



**SOFTWARE SOLUTIONS**

# **Forecasting, Prediction Models, and Times Series Analysis with Oracle Business Intelligence and Analytics**

## **Heartland OUG Spring 2013**

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# Presentation Agenda

- Understanding classification and forecasting (predictions)
- Use of Geneva Forecasting engine in Oracle OLAP
  - Holt-Winters and time series
  - Parameter choices
- ARIMA forecasting algorithm in R
  - Use Oracle R Enterprise
- Use of time dimension and time series functions in OBI



# Dan VlamiS and VlamiS Software Solutions

- Founded in 1992 by Dan VlamiS in Kansas City, MO
- Developed/implemented more than 200 Oracle BI systems
- Specializes in ORACLE-based:
  - Business Intelligence
  - Analytic Options to Oracle DB (OLAP, Data Mining, Spatial)
  - Data Warehousing
  - Training and mentoring
- Expert presenter at major Oracle conferences
- [www.vlamiS.com](http://www.vlamiS.com) (blog, papers, newsletters, services)
- Co-authored book “Oracle Essbase & Oracle OLAP”
- Beta tester for OBIEE 11g
- Reseller for Simba and Nokia map data for OBIEE
- HOL Coordinator for BIWA Summit 2013





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# BIWA Summit 2014, Jan 14-16

## Oracle HQ Conference Center

Business Intelligence, Warehousing and Analytics  
IOUG Special Interest Group



# Forecasting Today

- Predictions are the holy grail of BI systems and initiatives.
- Most all corporations have need for forecasting.
- Typical forecasting systems
  - Are stand alone or from ERP (not integrated to BI system)
  - Tend to use straight line or heuristic calculations.
  - Not always integrated into the business.
  - Are often tied directly to the budgeting process
- High level of angst surrounding forecasts.



# Forecasting Should...

- Should be integrated with rest of BI system.
- Should be another series of measures that are revealed in the context of historic information.
- Should be a part of the Common Enterprise Model.
- Should have visibility across functional areas and roles in corporations
- Should leverage most powerful calculation tools (database and BI system)
- Ideally adjusted based on an integrated view across corporate functions (marketing, operations, finance, etc.)



# Forecasting Methodologies

- Rule-based heuristic (last period, last period +5%, etc.)
- Cross-sectional methodologies (point in time)
- Time series (time sequenced data series)
- Mixed models
- Averages (moving, weighted, etc.)
- Linear and Non-linear regressions (line fitting)
- Transforms, projections, min/max



# Methodologies for Today

- OLAP Geneva Forecasting Engine
  - Holt Winters for time series
- Oracle R Enterprise
  - ARIMA
- ODM Classification and Regression (overview)
- OBIEE Time Series Functions (overview)





# OLAP Geneva Forecasting Engine

- FCOPEN function -- Creates a forecasting context.
- FCSET command -- Specifies the forecast characteristics.
- FCEXEC command -- Executes a forecast and populates Oracle OLAP variables with forecasting data.
- FCQUERY function -- Retrieves information about the characteristics of a forecast or a trial of a forecast.
- FCCLOSE command -- Closes a forecasting context.



# METHOD 'method'

- **AUTOMATIC** best fit for the data. (Default)
- **LINREG** linear regression ( $y=a*x+b$ ) is fitted to the data.
- **NLREG1** nonlinear regression  $x'=\log(x)$  and  $y'=\log(y)$  a polynomial model between  $x$  and  $y(y=c*x^a)$ .
- **NLREG2** nonlinear regression  $x'=x$  and  $y'=\ln(y)$  an exponential model between  $x$  and  $y(y=c*e^{ax})$ .
- **NLREG3** nonlinear regression  $x'=\log(x)$  and  $y'=y$  a logarithmic model between  $x$  and  $y(y=a*\log(x)+b)$ .
- **NLREG4** nonlinear regression method  $x'=1/x$  and  $y'=1/y$  an asymptotic curve ( $y=x/(a+bx)$ ).
- **NLREG5** nonlinear regression method  $x'=x$  and  $y'=\ln(y/(K-y))$  an exponential asymptotic curve ( $y=cKe^{ax}/(1+ce^{ax})$ ).
- **SESMOOTH** single exponential smoothing method intended for short term forecasts of non-seasonal data.
- **DESMOOTH** double exponential smoothing method exponential smoothing is applied to both the series and the trend term.
- **CROSTON** Croston's Intermittent Demand method. used for intermittent data where more than half of the observations are zero
- **HOLT/WINTERS** "triple" exponential smoothing. used on seasonal data



# Using “Holt-Winters”

- Triple “Exponential Smoothing” methodology
- Used for data suspected to be seasonal
- Needs multiple seasons
- Assumes regular periods
- Pre/post processing may be necessary (fiscal calendar 445, irregular holidays, “Black Swans”, outages, etc.)



