



CphMRA Meeting

# *AI:EV – Predicting Policy Impact*



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# Policy changes impacting our sales & market dynamics

## Stakeholders from multiple departments

4+7  
Policy

BHP & Market  
volume & prices

NRDL

GPO  
(SH)

EDL

Guotan

+ other factors



# Multiple analysis approaches to analyze policy impact on market

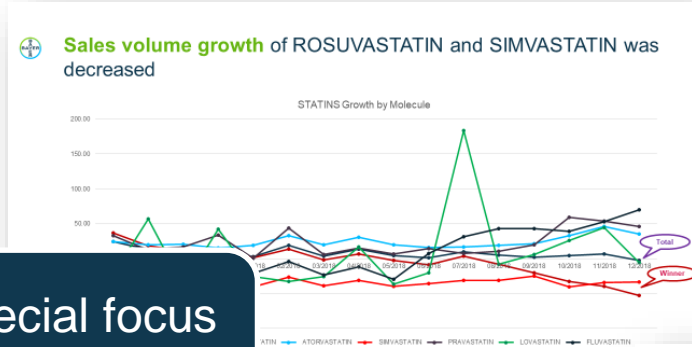


Special focus offline review

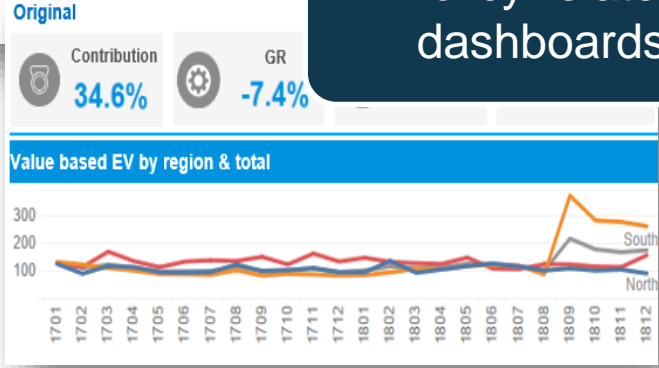
Common dashboards (Market insight, Internal sales)

External sources

Other external sources



Policy related dashboards



... all looking backwards



## AI:EV prestudy – Try to look into the future



Use advanced analytics to **predict** market, key product **sales volume & market share** for upcoming next 1 month, next 6 and 12 months in large hospital market.



