

Ask the Expert

Boosting Retail & CPG Profits: Are Your Promotions Optimized?

Arpit Jain, Senior Pre-Sales Solutions Architect





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Arpit Jain is a data scientist and customer adviser for commercial industries including CPG and retail. He has extensive experience in the development and implementation of advanced analytics solutions across different industries and provides subject matter expertise on supply chain topics such as demand forecasting, promotion analytics and optimization, and pricing optimization at SAS. Prior to joining SAS, Arpit worked as a demand analyst at Nestle, where he built statistical demand forecasting and promotion optimization solutions using SAS® technology.

Promotion Analytics

Understand, Optimize & Enhance the impact of Promotions

Arpit Jain



What is Promo Analytics for?

Business
department

Commercial

Supply chain

Operations

Marketing

Development

Main processes

- Financial & assortment planning
- Pricing strategy
- Promo planning
- Supplier's negotiations

- Demand forecasting
- Inventory planning

- Operations planning
- Workforce management

- Attraction of new consumers
- Prevent current consumers from leaving

- New stores opening
- New product launches

Examples of needs

- Assortment is not effective comparing to market
- Not optimal prices reduce margin and traffic
- Measure of true promo effectiveness
- Promo effectiveness is decreasing
- Best promo assortment formation

- Promo inventory level is too high
- High amount of out-of-stock of promo products
- Regular inventory level is too high
- High amount of out-of-stock of regular products

