

DSI x L'Oréal Capstone: Time Series Sales Prediction for Kiehl's

Key words: time series, prediction model, lightGBM

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Motivation



Resource Allocation



Inventory Management



Seasonal Trend Prediction



Business Planning

Research Questions

30 days sales prediction for Kiehl's stores across the United States



- Can we identify correlation between traffic and sales?
- Can we find determinants and external factors to build traffic model?
- Can we build sales prediction model based on traffic model?

Data

Data Description		
	External Dataset	Internal Dataset
	Census Data (state & city)	Sales Data (country level)
		Traffic Data (country level)
	U.S. Holiday Dataset	U.S. Traffic Data ★

Correlated

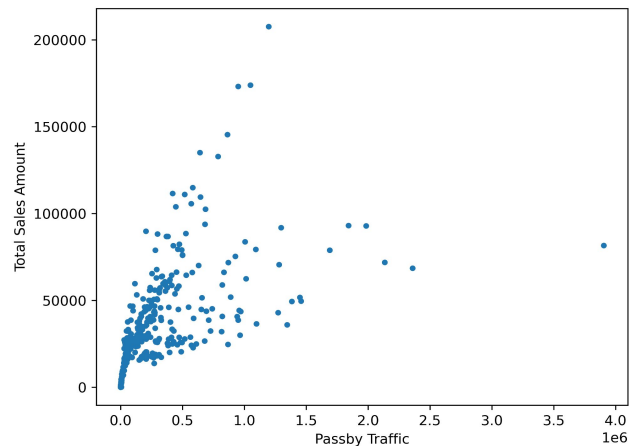


Fig 1: Scatter Plot of Traffic and Sellout

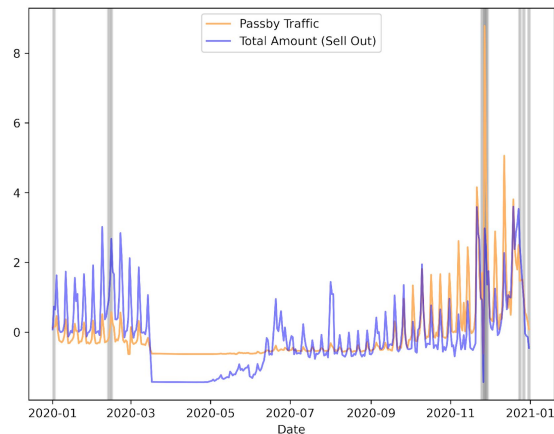



Fig 2: Passby Traffic and Sellout Amount over Time
(after standardization)

Data

- After data cleaning and preprocessing (before encoding), the dataframe involved in this project contains 89317 observations with 17 variables, ranging from 06/01/2016 to 08/18/2019
- **Categorical variables:** state, city, store type, flagship, sales zone, brand, holiday, weekday, year, month
- **Numerical variable:** number of entry (traffic), population (state-level), median household income (aka MHI, state-level), population density (city-level), mhi (city-level), income per capita (city-level)

From external census data source



Day	State	City	Store type	Flagship	Sales Zone	Brand	NbEntry	Holiday	Weekday	Year	Month	Population	MHI	Pop Density	mhi	per capita
2016/1/1	AZ	SCOTTSD ALE	Boutique	Standard	Mall	KIEHL'S	40	New Year's Day	Friday	2016	1	62.9	65913	1311.7	97409	70040
2016/1/1	CA	ARCADIA	Boutique	Standard	Mall	KIEHL'S	71	New Year's Day	Friday	2016	1	253.7	84097	5187.2	99588	47167
2016/1/1	CA	BERKELEY	Boutique	Standard	High Street	KIEHL'S	22	New Year's Day	Friday	2016	1	253.7	84097	11917.3	97834	56168
2016/1/1	CA	BREA	Boutique	Standard	Mall	KIEHL'S	52	New Year's Day	Friday	2016	1	253.7	84097	3889.3	108721	47945
2016/1/1	CA	CANOGA PARK	Boutique	Standard	Mall	KIEHL'S	25	New Year's Day	Friday	2016	1	253.7	84097	9942	84535	35589