\* Prediction based on IQVIA data, large hospital (>100 bed) market

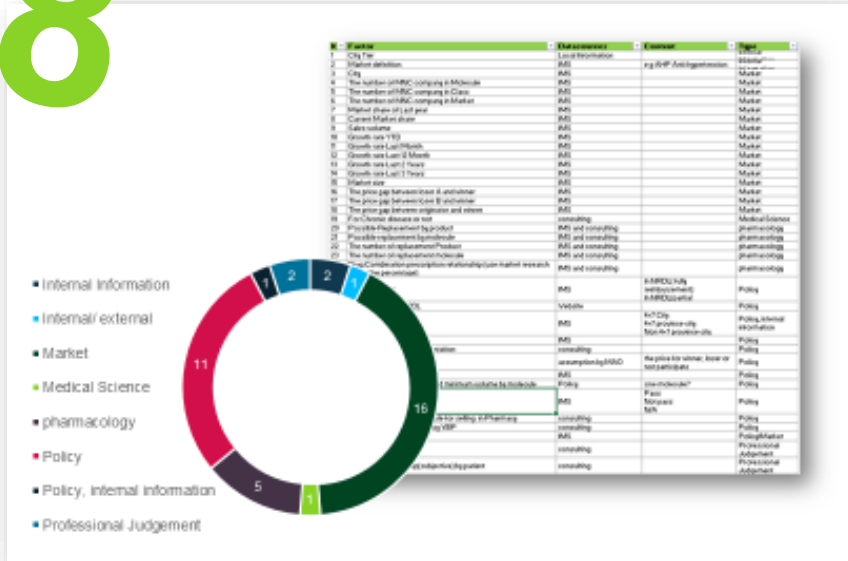


# *Methodology*



# Feature scoping

38 Factors defined by team



600+

Features included

ne	Volume	Volume	Volume	Volume	Volume	Volume	Volume	Volume	Volume	Volume
18720	20160	17280	20160	20160	10656	7776	16416	14400	20160	
18720	20160	17280	20160	20160	10656	7776	16416	14400	20160	
04682	992068	852852	3476	016	1105286	1160964	1340990	1216320	1122730	
04682	992068	852852	3476	016	1105286	1160964	1340990	1216320	1122730	
55610	25	25	130	00	276150	304130	343500	365100	318470	311660
66730	62	66	00	00	276150	304130	343500	365100	318470	311660
66730	62	66	00	00	73170	82210	77610	92910	75850	76860
66730	62	66	604	79600	73170	82210	77610	92910	75850	76860
926	2210	2210	22693	22693	225708	237832	234360	297500	269808	268184
926	2210	2210	22693	22693	225708	237832	234360	297500	269808	268184
927	2210	2210	22693	22693	225708	237832	234360	297500	269808	268184
03656	771946	714854	728147	852075	898562	1104845	1169371	1422673	1315412	1329699
03656	771946	714854	728147	852075	898562	1104845	1169371	1422673	1315412	1329699
94260	399390	396330	368094	377892	393426	392208	397650	431160	398856	370266
94260	399390	396330	368094	377892	393426	392208	397650	431160	398856	370266
0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0
20208	894446	750302	686448	705838	729344	843122	738794	882728	699986	658882
20208	894446	750302	686448	705838	729344	843122	738794	882728	699986	658882
0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0
39580	2146860	2228350	2040710	2198650	2055020	2315410	2293900	2614450	2206080	2055640
39580	2146860	2228350	2040710	2198650	2055020	2315410	2293900	2614450	2206080	2055640
68970	176760	205670	175160	201060	171260	210920	183020	191440	153010	172000
68970	176760	205670	175160	201060	171260	210920	183020	191440	153010	172000
01530	163758	162225	135457	206206	170975	180684	171948	256046	287364	340592



e.g. policy trend, market share, market size, statistical result, sales volume lag, and the product information





# Supervised learning model applied for

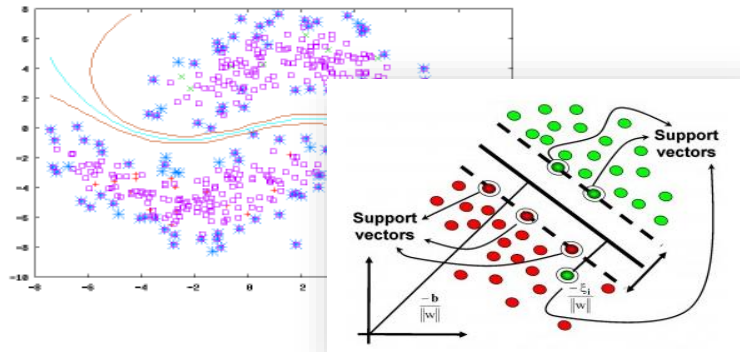


~12,000 pack-city combinations, 300+ companies, 500+ products



41 cities

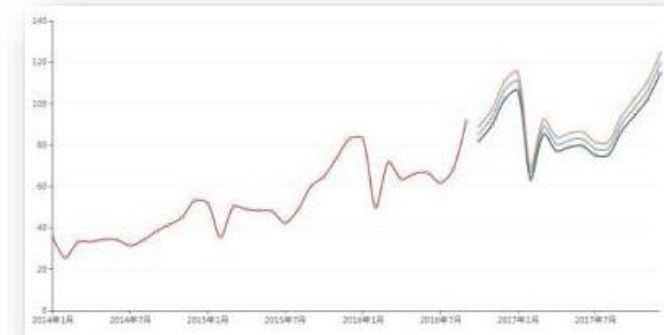
## Classify / Clustering



1. Split test data: key product/ or not, policy city/ not → select a feature set for each groups.
2. Classifier: Train a classifier model to predict will have sales volume or not.