- Low in-store service level

- Promo effectiveness is decreasing
- Current consumers are leaving

- New stores can't reach the required KPI

Data Requirements

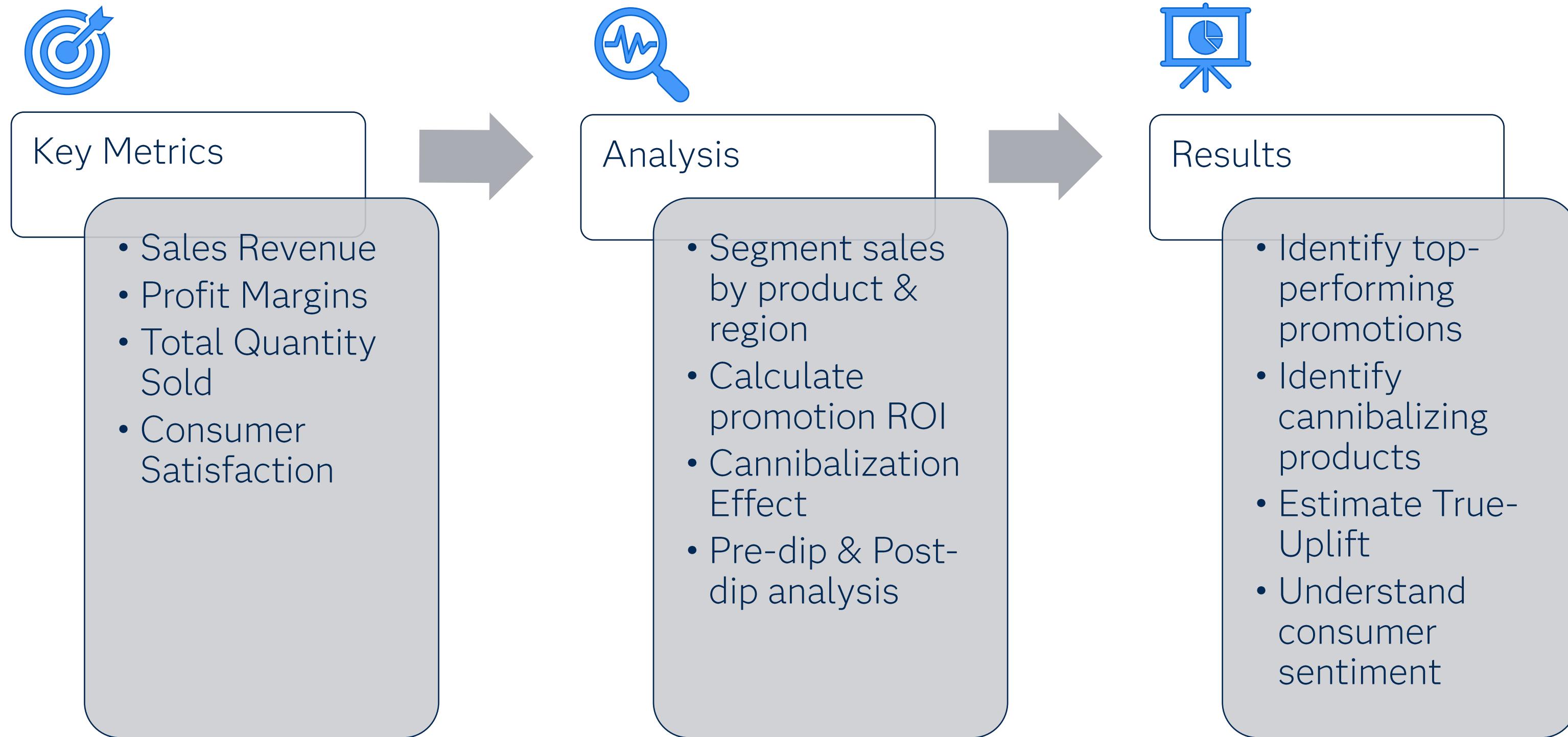
Data Requirements

Product Master	Store Master	Sales Transaction	Causal Factors	Finance Data
<ul style="list-style-type: none">• Product Key / EAN• Category• Product description• Manufacturer• Brand• Flavor• Volume• Product Group	<ul style="list-style-type: none">• Store ID• Market• Store size• Store address• Store demographics• Promotion Zone• Retail competition	<ul style="list-style-type: none">• Product Key• Store ID• Day / Week Key• Event ID• Unit Sales• Price• Discount	<ul style="list-style-type: none">• Product Key• Store ID• Day / Week Key• Event ID• Promotion Price• Promotion Type• Feature Type• Display Type	<ul style="list-style-type: none">• Product Key• Day / Week Key• Event ID• Cost of Promotions• COGS