# Exponential Smoothing

- Methodology for smoothing data and preferencing more recent periods when doing time series forecasts.
- Similar conceptually to a weighted moving average
- Weights decline according to an exponential function.  
 $\{1, (1-\alpha), (1-\alpha)^2, (1-\alpha)^3, \dots\}$
- Higher values give more weight to more recent periods
- Single (weighted average of most recent observation and the most recent smoothed statistic)
- Double (trend either up or down)
- Triple (period effect)



# FCSET Parameters

- **ALLOCLAST** {YES|NO}
- ALPHA {MAX|MIN|STEP} decimal
- **APPROACH** {'APPAUTO'|'APPMANUAL'}
- BETA {MAX|MIN|STEP} decimal
- COMPSMOOTH {YES|NO}
- CYCDECAY {MAX|MIN} decimal
- GAMMA {MAX|MIN|STEP} decimal
- **HISTPERIODS** integer
- MAXFACTOR decimal
- **METHOD** 'method'
- MINFCFACTOR decimal
- MPTDECAY {MAX|MIN} decimal
- NTRIALS integer
- **PERIODICITY** cycle-spec
- RATIO decimal
- **SMOOTHING** {YES|NO}
- TRANSFORM {'TRNOSEA'|'TRSEA'|'TRMPT'}
- TRENDHOLD {MAX|MIN|STEP} decimal
- WINDOWLEN integer



# Alpha, Beta, Gamma Setting

- Default Max is 0.3
- Default Min is 0.1
- Default Step is 0.1 ( $.05 \leq \text{divisible value} \leq 0.2$ )
- Greater value means nearer periods have more weight.
- Lower value means periods have more equal weight.



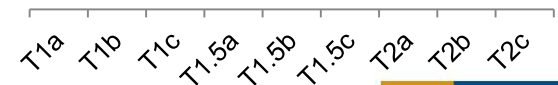
# Recommendations

- Be careful of accepting the APPAUTO setting
- Be aware of Embedded total time dimensions
- Match HISTPERIODS with PERIODICITY for best results
- PERIODICITY cycle-spec is hierarchical from higher grain to lower
  - Ex {52,7} 52 weeks in a year, 7 days in a week
  - Ex {4,13,7} 4 quarters in a year, 13 weeks in a quarter, 7 days in a week
  - Ex {12} 12 months in a year
- Months are challenging to incorporate with other periods



# Case Study Using Oracle OLAP

- Forecasted values from Oracle OLAP made no sense
- Client trying to use Best Fit – complicates study because don't know what method chosen
- Avoid tendency to inherit mistakes
- Problem in “HISTPERIODS” parameter
  - Solution: set HISTPERIODS to number of data points
- Problem in forecasting on hierarchical dimension – 12 month periods, 1 year period throwing off forecast
  - Solution: LIMIT TIME TO TIMELEVEL ‘PERIOD’
- 4-4-5 “periods” artificially inflate every 3<sup>rd</sup> period
- Added 3<sup>rd</sup> year – average of 2 years







# Example OLAP DML Forecast Program

```
vrb _handle int
```

```
" Removed error handling and definition of temporary variables such as DJOFCST2_C_SEASONAL  
LIMIT DJOFCST2_C_MEASURE_DIM TO 'QTY_HW'
```

```
_handle = FCOPEN('MyForecast')
```

```
limit djotime_d2 to djotime_d2_levelrel eq 'PERIOD'
```

```
SORT DJOTIME_D2 a DJOTIME_D2_END_DATE
```

```
"Set forecast parameters for 'best fit'
```

```
fcset _handle method 'HOLT/WINTERS' APPROACH 'APPMANUAL' SMOOTHING 'YES' MAXFCFACTOR 10.0 TRANSFORM 'TRSEA' -  
periodicity 12 histperiods 36 BETA MAX 0.5
```

```
"Execute the forecast - save seasonal and seasonal smoothed into the variables just defined
```

```
FCEXEC _handle time DJOTIME_D2 INTO DJOFCST2_C_STORED -
```

```
seasonal DJOFCST2_C_SEASONAL -
```

```
smseasonal DJOFCST2_C_SMSEASONAL backcast DJOFCST2_C_QTY
```

```
ALLSTAT
```

```
"Close the forecast
```

```
FCCLOSE _handle
```

