

Rossmann Store Sales Prediction





CONTEXT

Rossmann, Business Problem, Business Understanding 01

CHALLENGES

Main Problem, Possible Causes, Solutions 02

METHODOLOGY

Data Description, Hypothesis Creation, Exploratory Data Analysis, ML models 03

04

CONCLUSION

Business Performance, Machine Learning Model Performance

05

DEMONSTRATION

Model Deployment through Heroku, Telegram Bot

06

NEXT STEPS

Further actions to improve project



CONTEXT

R®SSMANN

drugstore chain with more than

3.000 stores

located in Europe

BUSINESS PROBLEM



Rossmann store managers are tasked with **predicting their daily sales** for up to **six weeks in advance**

BUSINESS UNDERSTANDING



CFO wants to **invest individually in each store**, according to their incomes





CHALLENGES

budget definition for individual investment of stores

PROBLEM

Inaccuracy of general sales prediction due to divergence between stores

All sales prediction is manually made by each Rossmann store

Only possible to check prediction results through a computer



Use of a Machine Learning method to make all stores sales predictions



Make prediction results available through a smartphone





METHODOLOGY



Data Description

Sales data for 1.115 Rossmann stores

Number of Columns: 18



DATE



Date of sales, holidays

STORE



Type, assortment

CUSTOMER



Quantity

PROMOTIONS



Period, consecutive promo

COMPETITION



Distance, open since



Descriptive Statistics

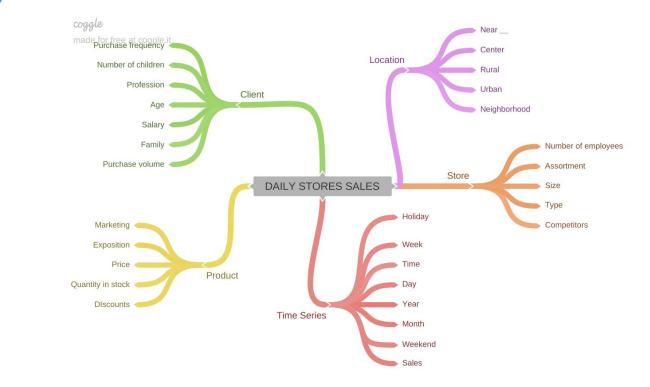


	mean	median	std	min	max	range	skew	kurtosis
store	558.429727	558.0	321.908651	1.0	1115.0	1114.0	-0.000955	-1.200524
day_of_week	3.998341	4.0	1.997391	1.0	7.0	6.0	0.001593	-1.246873
sales	5773.818972	5744.0	3849.926175	0.0	41551.0	41551.0	0.641460	1.778375
customers	633.145946	609.0	464.411734	0.0	7388.0	7388.0	1.598650	7.091773
open	0.830107	1.0	0.375539	0.0	1.0	1.0	-1.758045	1.090723
promo	0.381515	0.0	0.485759	0.0	1.0	1.0	0.487838	-1.762018
school_holiday	0.178647	0.0	0.383056	0.0	1.0	1.0	1.677842	0.815154
competition_distance	5935.442677	2330.0	12547.652996	20.0	200000.0	199980.0	10.242344	147.789712
competition_open_since_month	6.786849	7.0	3.311087	1.0	12.0	11.0	-0.042076	-1.232607
competition_open_since_year	2010.324840	2012.0	5.515593	1900.0	2015.0	115.0	-7.235657	124.071304
promo2	0.500564	1.0	0.500000	0.0	1.0	1.0	-0.002255	-1.999999
promo2_since_week	23.619033	22.0	14.310064	1.0	52.0	51.0	0.178723	-1.184046
promo2_since_year	2012.793297	2013.0	1.662658	2009.0	2015.0	6.0	-0.784436	-0.210075
is_promo	0.165966	0.0	0.372050	0.0	1.0	1.0	1.795644	1.224338