After transformation, there are 87126 observations with 184 variables

Feature Engineering

Starting Point Baseline Model ≡

Predicts traffic by taking the average of traffic in historical data on a given date.
RMSE: 25.0

Data Augmentation

One-hot Encoding

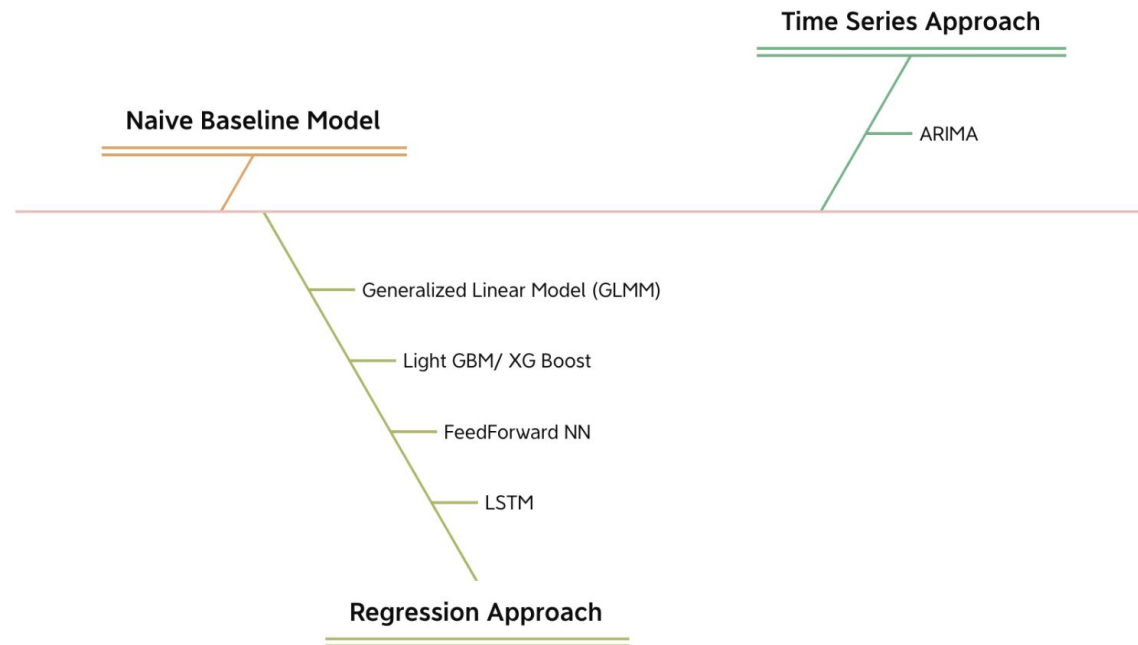
Geographical Encoding

Holiday Labels

Day of the Week

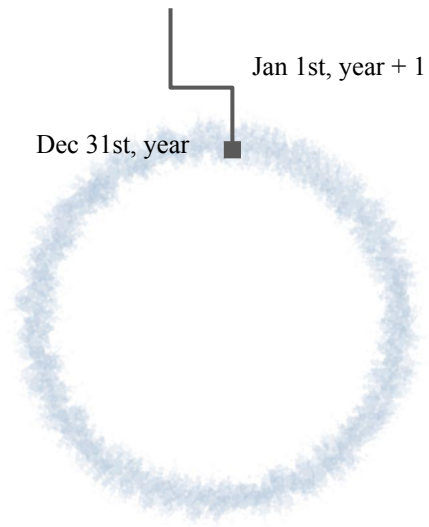
Methodology

How to construct
prediction model?