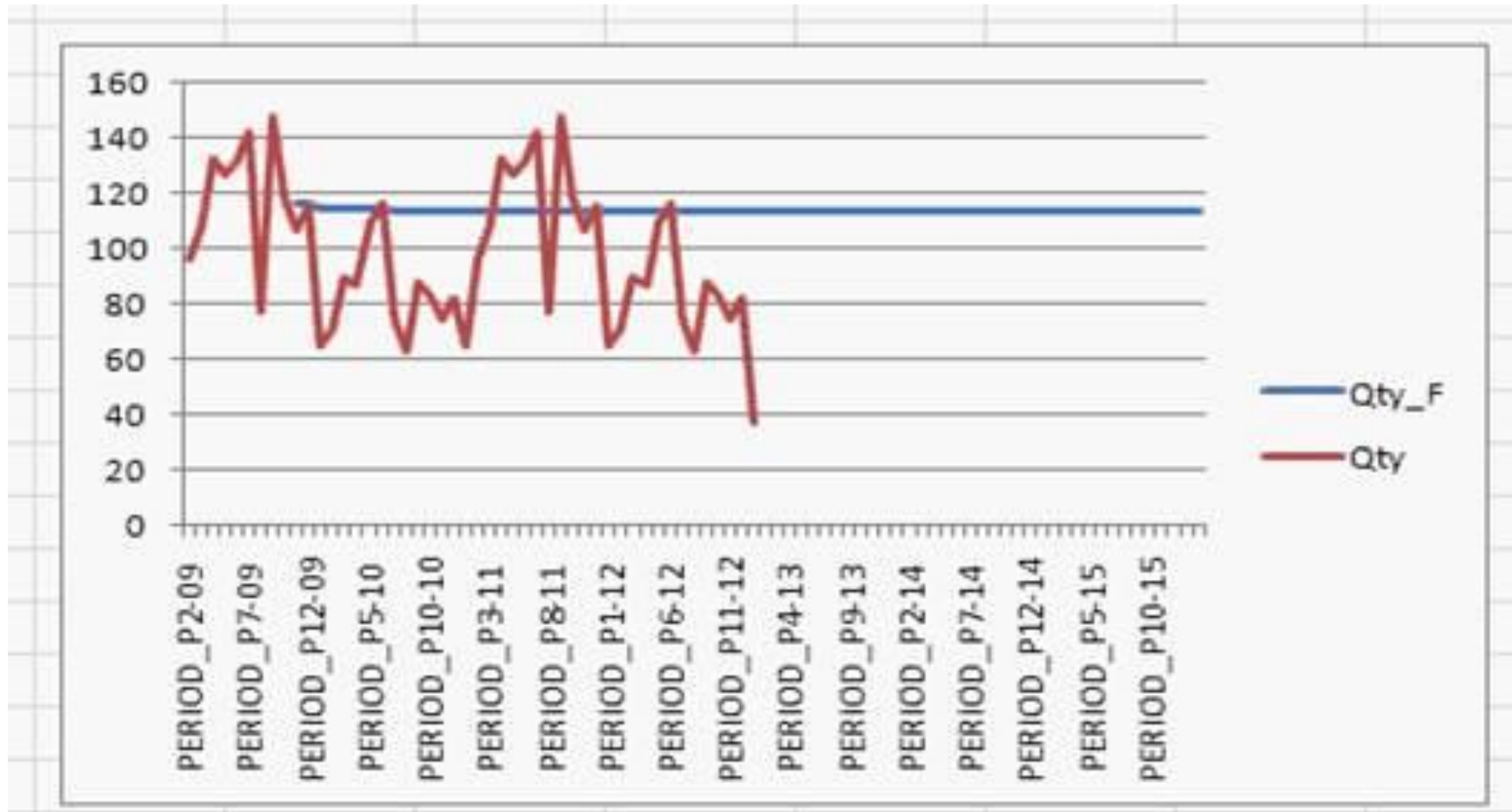
```
update
```

```
commit
```

```
return
```