Hypothesis MindMap

Hypothesis Creation











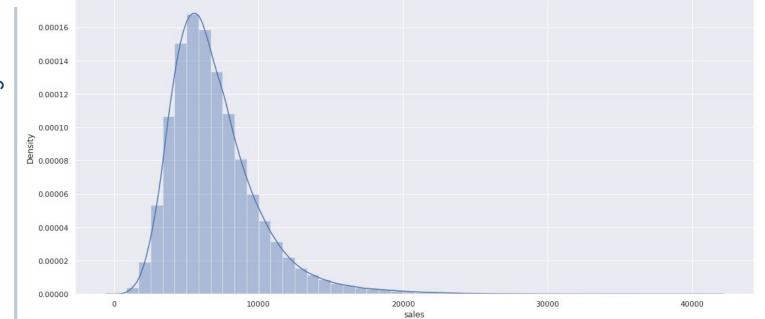
Hypothesis Creation

- 1. Stores with more diversified products should sell more
- 2. Stores with closer competitors should sell less
- 3. Stores with longer active promotions should sell more
- 4. Stores with more consecutive promotions should sell more
- 5. Stores, during school holidays, should sell more
- **6.** Stores, **in Christmas period**, should sell **more** than during other holidays
- 7. Stores, during weekends, should sell less
- 8. At the first half of the month, stores should sell more
- 9. In the first semester of the year, stores should sell less
- 10. Throughout the years, stores should sell more



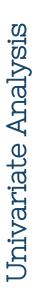








Exploratory Data Analysis





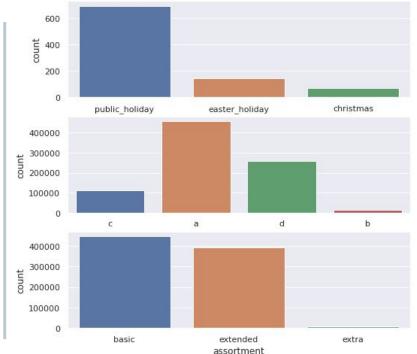


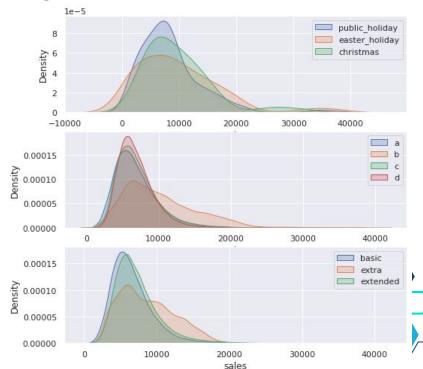
200000

100000

Exploratory Data Analysis





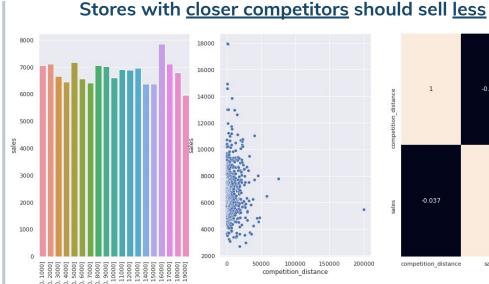


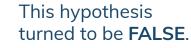




-0.037







Mean of sales does not vary much: in fact, the highest mean value is found for higher distances.

Not a strong correlation (heat map).





-0.35

-0.35





This hypothesis turned to be TRUE.

> The linear regression slope exhibits well this tendency.







This hypothesis turned to be **FALSE**.

 Even though data of 2015 is incomplete (until July), it was considered to observe the tendency of sales reduction throughout years.