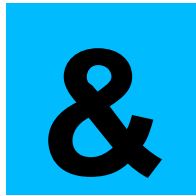


## Predicting



Gradient Boosting Regressor as main model to predict the next six months sales volume.

# Model Amendment – Next month 😊



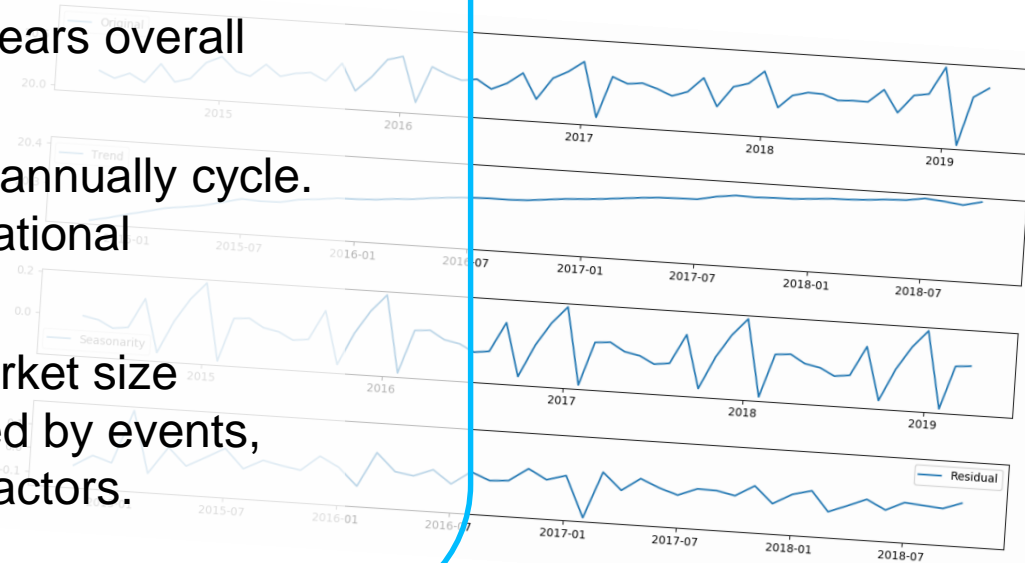
## Market Size Predictor

ARIMA model:

1. **Overall trend:** 5 years overall trend

2. **Seasonality:** The annually cycle. (Spring festival, National festival...)

3. **Residual:** The market size change that caused by events, policies, or other factors.



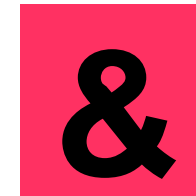
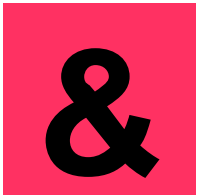
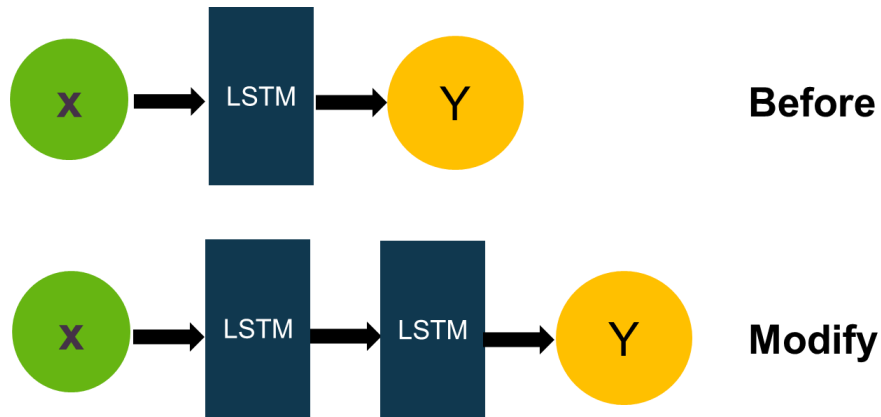