Post-Promotion Analytics

Understanding the effectiveness of recent promotions

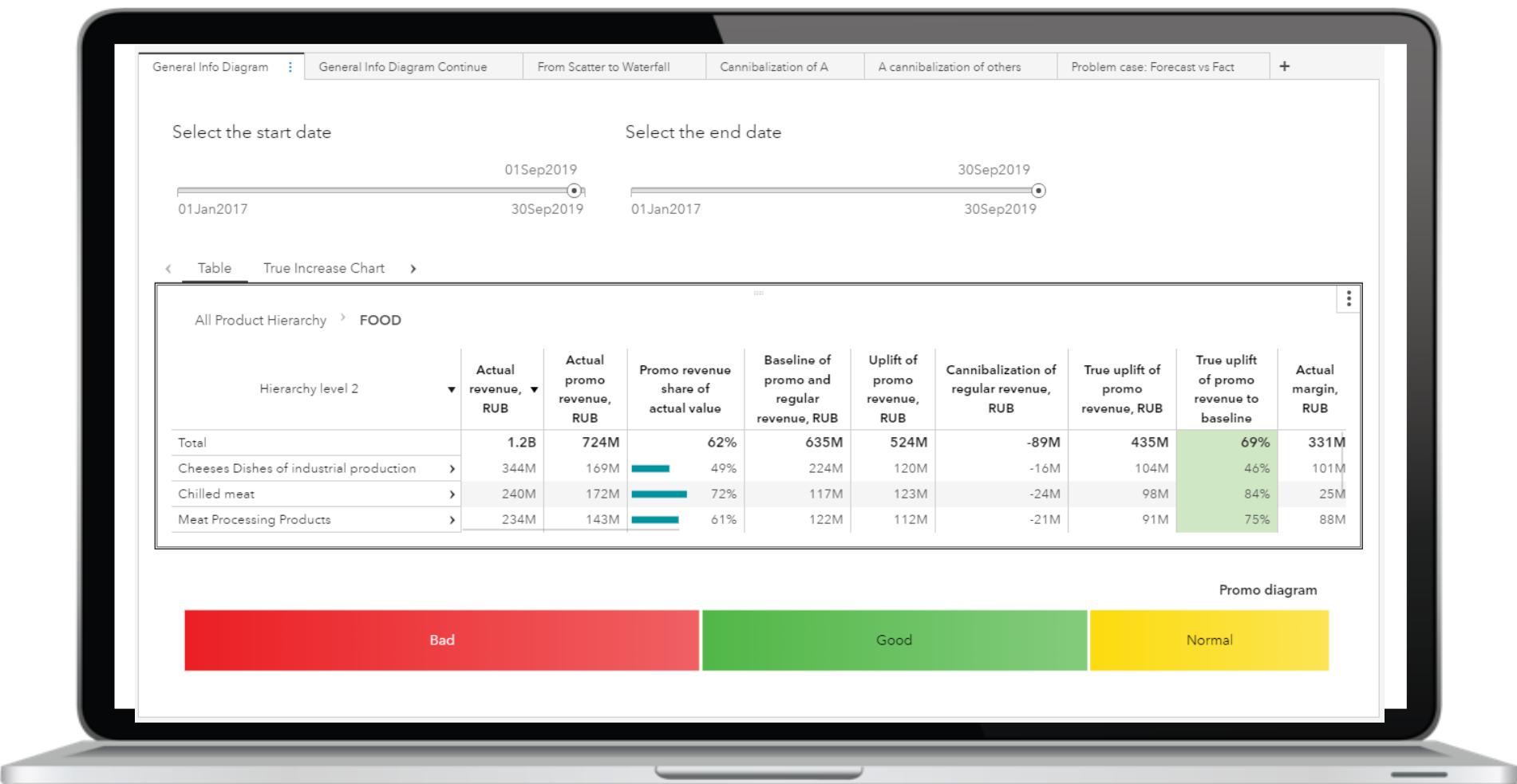
Post-Promo Analytics



Post-Promo Analytics

Understanding the effectiveness of recent promotions

Demonstration



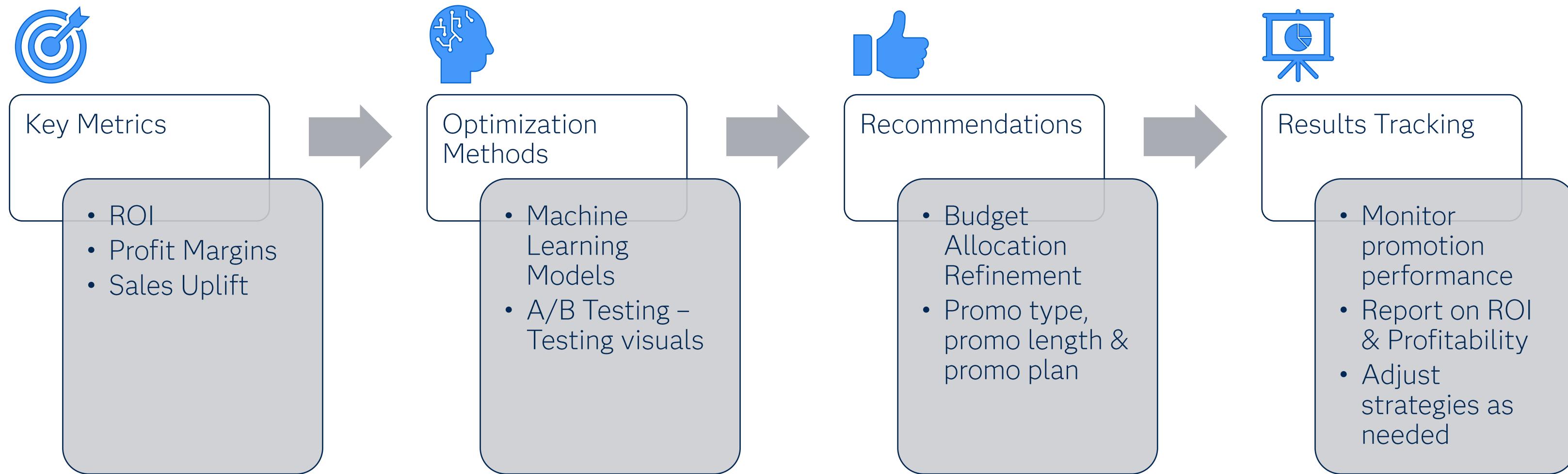
SAS Visual Analytics

Promotion Optimization

Maximize the impact and efficiency of promotions

Promotion Optimization

Maximize the impact and efficiency of promotions



Promotion Optimization

Calculate cross-elasticities between products

Cross-Elasticity: Impact of promotion of one product on the sales of another product.

1			Juice DOBRY Apple t/pack. 1L	from a mixture of fruits t/pack. 1L	Tomato juice with salt t/pack. 1L	Juice RICH Orange t/pack. 1L	RICH Exotic with/to t/pack. 2L	Juice DOBRY Apple t/pack. 2L	Nectar DOBRY Multifruit t/pack. 2L	Nectar DOBRY Orange t/pack. 2L	children from 3 years old t/pack. 2L
2		id	1604	1606	3289	13032	13034	28733	28734	28756	29038
3	Juice DOBRY Apple t/pack. 1L		1604	-4,91469	0	0	0,048177	0	0	1,271965	0
4	Nectar DOBRY Multifruit from a mixture of fruits		1606	0	-6,27866	0,652981	0	0	0	1,080248	0
5	DOBRY Tomato juice with salt t/pack. 1L		3289	0	0	-4,98708	0	0	0	0,923887	0