Hypothesis Final Table

Hypothesis	Conclusion	Relevance
H1 - Stores with more diversified products should sell more	TRUE	Medium
H2 - Stores with closer competitors should sell less	FALSE	Low
H3 - Stores with longer active promotions should sell more	TRUE	Medium
H4 - Stores with more consecutive promotions should sell more	FALSE	Low
H5 – Stores, during school holidays, should sell more	TRUE	Low
H6 – Stores, in Christmas period, should sell more than during other holidays	FALSE	Medium
H7 - Stores, during weekends, should sell less	TRUE	High
H8 - At the first half of the month, stores should sell more	TRUE	High
H9 - In the first semester of the year, stores should sell less	TRUE	High
H10 - Throughout the years, stores should sell more	FALSE	High

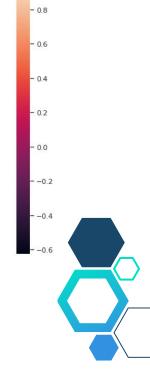


Exploratory Data Analysis

Numerical Attributes

Multivariate Analysis

store	1	-8.5e-06	0.0051	0.024	-4.7e-05	5.8e-05	0.00064	-0.014	-0.033	0.00046	0.0085	0.005	0.014	0.0049
day_of_week	-8.5e-06	1	-0.46	-0.39	-0.53	-0.39	-0.21	-4.5e-06	-0.0017	6.5e-05	0.00017	-0.002	0.00035	-0.0045
sales	0.0051	-0.46	1	0.89	0.68	0.45	0.085	-0.025	-0.0022	0.01	-0.091	0.056	0.052	-0.042
customers	0.024	-0.39	0.89	1	0.62	0.32	0.072	-0.078	-0.0074	0.0068	-0.15	0.04	0.11	-0.067
open	-4.7e-05	-0.53	0.68		1	0.3	0.086	0.0046	0.00033	0.0025	-0.0083	0.0015	0.0062	0.00013
promo	5.8e-05	-0.39	0.45	0.32	0.3	1	0.067	2.3e-05	-0.0038	0.0012	-0.00098	0.0007	0.0061	0.0042
school_holiday	0.00064	-0.21	0.085	0.072	0.086	0.067	1	-0.0029	0.035	0.00028	-0.0069	0.033	-0.0051	0.021
competition_distance	-0.014	-4.5e-06	-0.025	-0.078	0.0046	2.3e-05	-0.0029	1	-0.038	0.038	-0.1	-0.025	0.05	-0.046
competition_open_since_month	-0.033	-0.0017	-0.0022	-0.0074	0.00033	-0.0038	0.035	-0.038	1	-0.14	-0.032	0.11	0.028	-0.037
competition_open_since_year	0.00046	6.5e-05	0.01	0.0068	0.0025	0.0012	0.00028	0.038	-0.14	1	-0.015	-0.011	0.0035	-0.0094
promo2	0.0085	0.00017	-0.091	-0.15	-0.0083	-0.00098	-0.0069	-0.1	-0.032	-0.015	1	-0.024	-0.63	0.45
promo2_since_week	0.005	-0.002	0.056	0.04	0.0015	0.0007	0.033	-0.025	0.11	-0.011	-0.024	1	-0.16	-0.011
promo2_since_year	0.014	0.00035	0.052	0.11	0.0062	0.0061	-0.0051	0.05	0.028	0.0035	-0.63	-0.16	1	-0.28
is_promo	0.0049	-0.0045	-0.042	-0.067	0.00013	0.0042	0.021	-0.046	-0.037	-0.0094	0.45	-0.011	-0.28	1
	store	y_of_week	sales	customers	oben	ргото	ool_holiday	n_distance	nce_month	since_year	promo2	since_week	since_year	is_promo

