





Quantitative Business Forecasting

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Suhartono

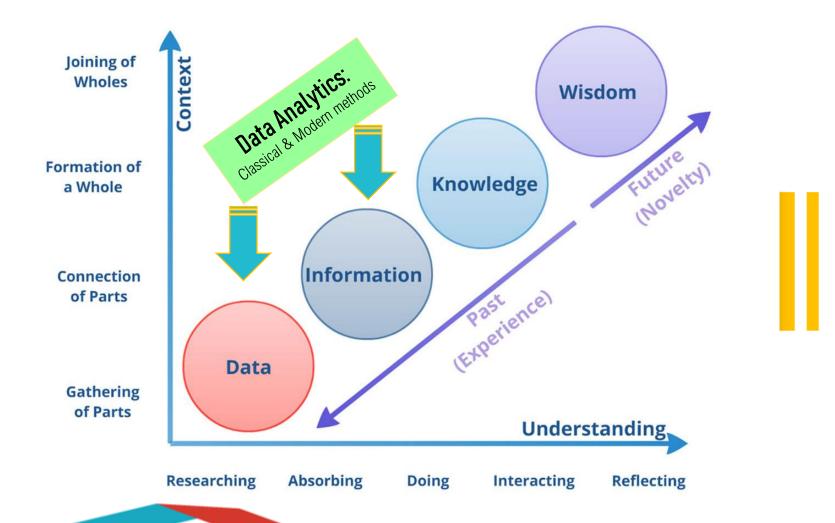
Laboratorium Analitika Data Ekonomi dan Finansial Departemen Statistika