6	Juice RICH Orange t/pack. 1L		13032	0	0	0	-3,91535	0	0	0	0,153261
7	Juice RICH Exotic with/to t/pack. 1L		13034	0,024318	0	0,131832	0	-5,56941	0	0	0,10591
8	Juice DOBRY Apple t/pack. 2L		28733	0,016456	0	0	0	0	-5,04456	0	0
9	Nectar DOBRY Multifruit t/pack. 2L		28734	0	0	0	0,084975	0	0	-5,39928	0
10	Nectar DOBRY Orange t/pack. 2L		28756	0	0	0	0,280095	0	0	0	-5,12235
11	Dobry Tomato juice with salt and pulp. restore h		29038	0,048291	0,034997	0	0	0	0	0	-5,79941
12	Juice RICH Apple light. t/pack 1L		42970	0	0	0	0	0	0	0,058081	0
13	Juice Ya Orange with pulp. d/d.p. from 3 years o		49673	0,389203	0,037948	0	0	0,201804	0,143552	0	0
14	Juice I Apple light. restore d/d.p. from 3 years t/		49675	0	0,077048	0	0	1,206878	0	0,076053	0,137983
15	Juice Ya Grapefruit with pulp. d/d.p. from 3 year		49676	0	0	0	0,183196	0	0,12396	0	0,357863
16	Nectar I am Cherry Light. t/pack 0.97L		49677	0	0,384584	0	0	0,074817	0	0,376241	0,041053
17	Nectar XXX XXX Orange t/pack. 0.95L		50749	0,201247	0	0	0	0	0,114816	0	0
18	Nectar XXX XXX Apple t/pack. 0.95L		50750	0,185245	0	0	0	0,053552	0	0,145995	0
19	Nectar DOBRY Pineapple t/pack. 2L		59017	0	0	0	0,018189	0	0	0	0
20	RICH Tomato juice with salt reduced t/pack 1L		62481	0,000694	0	0,35232	0	0	0	0	0,280626