# Model Amendment – 3 months after creation

## Prediction period expansion

Increase level of LSTM:



+ in-mkt.  
features

+ group further  
and train  
specifically

+ group further  
and train  
specifically



# Current feature set



## 6 months prediction

## 6-12 months prediction

volume_last_month_4	MKTNRDLSHORTDESC
volume_last_month_3	MKTGQCESHORTDESC
volume_last_month_2	Type
volume_last_month_1	IsKeyProduct
YTD_Volume_last_month_4	IsKeyCompetitor
YTD_Volume_last_month_3	CityTier
YTD_Volume_last_month_2	Originator
YTD_Volume_last_month_1	winner_flag
is_zero_last_month_4	4+7CityMar2019
is_zero_last_month_3	CompsDesc_company_count
is_zero_last_month_2	CompsDesc_product_count
is_zero_last_month_1	marketshare_last_month_4
MAT_Volume_last_month_1	marketshare_last_month_3
LYRQ_Volume_last_month_1	marketshare_last_month_2
LYFY_Volume_last_month_1	marketshare_last_month_1
LYMAT_Volume_last_month_1	policy_scope
LYYTD_Volume_last_month_1	predict_is_zero

Factors selected by machine

In market sales	LM CM competitor marketshare	In market sales	LM CM competitor marketshare DIFF
volume	LY CM volume	volume	LY CM volume DIFF
month gap to new year	LY CM volume	month gap to new year	LY CM volume DIFF
month gap to national day	LY CM total marketshare	month gap to national day	LY CM total marketshare DIFF
month	LY CM competitor marketshare	month	LY CM competitor marketshare DIFF
marketshare	LYNM volume	marketshare	LYNM volume
LY volume	LYNM marketshare	LY volume	LYNM marketshare
LY marketshare	LYNM LY volume	LY marketshare	LYNM LY volume DIFF
LM CM sales DIFF	LYNM LY competitor marketshare	LM CM sales DIFF	LYNM LY competitor marketshare DIFF
LM CM marketshare DIFF	LYNM LY total marketshare	LM CM marketshare DIFF	LYNM LY competitor marketshare DIFF
LM CM total marketshare DIFF	LYNM LY total marketshare	LM CM total marketshare DIFF	LYNM LY total marketshare DIFF

34

22





# *Results*



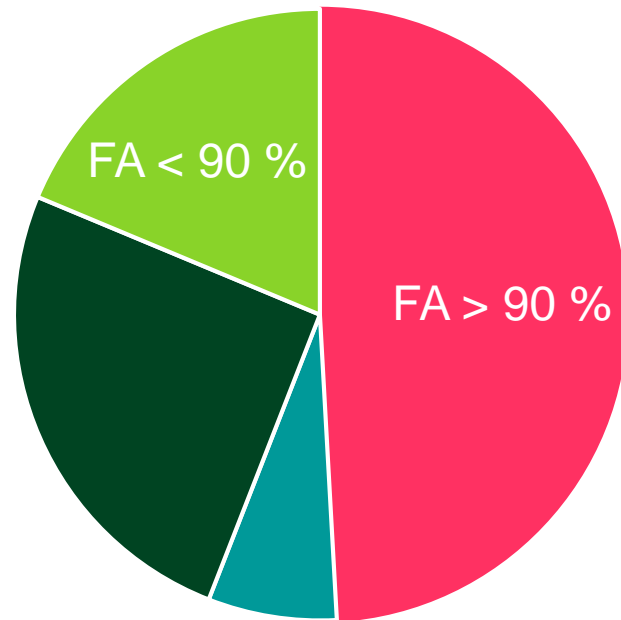
# Pack-level forecast accuracy overview



~12,000 pack-city combinations



41 cities



■ > 90 %   ■ < 90 > 80   ■ < 80% > 50%   ■ < 50 %



# Next months accuracy on selected indicators

Scope	Accuracy August	Accuracy July	Accuracy June	Accuracy May	Accuracy April
BHP EV policy city	99%	90%	98%	99%	91%
MS	98%	94%	97%	95%	92%
BHP MS Shanghai	87%	90%	88%	99%	84%
Market size	100%	99%	99%	97%	95%
Key product market size	97%	99%	92%	92%	95%

