

## PILOT ML FORECASTING

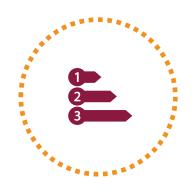
Go/NoGo - January 2022

## **MEETING OBJECTIVES**





- Data
- Architecture
- ML results



Share a first version of the new monthly forecasting process

- Monthly process
- Decision making matrix

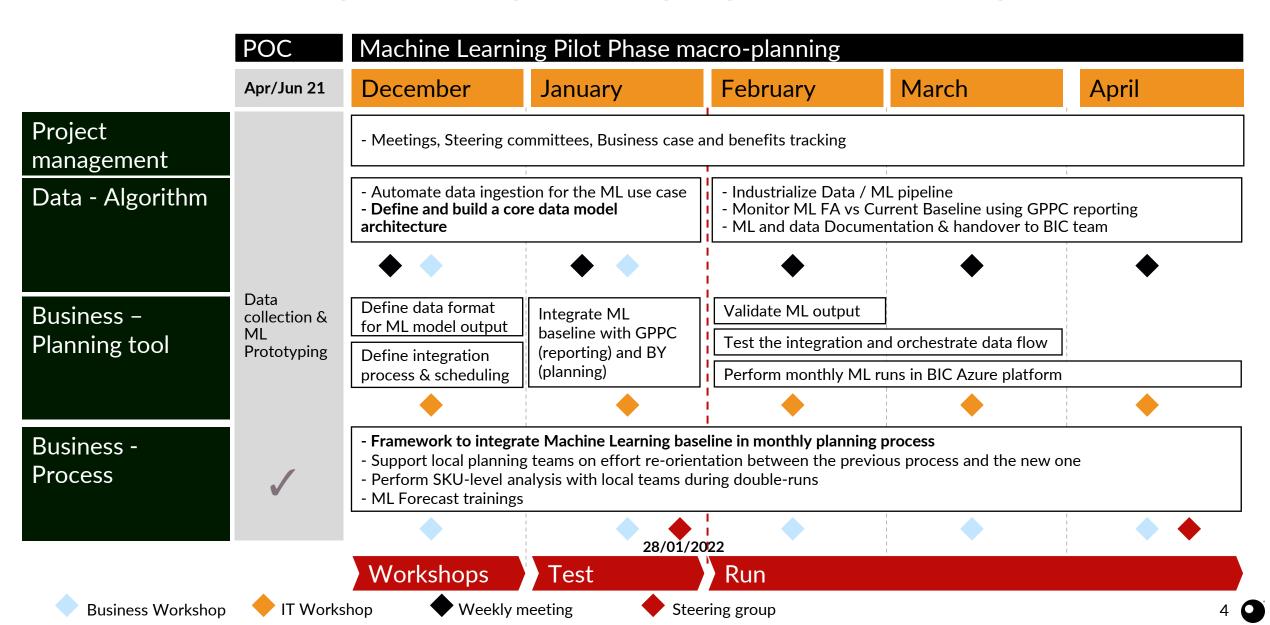


Confirm Go / No Go decision for pilot phase

- Planning
- Project objectives
- Machine learning input data
- Forecasting process with machine learning
- Machine learning results
- Next steps



## PILOT PHASE MACRO-PLANNING



- Planning
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### PILOT PHASE OBJECTIVES



## Data pipeline industrialization

- Define and build a core data model architecture
- Automate data ingestion for the ML use case
- ► Perform automated ML runs each month



### Sell-in Machine Learning Forecast

- Propose a Machine Learning baseline to demand planning teams
- ► Optimize Machine Learning results, using a maximum of performant data sources



## Statistical layer and business alerts definition

- ► Combine a performant ML baseline, with a robust statistical layer
- ➤ Define a "Machine Learning vs Statistical layer decision making matrix"



### **Forecast Process**

- Secure ML forecast adoption by global and local demand planning teams
- Perform training sessions

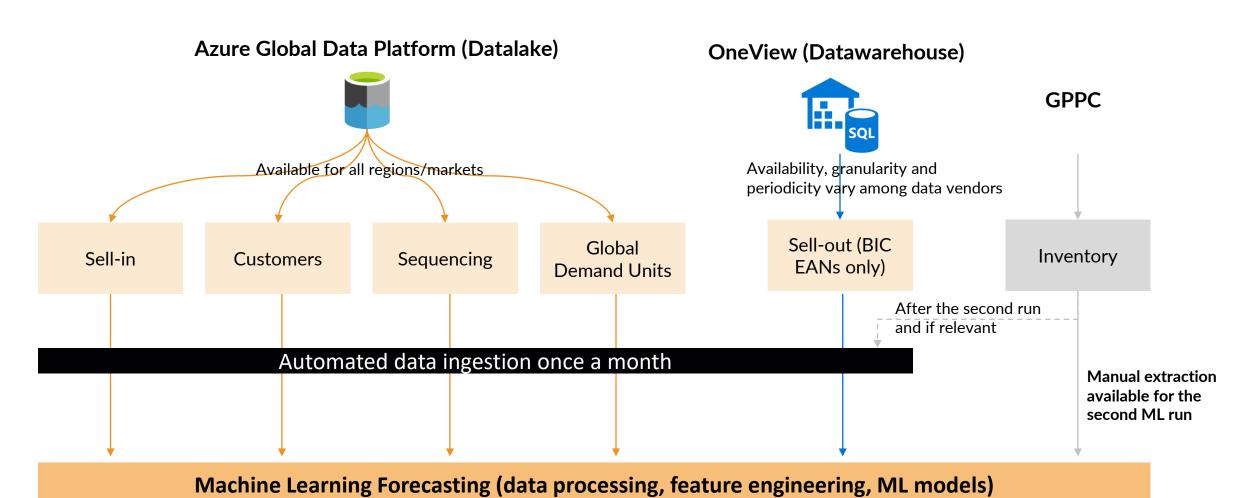


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## GLOBAL ARCHITECTURE FOR ML FORECASTING

Data architecture built within BIC's Azure subscription ready to be used for future roll-outs.



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## PROPOSED FORECASTING PROCESS AS DISCUSSED WITH DEMAND PLANNING TEAM

\*BIC Calendar

Data availability <= 3<sup>rd</sup> working day\* (ex. 09/02/22)

Beginning of month <= 8<sup>th</sup> working day (ex. 16/02/22)

Construction process <= 14<sup>th</sup> working day (ex. 23/02/22)

Collaborative review with Planning & Marketing <= 19th WD (ex. 28/02/22)

End of month
<= last working day of
month (ex. 03/03/22)</pre>

#### **Actors**

Global demand planning or Data teams



#### **Actions**

- Automated data ingestion except for Inventory
- Extract month end inventory data from GPPC

**Automated ML solution** 



- Machine Learning run (18month horizon)
- Automated forecasts file creation (to be integrated with GPPC)
- ► KPI calculation in GPPC

Global planning teams



- ► ML Baseline
- BY statistical baseline
- ► Forecasts modifications in BY

Local planning and Sales & Marketing teams



- "Forecast Revision" baseline to be shared with local Planning and Sales & Marketing teams
- Get insights from local planning and Sales & Marketing to adjust forecast

**Automated ML solution** 



- Validation of forecast baseline
- ► KPI calculation in GPPC

### **Outputs**

- Machine Learning Database
- ► Forecasts file for GPPC
- ► Forecasts baseline with first planning adjustments
- ► Forecasts baseline with validated forecast quantity
- ► Final Forecast in BY (18month horizon) and GPPC

+ Market input

= Final Forecast Performance

## DIFFERENTIATION BASED ON PRODUCT'S LIFE

	Step in product's life	How forecasts are used  High runners (A)  Low runners (B & C)			
		High runners (A)	Low runners (B & C)		
	New products (M0 to M+4)	Human expertise Machine Learning	Human expertise Machine Learning		
Scope	Catalogue products – Stationery	Machine Learning Human challenge	Machine Learning		
	Catalogue products – Shavers	<b>Machine Learning</b> Human challenge	Machine Learning		
	Catalogue products – Lighters	<b>Machine Learning</b> Human challenge	Machine Learning		
	Catalogue products – Other Products	<b>Machine Learning</b> Human challenge	Machine Learning		
	Promotions	Human expertise for promotions impact Machine Learning (with volume effect decomposition)			
	One Shot	Human expertise	Human expertise		
	End-of-life (M-3 to EOL)	Statistical rules Human challenge	Statistical rules Human challenge		

During the second phase, a detailed analysis carried out with planners at the product level will allow to adapt the forecasting method to use

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## **RESULTS SUMMARY FOR FRANCE & ITALY**

	FA 2019	POC FA 2019 ML Baseline	Pilot FA 2019 ML Baseline	FA 2020	POC FA 2020 ML Baseline	Pilot FA 2020 ML Baseline
France – Stationery <sup>(1)</sup>	65,7%	66,5% (+ <b>0,8</b> pts)	67,1% (+ <b>1,4</b> pts)	64,2%	66,5% (+1,3pts)	66,9% (+1,7pts)
France – Shavers <sup>(1)</sup>	71,6%		71,9% (+ <b>0,3</b> pts)	68,2%		71,0% (+ <b>2,8 pts</b> )
France - Lighter <sup>(1)</sup>	72,3%		73,3% (+ 1 pts)	60,8%		58,0% (- <b>2,8</b> pts)
Italy – Stationery <sup>(1)</sup>	52,9%	59,3% (+ <b>6,4</b> pts)	59,3% (+ <b>6,4</b> pts)	21,4%	38,1% (+ <b>16,7</b> pts)	40,6% (+ 19,2 pts)
Italy – Shavers <sup>(1)</sup>	72,0%	72,6% (+ <b>0,6</b> pts)	72,6% (+ <b>0,6</b> pts)	67,3%	69,7% (+ <b>2,4</b> pts)	70,6% (+ <b>3,3 pts</b> )
Italy - Lighter <sup>(1)</sup>	66,8%	59,0% (- <b>7,8 pts</b> )	62,8% (- <b>4 pts</b> )	21,2%	42,8% (+ <b>21,6</b> pts)	42,8% (+ <b>21,6</b> pts)
Italy- Other <sup>(1)</sup>	55,4%	59,6% (+ <b>4,2</b> pts)	60,8% (+ <b>5,4</b> pts)	56,0%	63,1% (+ <b>7,1</b> pts)	64,6% (+ <b>8,6</b> pts)
Italy – Mixed <sup>(1)</sup>	66,2%	68,1% (+ <b>1,9</b> pts)	69,2% (+ <b>3</b> pts)	54,3%	68,5% (+ <b>14,2</b> pts)	72,8% (+ <b>18,5</b> pts)

Forecast Accuracy comparison between BIC's final Forecast and Machine Learning output

## PILOT PHASE KPI MEASUREMENT

### Technology stack / Data Architecture

 Qualitative feedback on how well the architecture's components worked



Measurement of an adherence level to ML forecasts which would be used as-is or

overridden

► KPI to be defined. For example : 100% when ML forecasts are used as-is without overriding values

### **Global planning teams feedback**

▶ Qualitative feedback from global planning team. For example, on the course of forecasts review sessions, on how much ML is adding value to the business ...

► Performance measurement of machine learning models starting from April (2 months horizon) for February's ML run

# THANK YOU!