Promotion Optimization

Optimal promo discounts, optimal SKUs mix

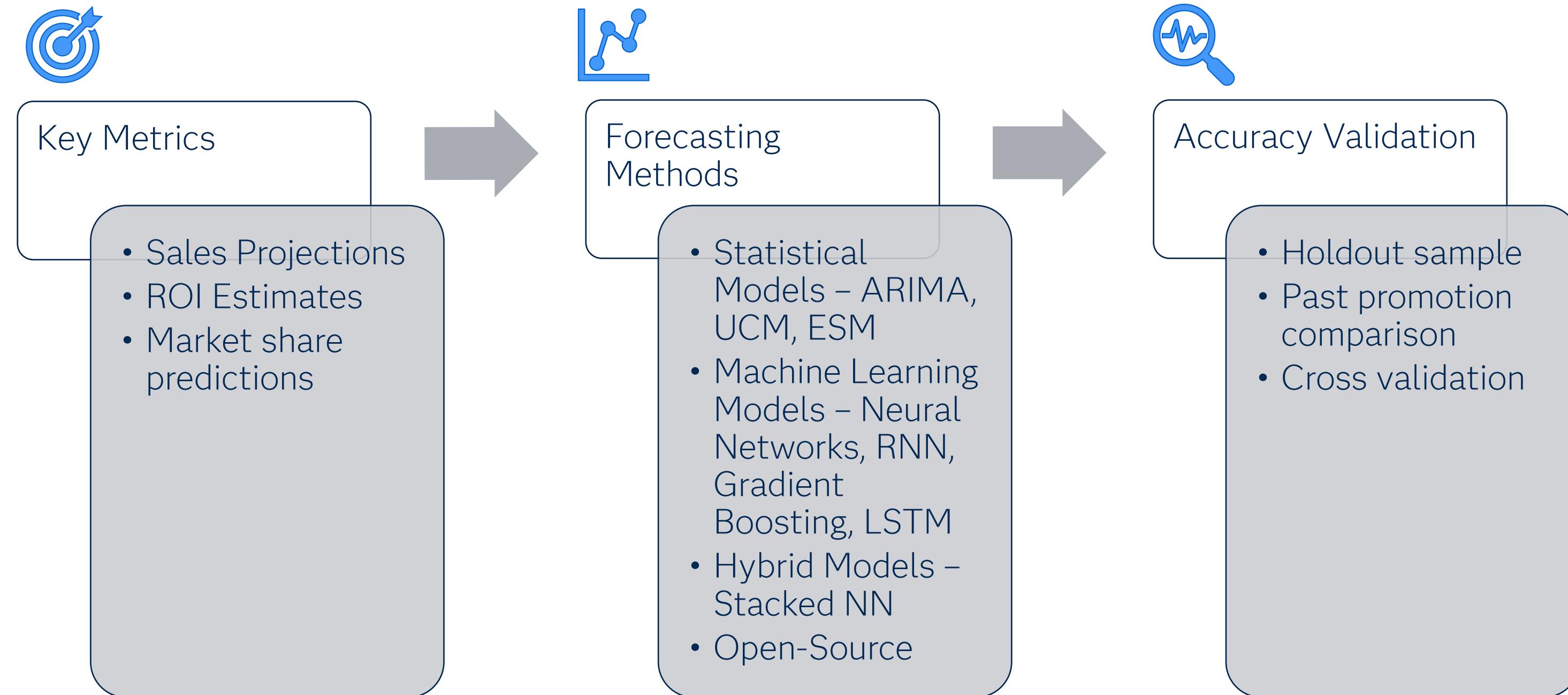
					MAPE	27%	Fcst, opt. discounts	Opt vs. acutals	Opt vs fcst	KPI not less	Recommended to delist			
					WMAPE (by revenue)	25%	19.628	-13%	0%		0			
					WMAPE (by margin)	26%	1.431.779	-8%	0%	1.800.000	0%			
					TOTAL APE	13%	523.117	-9%	0%					
Model parameters														
SKU code	SKU Name	Revenue, RUB	Margin, RUB	Num. of stores	Reg. price	Fcst Units with actual discounts	APE	Fcst Revenue with actual discounts	Fcst Margin with opt. discounts	Min discount	Max discount	Actual discount	Opt. discount	Recommended to delist
1604	Juice DOBROY Apple t/pack. (XX) 1L	62.860	23.365	52	101	279	70%	28.096	16.183	0,00	0,00	0,00	0,00	0
1606	Nectar GOOD Multifruit from a mixture of fruits t / p	25.891	10.695	52	100	463	22%	31.431	12.965	0,00	0,40	0,32	0,32	0
3289	Juice DOBROY Tomato with salt t/pack. (XX) 1L	47.561	17.652	52	101	691	1%	46.899	17.256	0,00	0,40	0,33	0,33	0
13032	Juice RICH Orange t/pack. (XX) 1L	11.809	6.879	52	132	83	7%	10.932	5.637	0,00	0,00	0,00	0,00	0
13034	Juice RICH Exotic with / to t / pack. (XX) 1L	3.429	2.101	52	147	29	28%	4.304	2.612	0,00	0,00	0,00	0,00	0
28733	Juice DOBROY Apple t/pack. (XX) 2L	36.526	19.048	52	181	128	39%	23.186	12.515	0,00	0,00	0,00	0,00	0
28734	Nectar GOOD Multifruit t / pack. (XX) 2L	19.301	10.531	52	186	66	40%	12.342	7.114	0,00	0,00	0,00	0,00	0
28756	Nectar GOOD Orange t/pack. (XX) 2L	17.912	9.503	52	185	65	35%	12.107	6.601	0,00	0,00	0,00	0,00	0
29038	Juice GOOD Tomato with salt with pulp. restore hom	10.989	5.891	50	185	42	33%	7.752	4.359	0,00	0,00	0,00	0,00	0
42970	Juice RICH Apple light. t/pack (XX) 1L	17.888	10.682	52	143	71	42%	10.101	5.957	0,00	0,00	0,00	0,00	0
49673	Juice Ya Orange with pulp. d/d.p. from 3 years restor	21.398	6.380	52	149	356	49%	32.015	9.635	0,00	0,40	0,40	0,40	0