Big scope

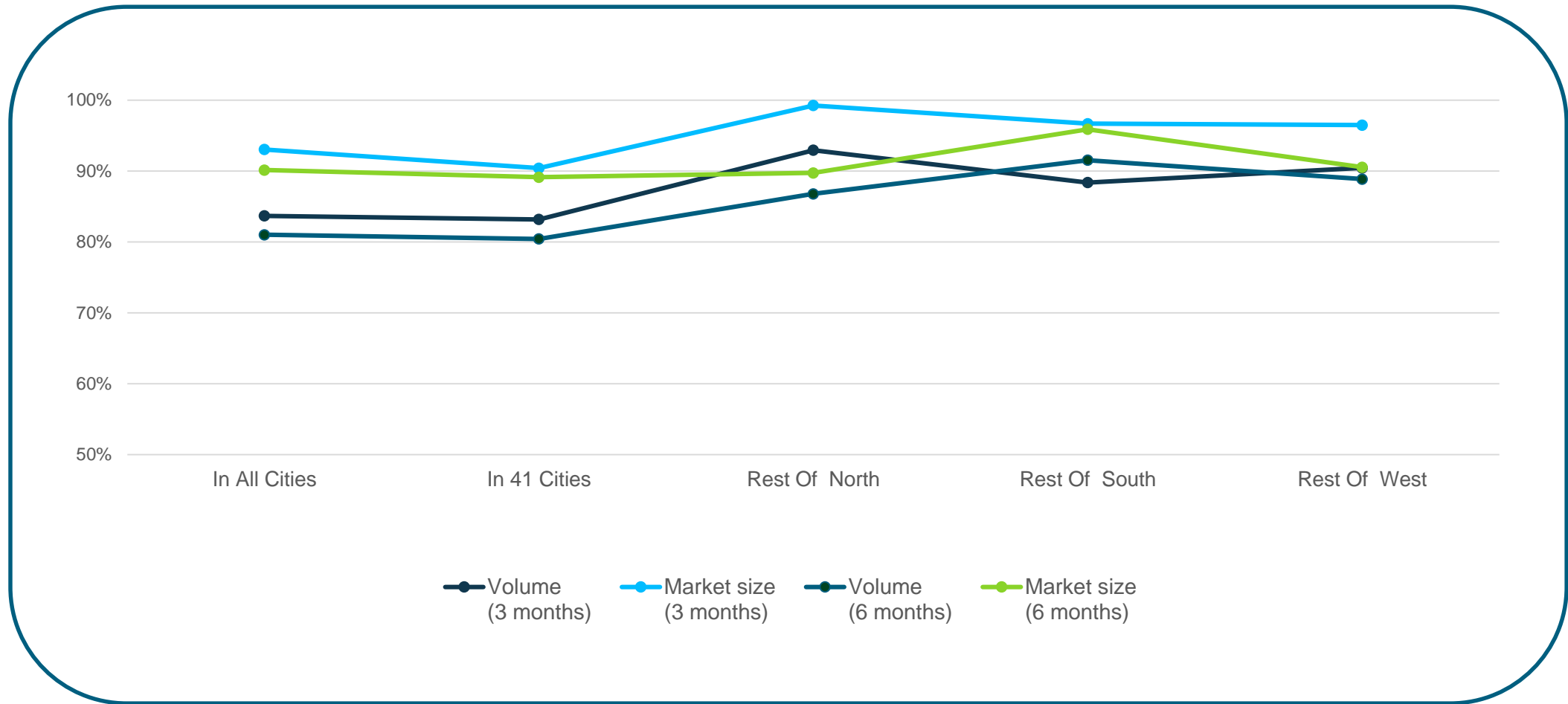
Forecast accuracy declining the smaller the scope

