

# It's about Time

How to make sense of your time series data?

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# How do I know that I have time series data?

## Definition

An ordered sequence of values of a variable at equally spaced time intervals.

## Usage

- ➊ Obtain an understanding of the underlying forces and structure that produced the observed data.
- ➋ Fit a model and proceed to forecasting, monitoring or even feedback and feedforward control.

## Applications

Economic Forecasting, Sales Forecasting, Budgetary Analysis, Stock Market Analysis, Yield Projections, Process and Quality Control, Inventory Studies, Workload Projections, Utility Studies, Census Analysis, and many, many more...

# Life is a Sine Wave

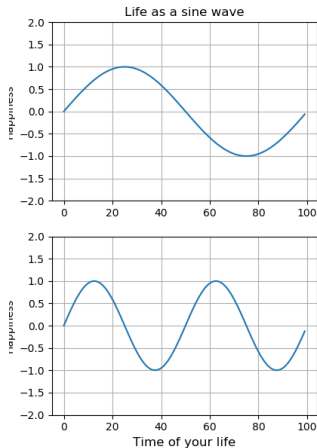


Figure 1: Life is different for everyone, but for all of us it's based on a sine wave.

# Time Series Decomposition

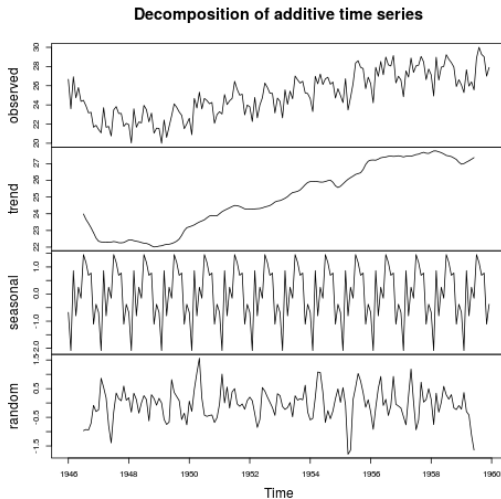
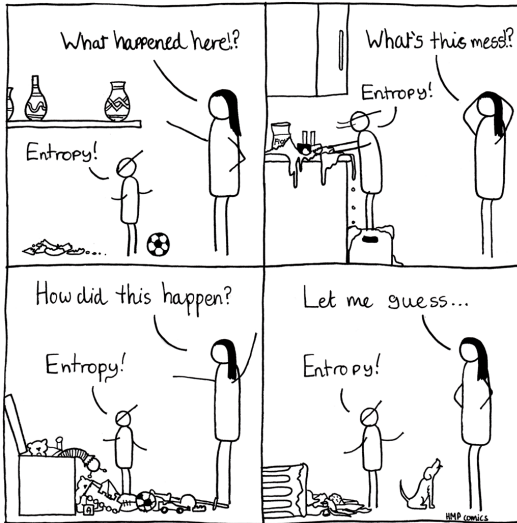


Figure 2: A decomposed population time series.

# Case Study: Tractor and Farm Equipment Manufacturing Company

- ▶ Established after WWII
- ▶ Slow but consistent growth since inception
- ▶ Struggles to keep inventory and production costs down because of variability in demand
- ▶ Pressure to reduce production costs
- ▶ Want to understand impact of marketing on sales

# Why is your bank password safe?



This is why we don't teach our children  
about entropy until much later...

# Case Study Decomposition

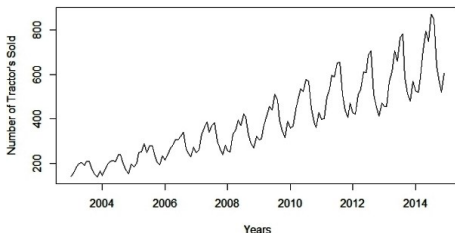


Figure 3: Original sales time series.