Institut Teknologi Sepuluh Nopember



Statistical Thinking



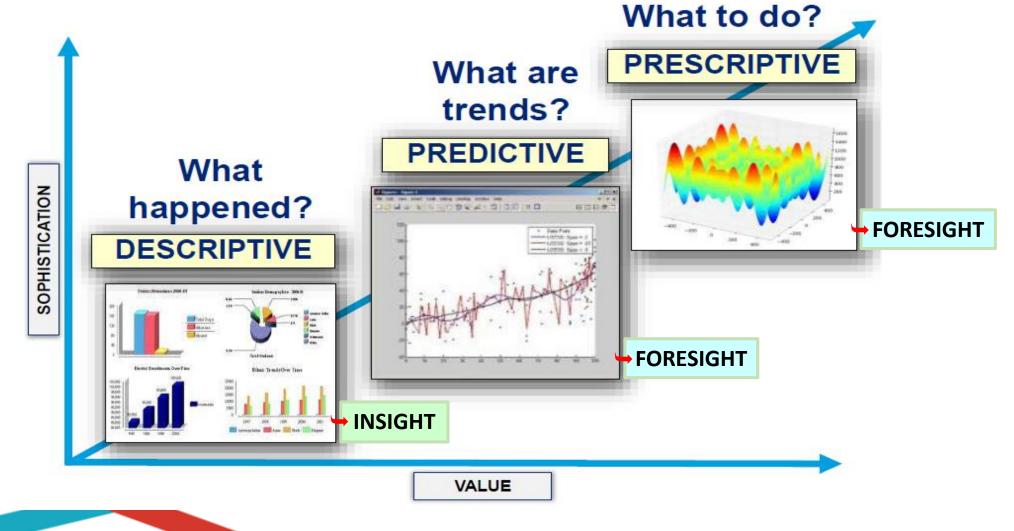






Data Analytics



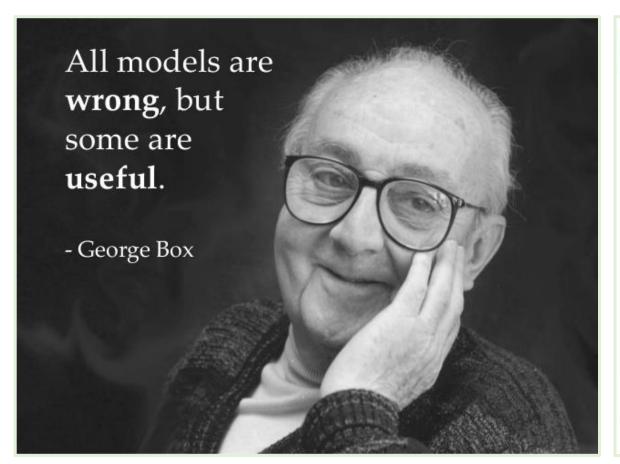


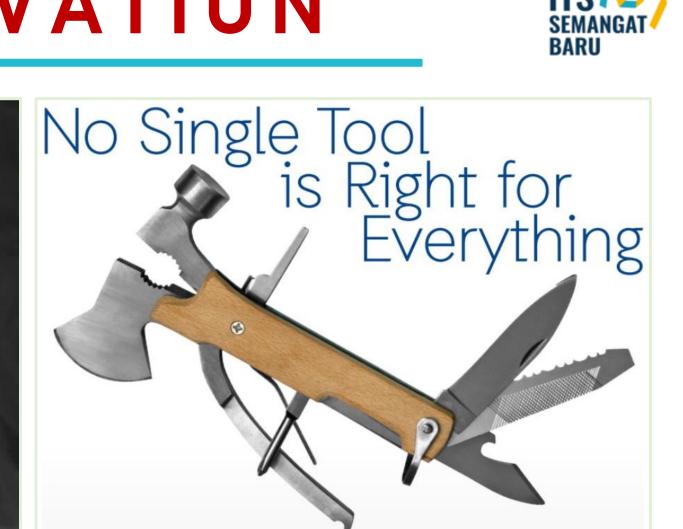
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MOTIVATION