Small scope





# 3 & 6 months product level accuracy (MTH) above 80%



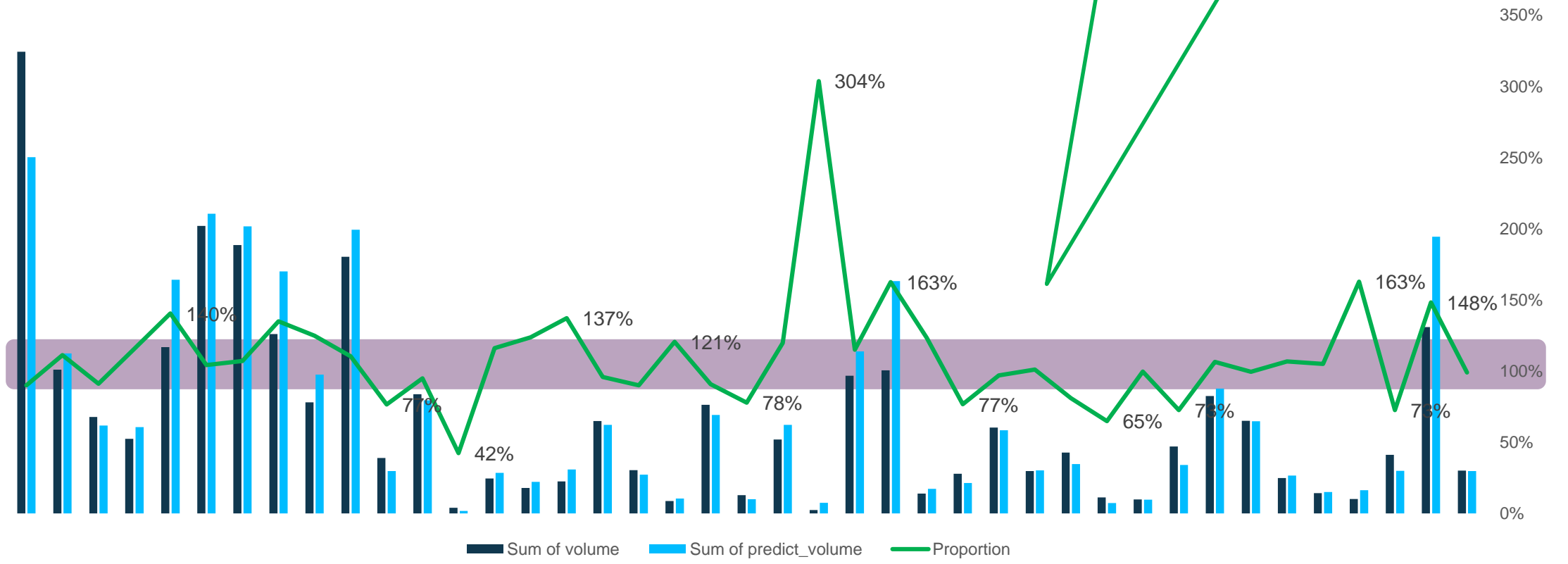


# Predicted proportion of actual

High volatility in monthly data

图表标题

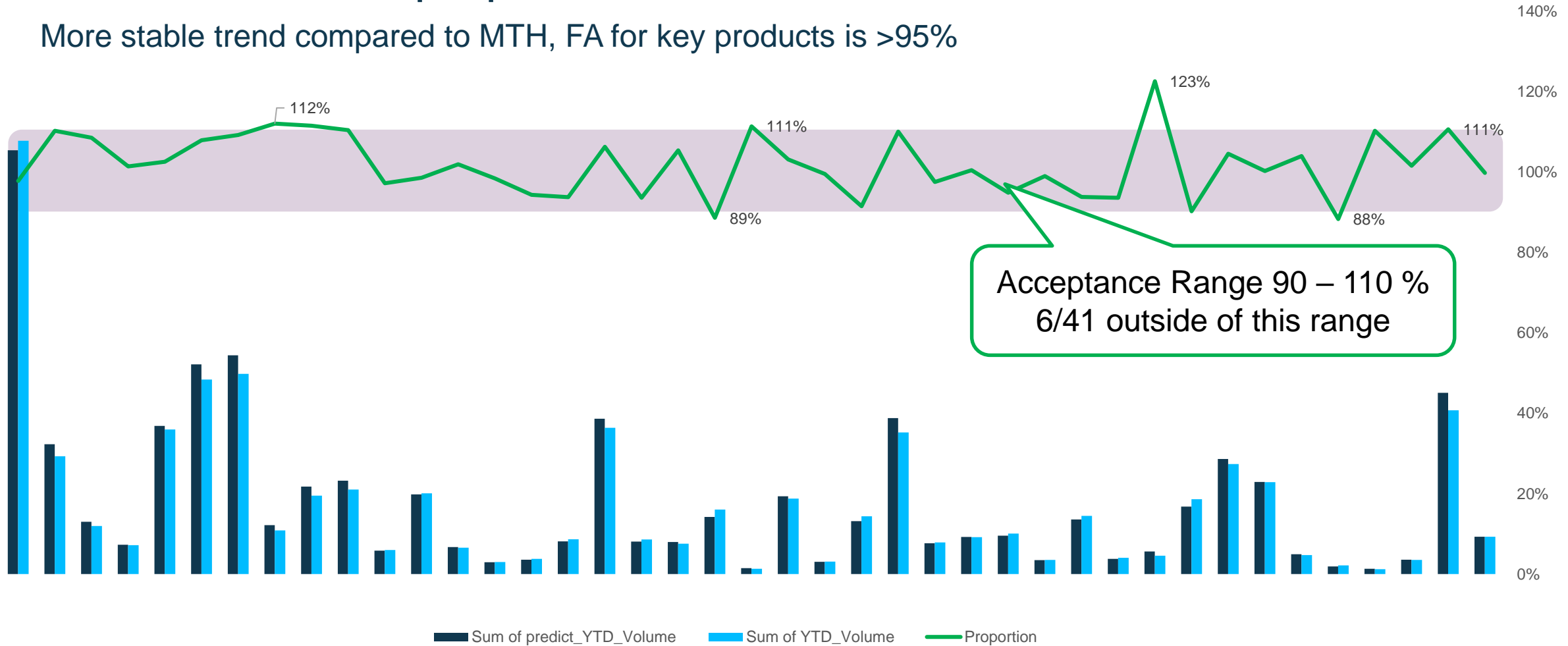
Acceptance Range 85 – 115 %  
14/41 inside of this range





# YTD – Predicted proportion of Actual

More stable trend compared to MTH, FA for key products is >95%







# *Learnings*



# Learnings

Initial policy impact is less than expected

Strong seasonal trends

MS, EV/ EI & Sales Volume are predictable with good accuracy

Low forecast accuracy provides additional insights

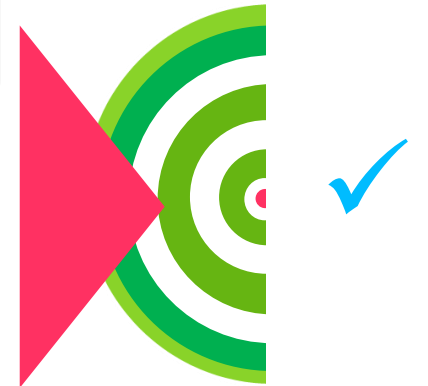
Real-time EV more relevant than further predict values

Potentially re-creatable

Need to gain experience in mid-term forecast accuracy

Regular reporting of MS/ EV future trend

Regular reporting of Outlier brands & cities





谢谢



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