49675	Juice I Applelight. restore d/d.p. from 3 years old t /	23.679	7.816	49	143	244	20%	21.898	7.244	0,00	0,40	0,37	0,37	0
49676	Juice I Grapefruit with pulp. d/d.p. from 3 years resto	16.747	5.275	49	143	241	29%	21.653	6.718	0,00	0,40	0,37	0,37	0
49677	Nectar I Cherry light. t/pack (XX) 0.97L	23.864	6.954	47	139	440	57%	37.085	9.468	0,00	0,40	0,39	0,39	0
50749	Nectar XXX XXX Orange t/pack. (XX) 0.95L	2.139	717	44	34	66	5%	2.234	747	0,00	0,00	0,00	0,00	0
50750	Nectar XXX XXX Apple t/pack. (XX) 0.95L	6.161	2.055	39	34	94	48%	3.192	1.071	0,00	0,00	0,00	0,00	0
59017	Nectar GOOD Pineapple t / pack. (XX) 2L	8.575	4.525	52	185	34	32%	6.242	3.508	0,00	0,00	0,00	0,00	0
62481	Juice RICH Tomato with salt recon. t/pack (XX) 1L	7.402	4.435	52	145	50	2%	7.240	4.333	0,00	0,00	0,00	0,00	0
64638	Juice DOBROY Apple t/pack. (XX) 0.2L	5.647	3.036	49	29	175	12%	5.152	2.850	0,00	0,00	0,00	0,00	0
67094	Nectar FRUIT GARDEN Orange with pulp. d/d.p. t/pac	70.761	18.160	51	173	910	10%	81.792	21.248	0,00	0,50	0,48	0,48	0
67095	Nectar FRUIT GARDEN Peach-apple with pulp. d/d.p.t	58.539	15.417	51	173	434	33%	39.023	10.089	0,00	0,50	0,48	0,48	0
67096	Nectar FRUIT GARDEN Multifruit with pulp d/d.p. t/pa	61.904	16.205	52	176	643	6%	57.831	14.910	0,00	0,50	0,49	0,49	0
77894	Juice SADY PRIDONIA Apple direct extraction t/pak. (12.486	8.254	52	93	151	13%	14.089	9.312	0,00	0,00	0,00	0,00	0
92181	Nectar J-7 Peach with soft t/pack (XX) 0.97L	11.683	4.442	52	135	170	18%	13.604	5.041	0,00	0,50	0,41	0,41	0
92182	Juice J-7 Apple Clarified d/d.p. t/pack (XX) 0.97L	30.574	11.327	52	135	385	1%	30.795	11.271	0,00	0,50	0,41	0,41	0
92183	Juice J-7 Orange with pulp. d/d.p. t/pack (XX) 0.97L	26.600	8.862	52	135	389	17%	31.090	10.299	0,00	0,50	0,41	0,41	0
92185	Nectar J-7 Pineapple with soft t/pack (XX) 0.97L	15.773	5.687	52	135	304	54%	24.271	8.716	0,00	0,50	0,41	0,41	0