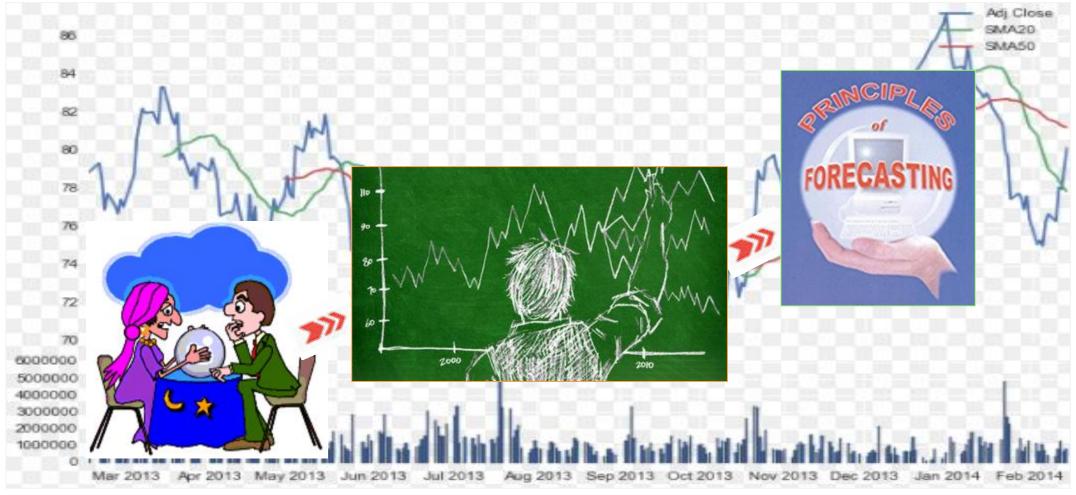






MOTIVATION

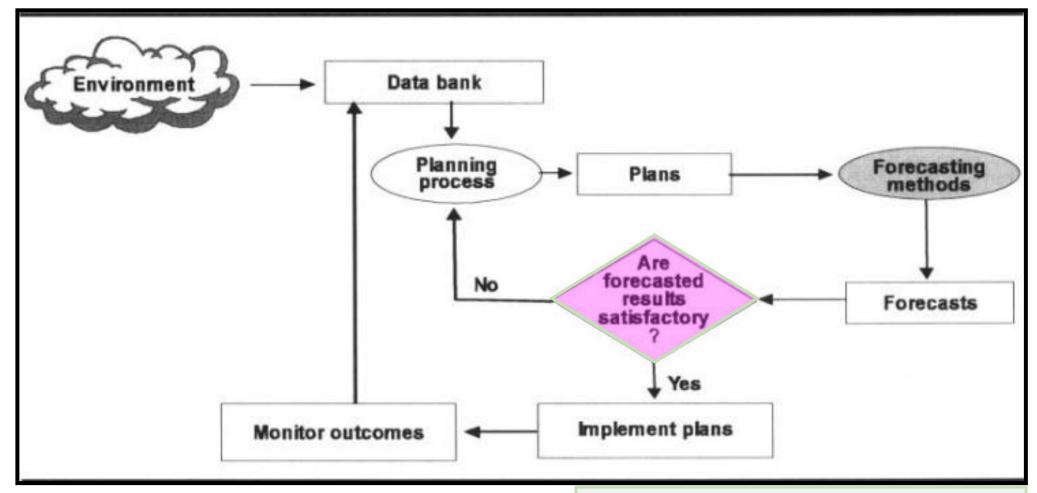






Forecasting & Planning





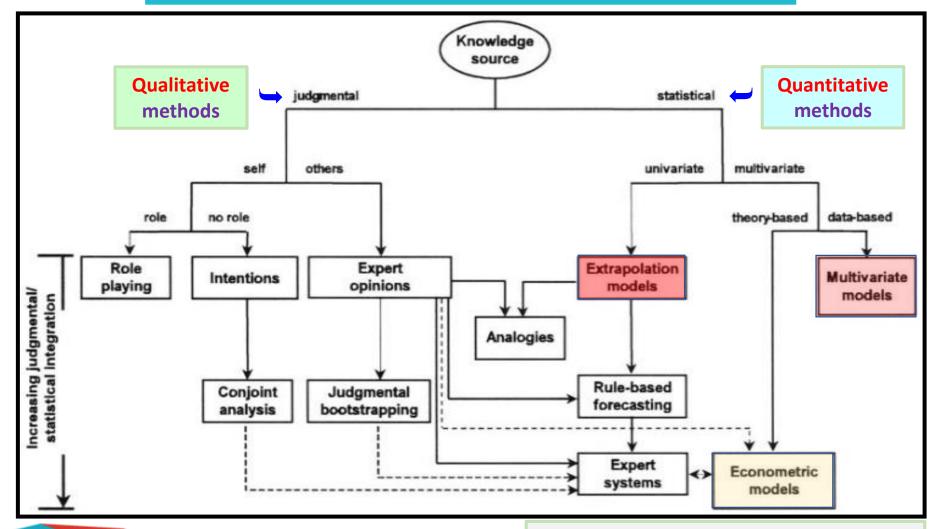
Source: J.S. Armstrong, "Principles of Forecasting"

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Forecasting Methods





Source: J.S. Armstrong, "Principles of Forecasting"