- ▶ **Trend** – overall direction of the series (upwards, downwards)
- ▶ **Seasonality** – monthly or quarterly patterns
- ▶ **Cycle** – long term business cycles
- ▶ **Remainder** – random noise left

## Model

$$Y_t = f(Trend_t, Seasonality_t, Remainder_t)$$

## Moving Average

$$MA = \frac{\sum_{i=-m}^m Y_{t+i}}{2m}$$

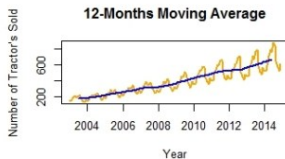
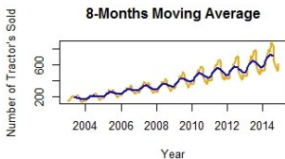
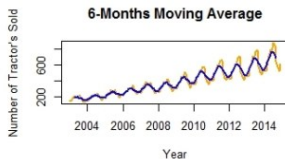
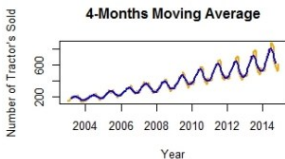


Figure 4: Smoothing the trend with moving average.



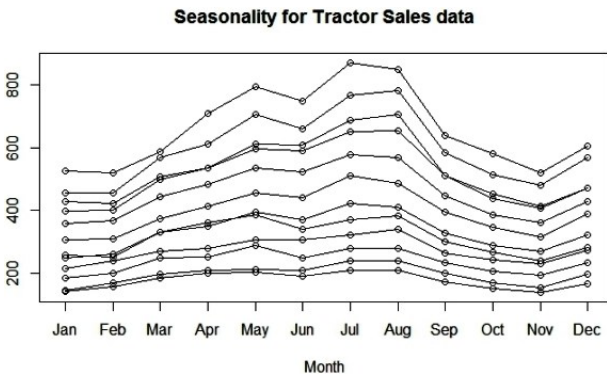


Figure 5: Looking into seasonality.

## Multiplicative Time Series

$$Y_t = Trend_t * Seasonality_t * Remainder_t$$

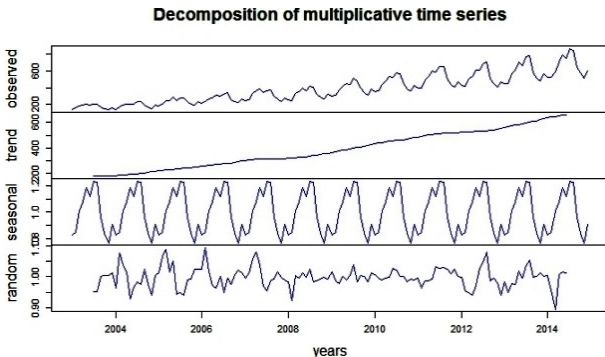


Figure 6: Decomposing a multiplicative TS model.

- ▶ **Trend:** 12-months moving average looks quite similar to a straight line hence you could have easily used linear regression to estimate the trend in this data.
- ▶ **Seasonality:** seasonal plot displays a fairly consistent month-on-month pattern. The monthly seasonal components are average values for a month after removal of trend. Trend is removed from the time series using the following formula:

$$Seasonality_t * Remainder_t = \frac{Y_t}{Trend_t}$$

- ▶ **Irregular Remainder (random):** is the residual left in the series after removal of trend and seasonal components. Remainder is calculated using the following formula:

$$Remainder_t = \frac{Y_t}{Trend_t \times Seasonality_t}$$



Figure 7: A sugar cane juice producer in India.

► **1st Pass Integrated:**

No differencing (d=0):

$$Y'_t = Y_t$$

1st differencing (d=1):

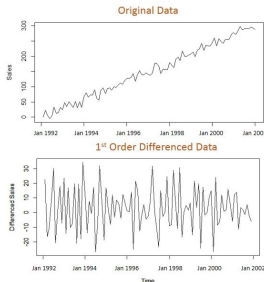
$$Y'_t = Y_t - Y_{t-1}$$

2nd differencing (d=2):

$$Y'_t = Y_t - Y_{t-1} -$$

$$(Y_{t-1} - Y_{t-2}) =$$

$$Y_t - 2 \times Y_{t-1} + Y_{t-2}$$



► **2nd Pass AutoRegressive:**

$$Y_t = c + \phi_1 Y_{t-1} + \phi_2 Y_{t-2} + \dots + \phi_p Y_{t-p} + e_t$$

► **3rd Pass MovingAverage:**

$$Y_t = c + e_t + \theta_1 e_{t-1} + \theta_2 e_{t-2} + \dots + \theta_q e_{t-q}$$

