

Forecasting digital interactions

Philip Stubbs, Atlantic Insight Thursday 14 October 2021



The importance of data accuracy

#1 Make sure your historical data is accurate

#2 Understand in detail the sources of digital volume data

#3 Automate the assembly of digital volumes

#4 Understand when demand was constrained







#5 Data Preparation – fix data before passing it into a model

Understand the historical variation in digital volumes



#8 Keep a diary of reasons for variation in digital volumes

#6 Understand the variation in historical volumes

#9 Use terms Trend and Seasonality correctly



Trend is a long term increase or decrease in the data

Seasonality is where a time series is affected by seasonal factors such as time of year or day of the week



Understand future impacts to digital volumes

#10 Identify individuals whose actions contribute to variation

#11 Get accurate information about digital volume drivers with enough time to change the resource plan

#12 Challenge the accuracy and timeliness of drivers



#13 Keep very close to website changes & changes to customer communications

#14 Be conservative with respect to promises of a new system and/or process



Forecasting process



#15 Judgement from stakeholders can improve accuracy

#16 Study errors & share with clients and suppliers

#17 Graph forecasts next to recent actuals



Other forecasting points



#18 Channel shift will change the total number of enquiries

#19 Using scenarios more important when uncertainty greater

#20 Automate forecasting when conditions are right

- Automated forecasts are more accurate analysis + judgement
- Errors on these digital channels don't have CX / commercial impact
- Forecast analyst support is limited



Time series models

#21 Understand Time Series models

Only look at historical values of what we are trying to forecast – no external variable involved

- Simple averages
- Weighted averages
- Bespoke model
- Exponential Smoothing, eg Holt Winters
- ARIMA





Causal models



#23 Make sure three assumptions are in place before you use a Causal model:

- 1. A strong relationship exists with the driver
- 2. Expect the relationship will continue to exist in the future
- 3. Reliable forecasts of the driver exist

#24 Time Series models often outperform Causal models

#25 Dig a little deeper to find a causal model



Initial Regression Model

This Weeks Emails = 834 + 0.134 * This Weeks Orders

Lagged Regression Model

This Weeks Emails = 892 +

0.079 * Last Weeks Orders + 0.053 * Week -3 Orders



Final points



#26 Select model based on its forecasting ability, and not on model fit

#27 Try Machine Learning models

#28 Look into SAS, R or Python

#29 Further Reading (buy the latest editions)









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