MOTIVATION

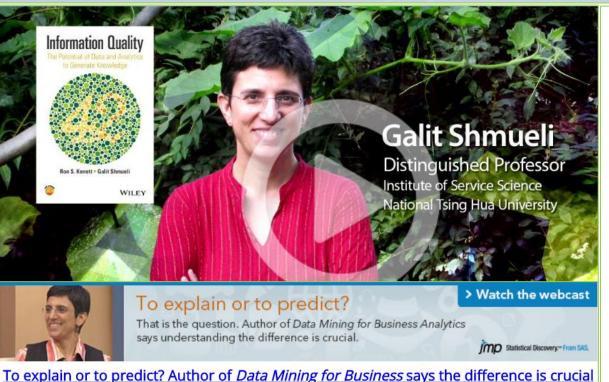


ANALYTICALLY SPEAKING

To Explain or Predict? That Is the Question

With Galit Shmueli

http://www.galitshmueli.com/content/explain-or-predict



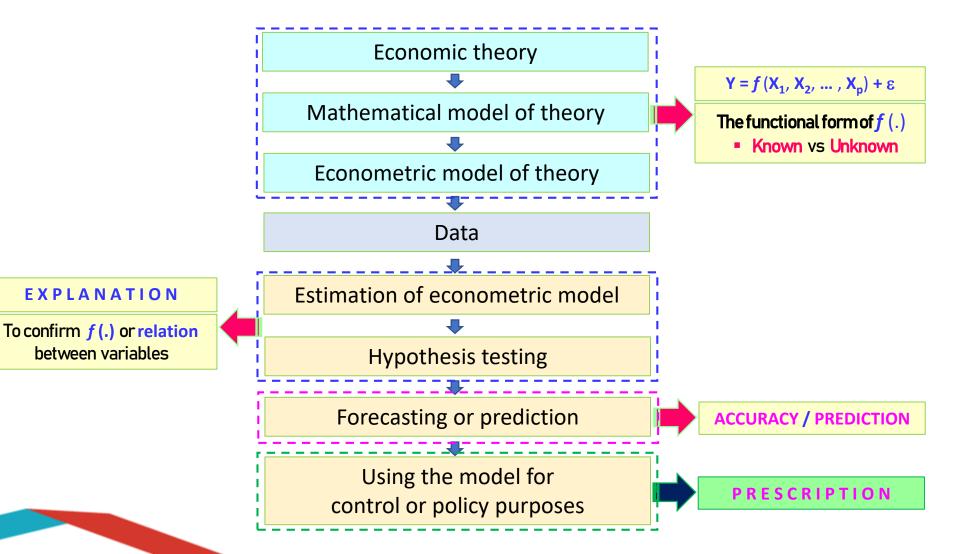
In her highly acclaimed paper,
To Explain or to Predict?, Galit
Shmueli writes "statistical
modeling is a powerful tool for
developing and testing
theories by way of causal
explanation, prediction, and
description." But while it is
common to conflate
explanation and prediction,
understanding the distinction
is crucial.



Econometrics Modeling

Source: Gujarati (2004)

between variables





Qualitative

methods

Quantitative methods

Selecting Forecasting Methods



Sales-forecasting methods used by firms

	Regularly used (percentage)		Regularly used (percentage)
Expert opinions		Extrapolation	
Internal		Naïve	30.6
Sales force	44.8	Moving average	20.9
Executives	37.3	Rate of change (percentage)	19.4
External		Rate of change (units)	15.7
Industry survey	14.9	Exponential smoothing	11.2
Analogies		Regression against time	6.0
Leading indicators	18.7	Box-Jenkins	3.7
Econometric			
Multiple regression	12.7		
Econometric methods	11.9		

Quantitative methods

Source: J.S. Armstrong, "Principles of Forecasting"



Quantitative Forecasting Methods



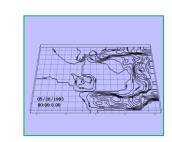
Approach:

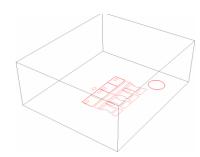
Causal Approach

$$Y_{t} = f(X_{1t}, X_{2t}, ..., X_{6t}) + \varepsilon_{t}$$

Time Series Approach

$$Y_{t} = f(Y_{t-1}, Y_{t-2}, ..., Y_{t-k}) + \varepsilon_{t}$$





- Variable: Univariate vs Multivariate
- Technique: Classical Models vs Machine Learning

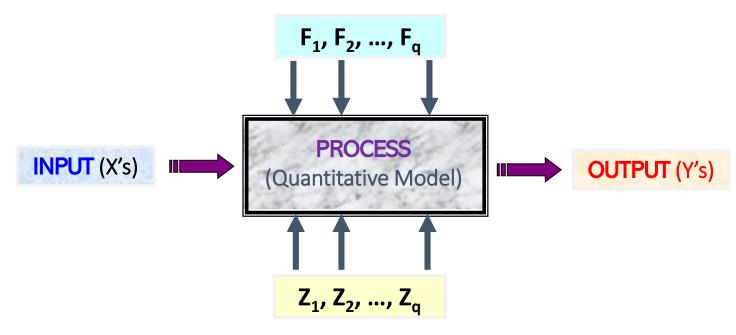


Data Analytics for Sales Forecasting



Controllable Factors

Different
regions have
different
demography,
social
structures,
and cultures.