Methodology

Sine and Cosine Transformation



Lookback variables

Date	Nb of entry	1 day lookback	N day lookback
Jan 1st	Y_{t-2}	Y_{t-3}	Y_{t-2-N}
Jan 2nd	Y_{t-1}	Y_{t-2}	Y_{t-1-N}
Jan 3rd	Y_t	Y_{t-1}	Y_{t-N}

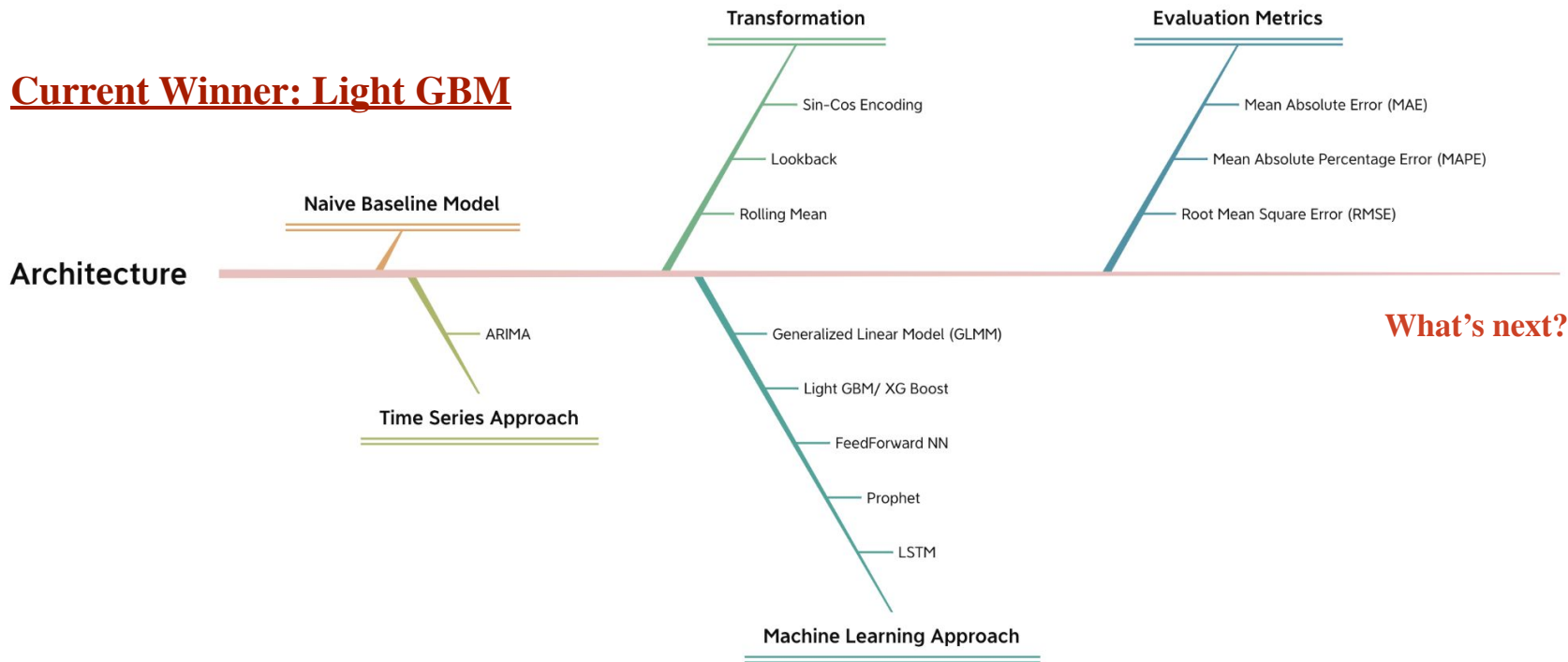
Model Evaluation

<u>Model Comparison</u>	MAPE	MAE	RMSE
Light GBM <ul style="list-style-type: none">- Sin-Cos transformation on 365 day- Lookback on 30 days- Rolling window	19.82	5.13	8.19
Light GBM	20.36	5.30	8.43
XGBoost			14.82
FeedForward NN			12.82
Generalized Linear Model (GLMM)			23.93
Prophet (with holiday parameter)			22.09
Baseline (fitting on previous year)			25



Results

Current Winner: Light GBM



Discussion

Is there any limitations besides methodology itself?

Discussion

How to combine this traffic prediction with sales model?

Capstone Mentor

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Thanks for Listening



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