Promotion Forecasting

Predicting the impact of upcoming Promotions

Promotion Forecasting

Predicting the impact of upcoming Promotions



Promotion Forecasting

Clean data

Typical cleansing steps

1. Detect outliers in sales, inventory, prices, etc.
2. Imputing missing data
3. Out-of-stock detection
4. Product listing/delisting mapping (same SKU but with different ID)
5. Promo extension detection

33337	
09.02.2018	12
23.03.2018	15
15.06.2018	5
06.07.2018	5
17.08.2018	1
28.09.2018	6
19.10.2018	327
30.11.2018	14
18.01.2019	18
08.02.2019	3

33804	
15.06.2018	1
06.07.2018	0
17.08.2018	2
07.09.2018	0
19.10.2018	610
09.11.2018	0
30.11.2018	0
18.01.2019	0
08.02.2019	1

#	Campaign	Code	Page	Key Section
1	201508	23296	2	
2	201508	30717	2	
3	201508	23346	2	
4	201508	28273	3	
5	201508	31830	5	PLT
6	201508	31831	5	PLT

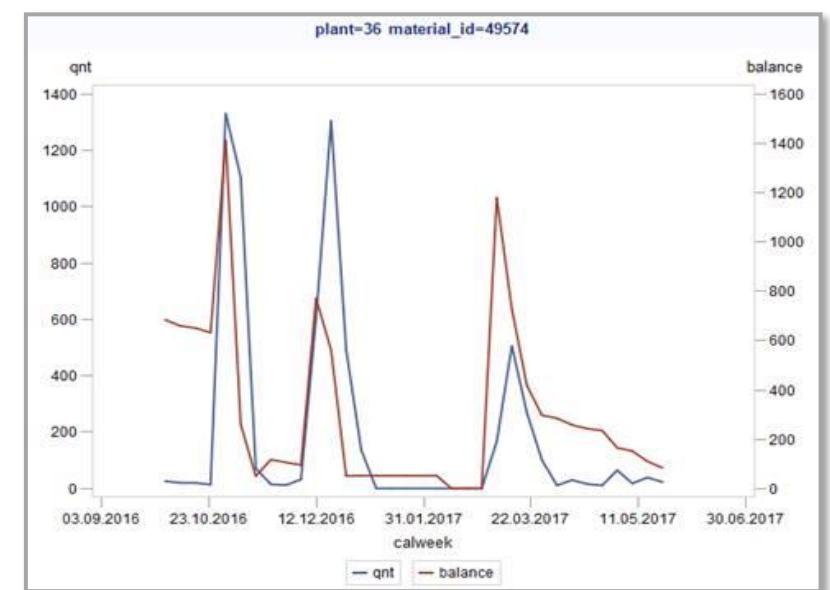
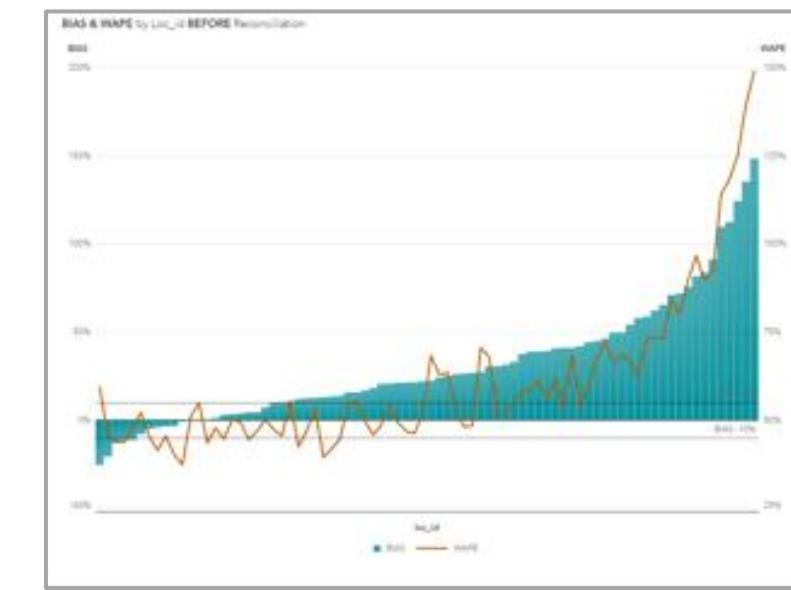
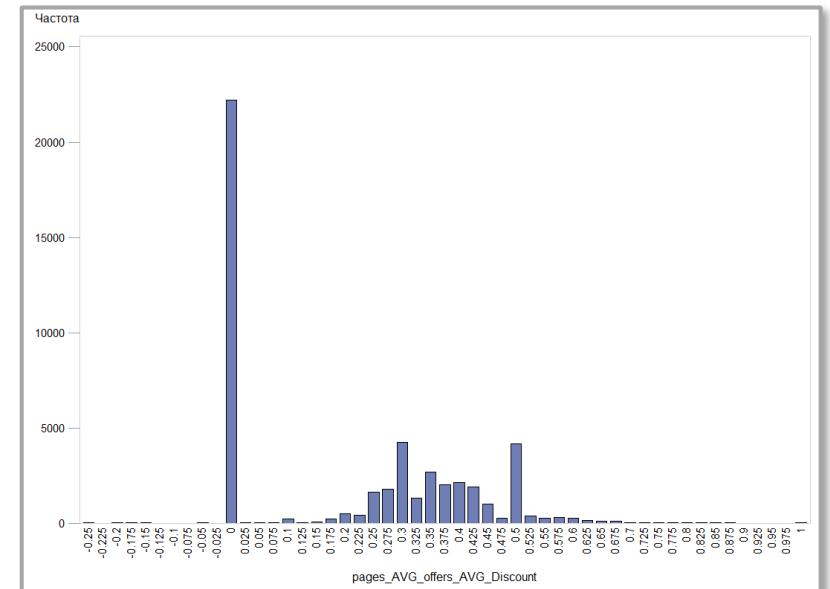
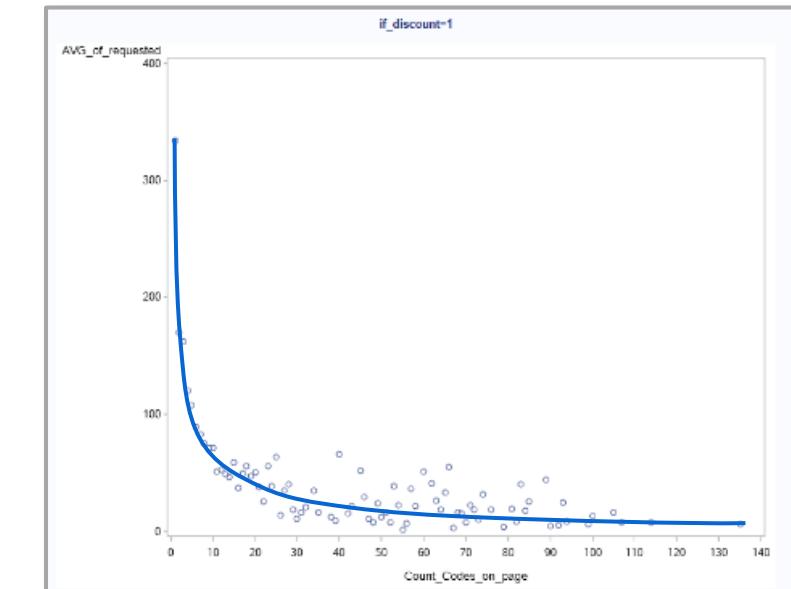
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1	Activelle	377
2	Amber Elixir	124
8	Bioclinic	77
9	Born To Fly	24
10	Colourbox	2094
11	DO NOT USE (EXPIRED) SunMoonEclipse	1

Promotion Forecasting

Explore data

Typical data to explore

1. Category trend-seasonality
2. Products classification high sales/low sales, new/old
3. Insights search
4. Price elasticity estimation
5. Features correlation estimation



Promotion Forecasting

Datamart & Forecasting Models – Statistical & Machine Learning

Basic Datamart:

- Sales lags for “forecasting horizon”, month, quarter, half-a-year, year
- Sales percentiles for “forecasting horizon”, month, quarter, half-a-year, year
- Promo types encoding
- Holidays flags
- Prices, discounts
- Product hierarchy encoding
- Date decomposition (month, DoW, WoY)