The number of sales in a certain company.

Uncontrollable Factors

Adalah mudah untuk membuat prediksi dengan metode kuantitatif, tetapi tidak mudah untuk bisa membuat eksplanasi yang valid dan reliable tentang angka-angka prediksi itu, khususnya tentang jaminan asumsi/skenario yang digunakan adalah benar dimasa yang akan datang.



Quantitative Forecasting Methods



Approach	Variable	Technique	Methods
Causal Univariate Support Vector Machine (SVM)	Univariate	Classical models	Multiple Regression, MARS,
		Machine Learning	NN, SVR, ANFIS, Deep Learning,
	Hybrid	Regression & NN,	
	Multivariate	Classical models	Multivariate Linear Regression (MLR),
		Machine Learning	Multi output NN, Deep Learning,
		Hybrid	MLR & Multi output NN,
Time Series	Time Series Univariate	Classical models	Time Series Regression (TSR), ARIMA,
Deep Feed Forward (DFF)		Machine Learning	NN, SVR, ANFIS, Deep Learning,
		Hybrid	ARIMA & NN, ARIMA & Deep Learning,
	Multivariate	Classical models	VARIMA, GSTAR,
		Machine Learning	Multi output NN, Deep Learning,
		Hybrid	VARIMA & NN, GSTAR & NN,

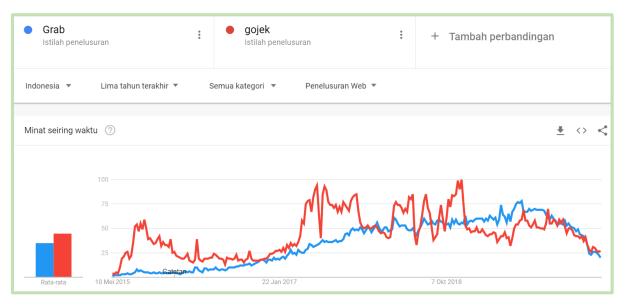


Time Series PATTERN



General time series "pattern"

- Stationer
- Trend
- 🖎 Seasonal
- Cyclic
- Calendar Variation



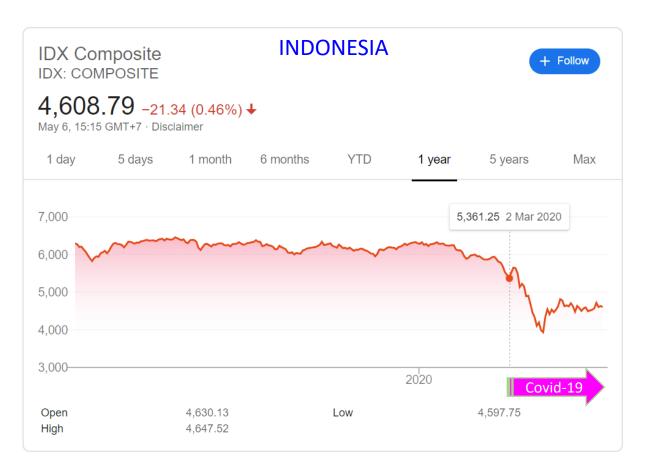
Source: https://trends.google.com/trends/?geo=US

→14



Trend & Cyclic Pattern









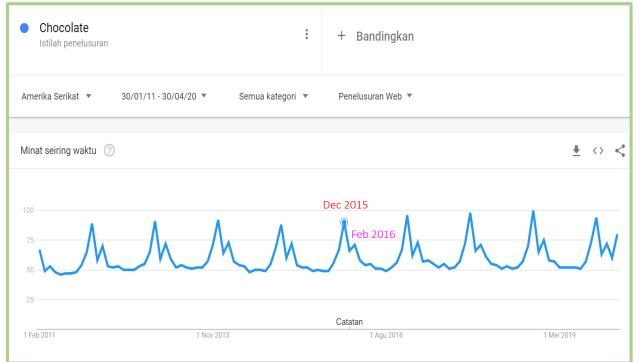
Seasonal & Calendar Variation Pattern



INDONESIA

Cokelat + Bandingkan Makanan manis Indonesia ▼ 30/01/11 - 30/04/20 ▼ Semua kategori 🔻 Penelusuran Web ▼ Minat seiring waktu (?) **±** ↔ < Catatan 1 Agu 2016 1 Mei 2019 1 Nov 2013

