Springboard—Data Science Career Tack
Proposal of Capstone Project 1\_Grocery Store Sales Forecasting
Capstone Project 1 Proposal
By Henry Ruan
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My first capstone project will be <u>based on a-the</u> Kaggle competition project: <u>"Corporacion Favorita Grocery Sales Forecasting."</u> Grocery store sales forecasting is a critical factor for store managers to growth their sales revenue, maintain a healthy level of inventory and improve their operation profit.

In this project, I'll apply consider the application of several latest data science tools and machine learning algorithms and Ppython packages, such as linear regression and ARIMA; Time series to build automatic models to predict the unit sales for thousands of items sold at different Favorita stores located in Equadorof this company.

The all am planning on considering various business activities and all level operations operational aspects that might have the an impacts on predicting the store sales, for example store location, available inventory, pricing, operation hours, holiday/events and promotion on specific type of products/items, etc.

I am planning on studying approaches to address this project by considering the following three layers: As the limitation of the time and resources, to make this capstone project practice is easy to manage and deliverable, I'll breakdown the whole project in three typical levels.

- First level is to predict the sales on typical products/items at one grocery store/location
- Second level is to predict the sales on typical products/items at multiple grocery stores/locations
- Third level is to predict the sales on typical products/items at multiple grocery stores/locations on different periods (daily, weekly, monthly, quarterly)

The overall approach taken to solve the problem is to first build familiarity with the data and business domain through descriptive statistics and data visualization, and then use this understanding to develop and test a set of hypotheses to explain the increase in sold items experienced by the stores.

Hypothesis to focus on and test: Some hypothesis to consider are:

Hypothesis 1: Type of products sold are dependent on territory/holiday/events/seasonal patterns

Hypothesis 2: <u>T</u>the <u>amountnumber</u> of units to be sold are highly dependents on the pricing, availability.

Hypothesis 3: Certain grocery stores reload the inventory more frequently.

In practice, to get the predictive models implemented, there will be a potential improvement on the overall <a href="management">management</a> of <a href="management">groceries</a> and on their activities among the stakeholders: customers, nearby community, logistics/supply chain, sales/marketing, management, operation & financing, competitors.

The data sets are available at <a href="https://www.kaggle.com/c/favorita-grocery-sales-forecasting/data">https://www.kaggle.com/c/favorita-grocery-sales-forecasting/data</a>. These data sets <a href="mailto:seem to be already mostly clean">seem to be already mostly clean</a> and saved as CSV files.

The training data<u>set</u> includes dates, store and item information, whether that item was being promoted, as well as the unit sales. Additional files include supplementary information that may be useful in building <u>the</u> models:

- holidays\_events.csv
- · items.csv
- oil.csv
- · sample\_submission.csv
- stores.csv
- test.csv
- train.csv
- transactions.csv

My project plan and schedule in alignment with curriculum and requirements, as suggested by my mentor, and with deadlines estimated by me, is as follows:

- (1) Complete and submit capstone project proposal, per unit 4.3
- (2) Complete and submit the Data Wrangling notebook+report after completing all mini-projects in unit 5 of the curriculum
- (3) Complete and submit the Data Storytelling notebook+report, per unit 7
  - (4) Complete and submit the application of Inferential Statistics notebook+report to my CP dataset after finishing all mini-projects in unit 8 of the curriculum
  - (5) Complete and submit the Milestone Report for my CP See unit 8.5 of the curriculum.
- (6) Build a baseline model notebook for your CP using "pmdarima"
  - (6.1) To assess the performance of my model/s, do the following:
  - (a) Split the dataset into training and test set, determining an appropriate split. For time series forecasting, it is customary for the training set to be in the chronological past, and for the test set in the chronological future of the training set.
  - (b) Build time series model/s using the training set.

	<ul><li>(b) Build time series model/s using the training set.</li><li>(b) Build time series model/s using the training set.</li></ul>
	(c) Compute RMSE, MAPE (*) and R^2 for the training set. This might be implemented in different ways depending on the package you are using (see "Resources" below).
	(d) Compute RMSE, MAPE and R^2 for the test set.
	(e) Determine if there is overfitting, by comparing (c) and (d).
	(f) Show the following graphs:
	(f.1) Predicted vs Actual scatterplot.
	(f.2) Predicted vs Residuals scatterplot.
	(f.3) Histogram of Residuals, identifying positive and negative residual worst-cases for 90% and 95% of the histogram density.
	(6.2) Present an analysis of the performance of the model/s in light of (6.1).
	(7) Discuss with <u>my mentor</u> applicable extensions to (6).
	Options include (see "Resources" below):
l	<ul><li>Prophet.</li><li>Deriving features to apply a regression model, and</li><li>H2O.</li></ul>
	(8) Implement the extensions agreed upon in (7).
	(9) Prepare a first draft of CP Final Report, per the Capstone Project Guidelines (CPG) document, and submit it to <a href="mailto:my mentor">my mentor</a> for review <a href="Due on 5/28/2020">Due on 5/28/2020</a>
 	(10) Use feedback from my mentor after (9) to generate the CP Final Report, Presentation Slide Deck, and cleaned-up code, and submit them as deliverables for your CP.

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## (\*) Resources provided by my mentor:

(0) This reference is well-known and respected and covers the theory of ARIMA-type of models:

Although the examples are in R, the explanations are language-agnostic. Most of the approaches presented here require manual intervention, which is why the approaches below would be appropriate when automating the forecasting with many time series.

(1) pmdraima (formerly Pyramid) is a Python implementation of auto-arima (maybe discussed (1) pmdraima (formerly Pyramid) is a Python implementation of auto-arima (maybe discussed in fpp above?). Some associated resources are:

https://github.com/tgsmith61591/pmdarima

https://github.com/tgsmith61591/pmdarima

https://www.alkaline-ml.com/pmdarima/

https://www.alkaline-ml.com/pmdarima/

Even more resources can be found by searching pyramid python time series

(2) Prophet was developed by Facebook:

https://facebook.github.io/prophet/

https://facebook.github.io/prophet/

https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-quick-start-of-time-series-forecasting-with-a-practical-https://towardsdatascience.com/a-practical-https://towardsdatascience.com/a-practical-https://towardsdatascience.com/a-practical-https://towardsdatascience.com/a-practical-https://towardsdatascience.com/a-practical-https://towardsdatascience.com/a-practical-https://

example-using-fb-prophet-31c4447a2274

More resources ... search for prophet time series

(3) More recent approaches incorporate the concept of Auto-ML. An example is H2O ... <a href="https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/blog/time-money-automate-time-series-forecasts-driverless-ai/https://www.h2o.ai/htt

(4) Time series forecasting as a supervised machine learning problem

https://machinelearningmastery.com/time-series-forecasting-supervised-learning/https://machinelearningmastery.com/convert-time-series-supervised-learning-problem-python/https://machinelearningmastery.com/convert-time-series-supervised-learning-problem-python/

- (5) Backtesting
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- (5.0) General:
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https://medium.com/cindicator/backtesting-time-series-models-weekend-of-a-data-scientist-https://medium.com/cindicator/backtesting-time-series-models-weekend-of-a-data-scientist-92079cc2c540

- (5.1) pmdraima:
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https://alkaline-

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ml.com/pmdarima/auto\_examples/model\_selection/example\_cross\_validation.html#sphx-glr-auto-examples-model-selection-example-cross-validation-py

(5.2) prophet: