Consultant Interventions and Behavioral Change in Supply Chain Forecasting

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Judgmental adjustments in forecasting



- Actuals (X).
- Model Forecast (MF) or Statistical Forecast or System Forecast.
- **D** Expert Forecast (EF) or Judgmental Adjustment. This is usually used as the Final Forecast (FF).

Lessons learned from the literature

- Model-based forecasts for SKU-level data are **frequently** adjusted by experts. (Franses and Legerstee, 2009)
- Experts adjust model-based forecasts **upwards** more than downwards. (Franses and Legerstee, 2009)
- **Positive** adjustments are **far less effective** than negative **ONES.** (Fildes et al., 2009)
- The **optimism bias** leads to positive adjustments having larger errors than negative ones. (Trapero et al., 2013)
- Expert forecasts often **significantly differ** from model forecasts. (Franses and Legerstee, 2010)
- However small adjustments are also the case: ownership. (Fildes et al., 2009; Gonul et al., 2009)

Lessons learned from the literature (cont'd)

- Experts can reduce forecasting error when adjustments size is **not too large**. (Trapero et al., 2013)
- Where the forecasters' principal **motivation** is towards improved accuracy, they can add substantially to forecast accuracy. (Fildes et al., 2009)
- Small improvements to statistical forecasts can translate to significant gains in terms of utility. (Syntetos et al., 2009)
- Big losses in judgmental adjustments are most probable to be followed by **another big loss** as a result of a large adjustment. (Petropoulos et al., 2016)
- Combination leads to improvements. (Fildes et al., 2009; Franses and Legerstee, 2011)

Adjusting the adjustments

• Franses & Legerstee (2011, ESwA):

$$FF_t = \alpha_t EF_t + (1 - \alpha_t)MF_t \Leftrightarrow$$

$$FF_t = MF_t + \alpha_t (EF_t - MF_t) \Leftrightarrow$$

$$FF_t = MF_t + \alpha_t FD_t$$

- **Damping** the judgmental adjustments!
- Blattberg-Hoch approach (50% model + 50% manager; 1990, MnSc).
- Petropoulos et al. (2016, EJOR): will this lead to a change in forecasters' behaviour with regards to how they perform judgmental interventions?

A laboratory experiment

- The participants were asked to judgmentally adjust the one-stepahead statistical forecasts for **30 time series**.
- For each time series, **36 months of history** are available.
- In the past 36 months there has been no special events (promotional activity or competitors' activity) and as such there has been no need for any judgmental adjustment.
- In the immediate next period there is reliable information that: either

Our company is going on a heavy promotion of a Buy-One-Get-One-Free type

or

Our main competitor is going on a heavy promotion of a Buy-One-Get-One-Free type

• Participant were given information with regards to who is running the promotion for each of the time series.

A laboratory experiment (cont'd)

- A computer forecast prepared from an Exponential Smoothing based algorithm has also been prepared for the next period - this is a time series forecast and is not capturing the forthcoming special event.
- Participants were asked to make their own judgmental forecast (by adjusting the statistical forecast) so as to take into consideration the additional promotional information.
- Two tasks:

"The **final forecast** that is going to be used for the decision to be taken (e.g. ordering, replenishment) will be...

- ...solely **your forecast** (judgmentally-adjusted forecast)."
- ...the simple average (50-50%) of your forecast (judgmentallyadjusted forecast) and the computer forecast (statistical forecast)."

Positive versus negative adjustments



Low versus high noise series

Judgmentally adjust the forecast Task 1, Series 4 out of 15 In this task, your judgmentally-adjusted forecast will be the final (operational) 1500 Historical data forecast. ---- Statistical forecast ---- Judgmentally-adjusted forecast 1400 In the next period, a promotion will be applied by 1300 our company 1200 Sales Enter your percentage (%) adjustment: 35 1100 Submit 1000 006 Judgmentally adjust the forecast 2014.0 2014.5 2015.0 2015.5 2016.0 2016.5 2017.0 1400 Time Historical data ____ ----Statistical forecast Judgmentally-adjusted forecast 1200



Measuring behavioral change

- A representative uplift for promotions was not provided: **avoiding anchoring**.
- Each participant may **interpret** the effect of the promotions **differently**.
- Behavioral change is measured per participant.
- We measure:

 MAA_1 : mean absolute adjustment for task 1, where FF = EF

 MAA_2 : mean absolute adjustment for task 2, where $FF = 0.5 \cdot (MF + EF)$

• Behavioral change (%):

$$100 \cdot \left(\frac{MAA_2}{MMA_1} - 1\right)$$

Evidence on behavioral change



The effect of direction of adjustment



The effect of noise in series



Key insights

- Forecasters' behavior does change, once they are told that their judgmentally adjusted forecast is further adjusted.
- The **degree of change** does not directly reflect to the degree of the consequent adjustment.
- Also, there exist great differences across the behavior of different forecasters.
- Forecasters' size of interventions is affected by
 - The direction of the adjustment.
 - $_{\circ}$ The degree of noise in data.
- However, neither seem to interact with the behavioral change associated with adjusting the adjustment.

Future work

- A second experiment... Two tasks:
 - Task 1, as before: the final forecast will be the judgmentallyadjusted forecast
 - Task 2: provide evidence on the performance of 50-50% combination, but still make explicit that the final forecast will be the judgmentally-adjusted forecast.
- We expect (hopefully!) that participants will change their behavior in Task 2, but to the **opposite direction**.
- Possible solution (?): Provide feedback with regards the efficiency of combined system + expert forecast.
- **Case study**: multibillion North American company in the consumer packaged goods industry.

Questions?

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