UNITED STATES



Seasonal & Calendar Variation Pattern

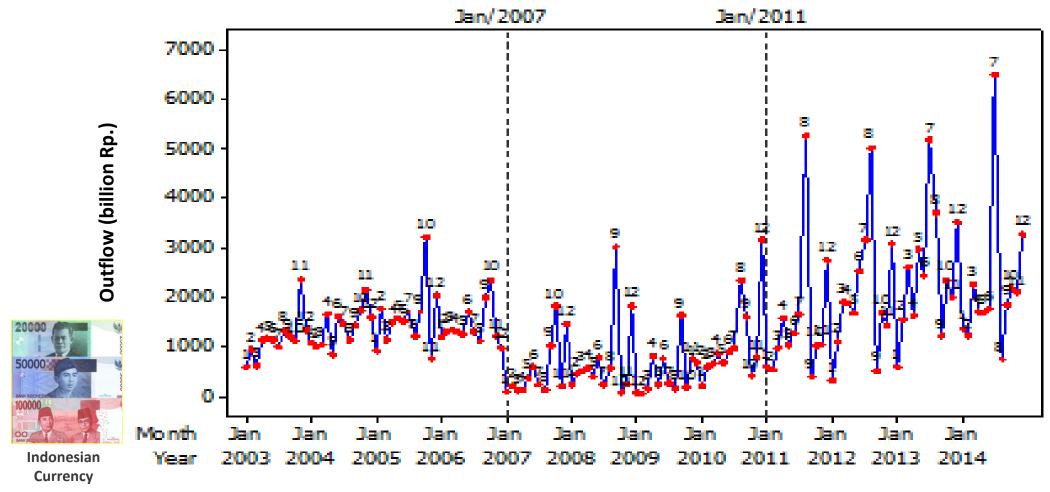
Seasonal Pattern

Source: https://trends.google.com/trends/?geo=US



Outflow Data in Indonesia





Currency

20000

50000

Indonesian

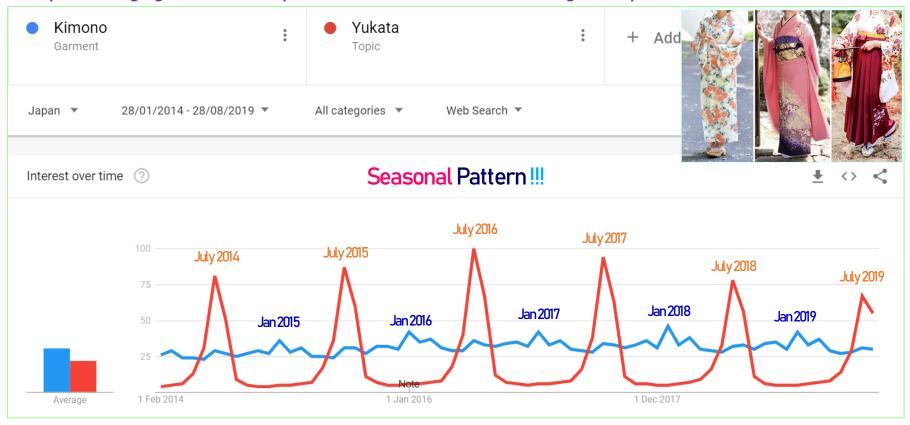
Seasonal & Calendar Variation Pattern



Seasonal or Calendar Variation Pattern



https://trends.google.com/trends/explore?date=2014-01-28%202019-08-28&geo=JP&q=%2Fn1%2F049kc,%2Fn1%2F01332v



