in this paper presented accurate forecasting capabilities, they should be used cautiously.
- The dynamics of the analyzed data, as well as the diversity of exogenous factors that can affect the daily notifications of COVID-19

Acknowledgments







Thank you!