# White Noise and ARIMA

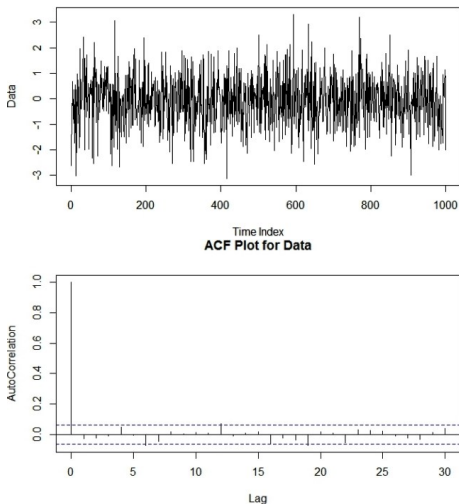


Figure 8: Checking for juice.

# Ever wanted to play Nostradamus?



Come the millennium, month 12,  
In the home of greatest power,  
The village idiot will come forth  
To be acclaimed the leader.

~ Nostradamus

AZ QUOTES

# Step-through Time Series Forecasting

Step 1: Plot the data

Step 2: Difference data to make it stationary on mean

*1st Differencing (d=1):  $Y'_t = Y_t - Y_{t-1}$*

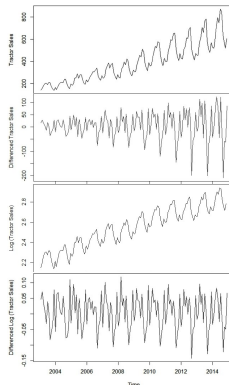
Step 3: Log transform data to make it stationary on variance

*Log of sales:  $Y_t^{new} = \log_{10}(Y_t)$*

Step 4: Difference log transform data to make it stationary on both mean and variance

*1st Differencing (d=1) of log of sales:*

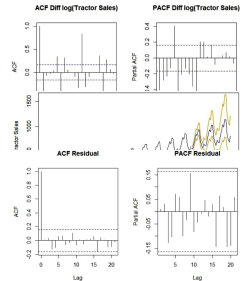
*$Y_t^{new'} = \log_{10}(Y_t) - \log_{10}(Y_{t-1})$*





# Step-through Time Series Forecasting

- Step 5: Plot ACF and PACF to identify potential AR and MA model
- Step 6: Identification of best fit ARIMA model
- Step 7: Forecast sales using the best fit ARIMA model
- Step 8: Plot ACF and PACF for residuals of ARIMA model to ensure no more information is left for extraction



# Akaike Information Criterion (AIC)

## Definition

Selects the best fit model by balancing between goodness-of-fit and number of parameters used in the model. Alternative: Bayesian Information Criterion (BIC).

## Formula

$$AIC = 2K - 2\ln(L)$$

# Case Study Regression with ARIMA Errors

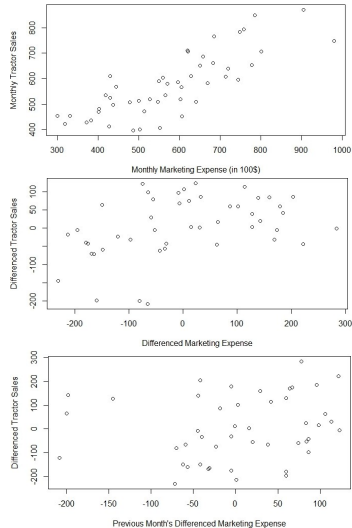


Figure 9: Marketing Expense, Differenced, and Previous Months

# CeADAR Demonstrator Applied Time Series Analysis and Prediction

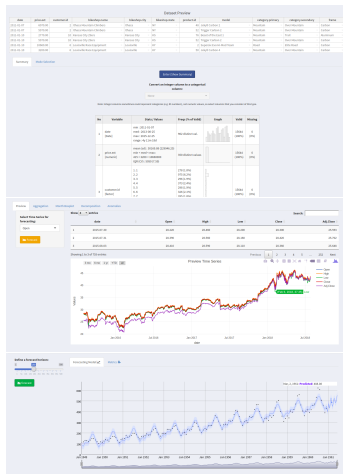


Figure 10: CeADAR Applied Time Series and Prediction demonstrator.



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TimeSeries use case and data: [ucanalytics.com](http://ucanalytics.com)

Images: AZ Quotes, Michael Sloan, [coolperthnights.com](http://coolperthnights.com)