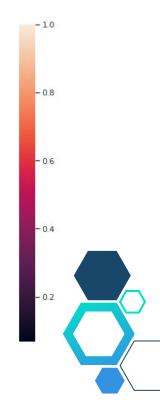


Exploratory Data Analysis

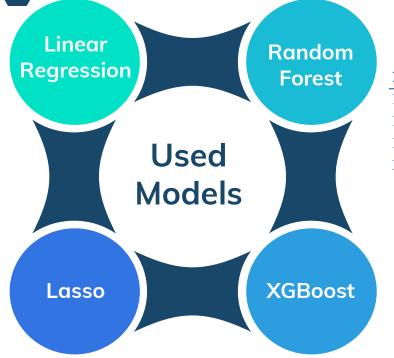


Categorical Attributes





Machine Learning Modelling



Model's Performance

Model Name	MAE CV	MAPE CV	RMSE CV
Linear Regression	2081.73 +/- 295.63	0.3 +/- 0.02	2952.52 +/- 468.37
Lasso	2116.38 +/- 341.5	0.29 +/- 0.01	3057.75 +/- 504.26
RF Regressor	837.68 +/- 219.1	0.12 +/- 0.02	1256.08 +/- 320.36
XGBoost Regressor	1030.28 +/- 167.19	0.14 +/- 0.02	1478.26 +/- 229.79

MODEL'S CHOICE



XGBoost Regressor

Lighter than RF
 Regressor, with not
 much difference in error



3.5



Business Performance

285,860,497.77\$

Total Stores Sales Prediction for the next 6 weeks

± 745,482.07 \$

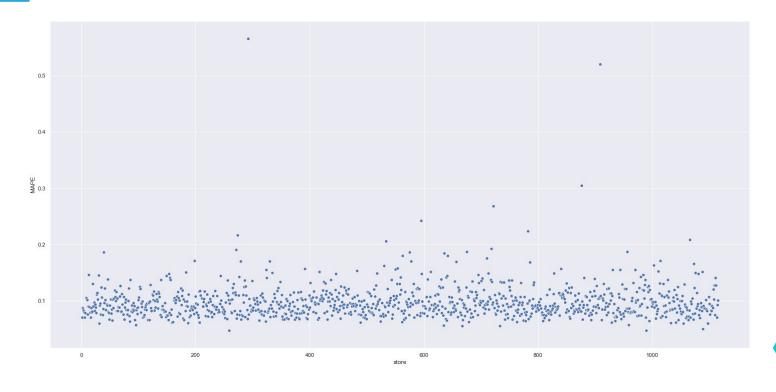
Margin of error (Worst/Best Scenario)

< 10 %

Average percentage error of the stores sales

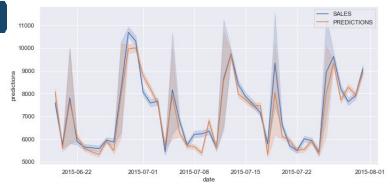


Business Performance

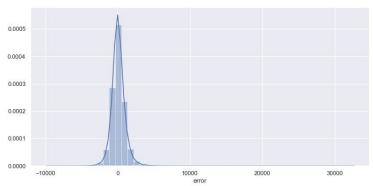


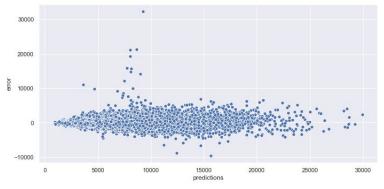


Machine Learning Performance









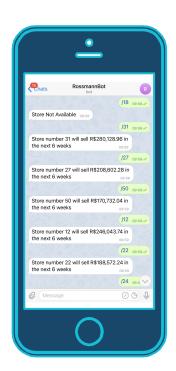




DEMONSTRATION

Model Deployment - Heroku

Telegram Bot





6. Next Steps

- Improve Model's performance;
- Test usability of deployment;
- Check for other ML models that could fit well into this study case.





THANKS!

Any questions?

You can find me at:

- LinkedIn: Diandra Melo
- GitHub: @diandramelo
- Gmail: dcxsmelo@gmail.com





I would like to give an special thanks to **Meigarom Lopes** for providing the orientation needed for me to achieve these results, as well as for improving my Data Science knowledge throughout the course **Data Science in Production**.

- Forecast sales competition Rossmann Store Sales from <u>kaggle.com</u>
- Presentation template by <u>SlidesCarnival</u>