Basic training grouping:

- Model per product category
- Basic algorithms ARIMA, LSTM, Random Forest, Gradient Boosting

Promotion Forecasting

Enrich Datamart and tune models

Typical hypothesis for forecast improvement

- Move some promo types / products / categories to separate forecasting models
- Add cross-effects features (max discount in category, price indices, number of products in category, etc.)
- Add information from site (banners size, placeholder type, etc.)
- Add elasticity information
- Reconcile on a lower level (top levels are forecasted with TS, lowest levels are forecasted with ML)



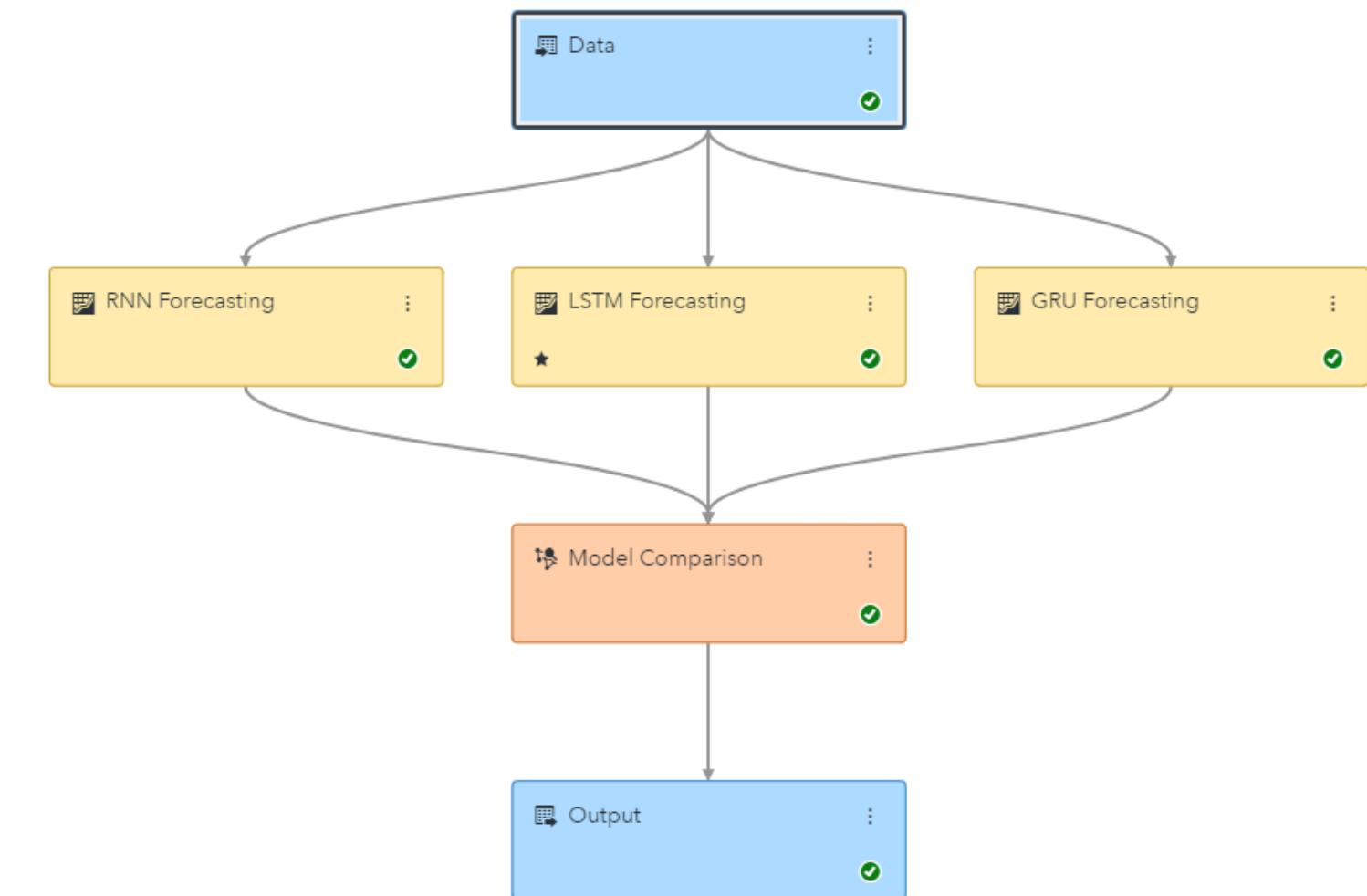
Promotion Forecasting

Enrich Datamart and tune models

Ensemble forecast

- Models stacking
- Choosing the best model for each group

	Naïve	Time series	ML	Ensemble
Promo	37%	41%	56%	58%
Reg	51%	67%	66%	69%



SAS Visual Forecasting

Thank You

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