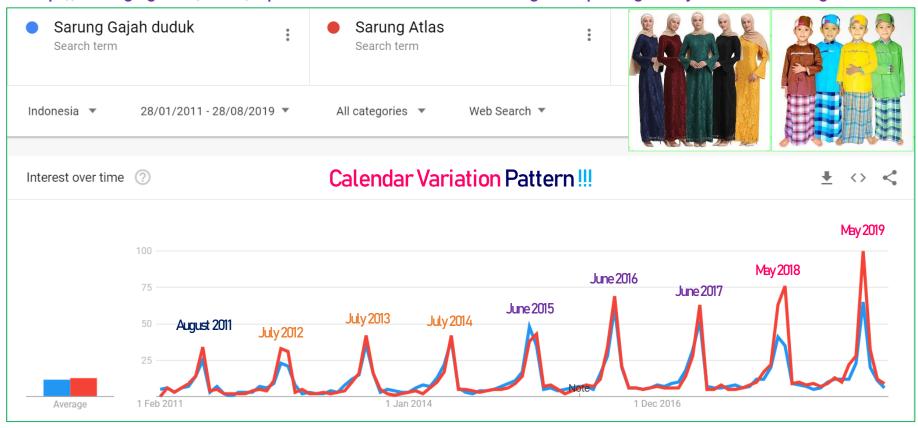
The Gregorian calendar is a solar calendar system.



Seasonal or Calendar Variation Pattern



https://trends.google.com/trends/explore?date=2011-01-28%202019-08-28&geo=ID&q=Sarung%20Gajah%20duduk,Sarung%20Atlas



The Lunar calendar is a calendar based upon the monthly cycles of the Moon's phases.



25 years of time series forecasting De Gooijer & Hyndman (International Journal of Forecasting, 2006)



□ 1 25 years of time series forecasting		
Introduction		
Exponential smoothing		
Preamble		
─ I Variations		
State space models		
Method selection		
Robustness		
Prediction intervals		
Parameter space and model properties		
ARIMA models		
- Preamble		
- Univariate		
Transfer function		
- Multivariate		
- Seasonality		
State space and structural models and the Kalman filter		

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□ I Nonlinear models ✓		
Preamble		
Regime-switching models		
Functional-coefficient model		
- Neural nets ✓		
Deterministic versus stochastic dynamics		
Miscellaneous		
Long memory models		
ARCH/GARCH models		
Count data forecasting		
Forecast evaluation and accuracy measures		
- Combining ✓		
Prediction intervals and densities		
A look to the future Spatial-Temporal		
Acknowledgments ❖ Linear-Nonlinear ❖ Calendar Variation-		
References Seasonality-Trend		



SOME REMARKS



The M3-Competition: results, conclusions and implications

- (1) Statistically sophisticated or complex methods do not necessarily provide more accurate forecasts than simpler ones.
- (2) The relative ranking of the performance of the various methods varies according to the accuracy measure being used.
- (3) The accuracy when various methods are being combined outperforms, on average, the individual methods being combined and does very well in comparison to other methods.
- (4) The accuracy of the various methods depends upon the length of the forecasting horizon involved.

Makridakis & Hibon (International Journal of Forecasting, 2000)



What should we be prepared for the future?



"The best way to predict the future is to create it"

Peter Ferdinand Drucker



19 November 1909 – 11 November 2005, An Austrian Born Management Consultant, Educator and Author, University of Frankfurt

Sources: Six Tools for Common Cause Variability Reduction for Pharmaceutical QA/QC and Manufacturing, Lynn Torbeck, Torbeck and Assoc.



References



- 1. Hanke, J.E. and Reitsch, A.G. (1995, 2005, 2008)

 **Business Forecasting*, 5th, 7th and 9th edition, Prentice Hall.
- 2. Armstrong, J.S. (2002)

 Principles of Forecasting: A Handbook for Researchers and Practicioners, Kluwer Academic Publisher.
- 3. De Gooijer, J.G. and Hyndman, R.J. (2006) 25 years of time series forecasting, International Journal of Forecasting, 22, 443-473.
- 4. Makridakis, S., Spiliotis, E., Assimakopoulos, V. (2020) *The M4 Competition: 100,000 time series and 61 forecasting methods*, International Journal of Forecasting, 36 (1), 54-74.
- 5. https://trends.google.com/trends/?geo=US