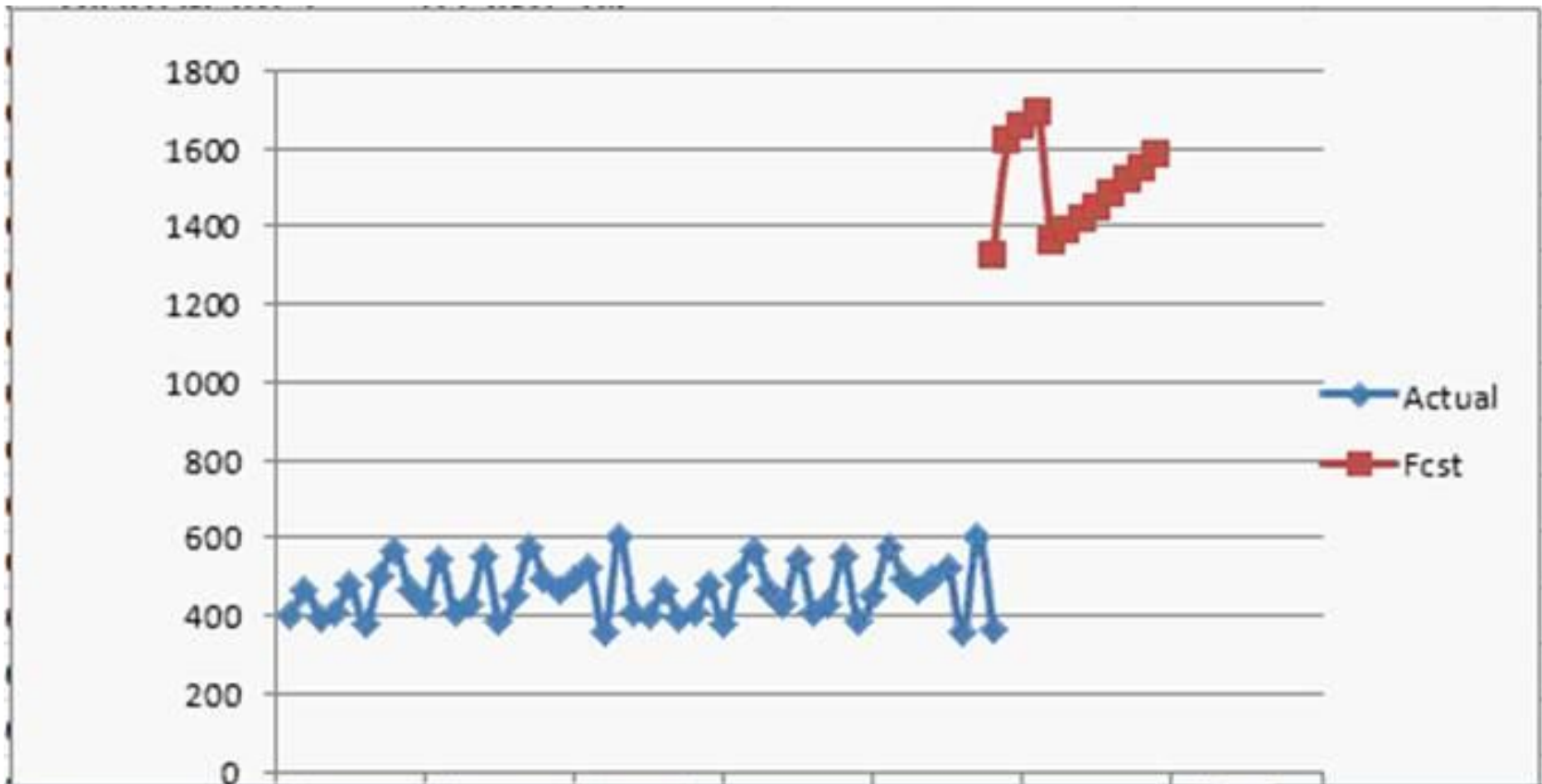


# Forecasts Did Not Make Sense



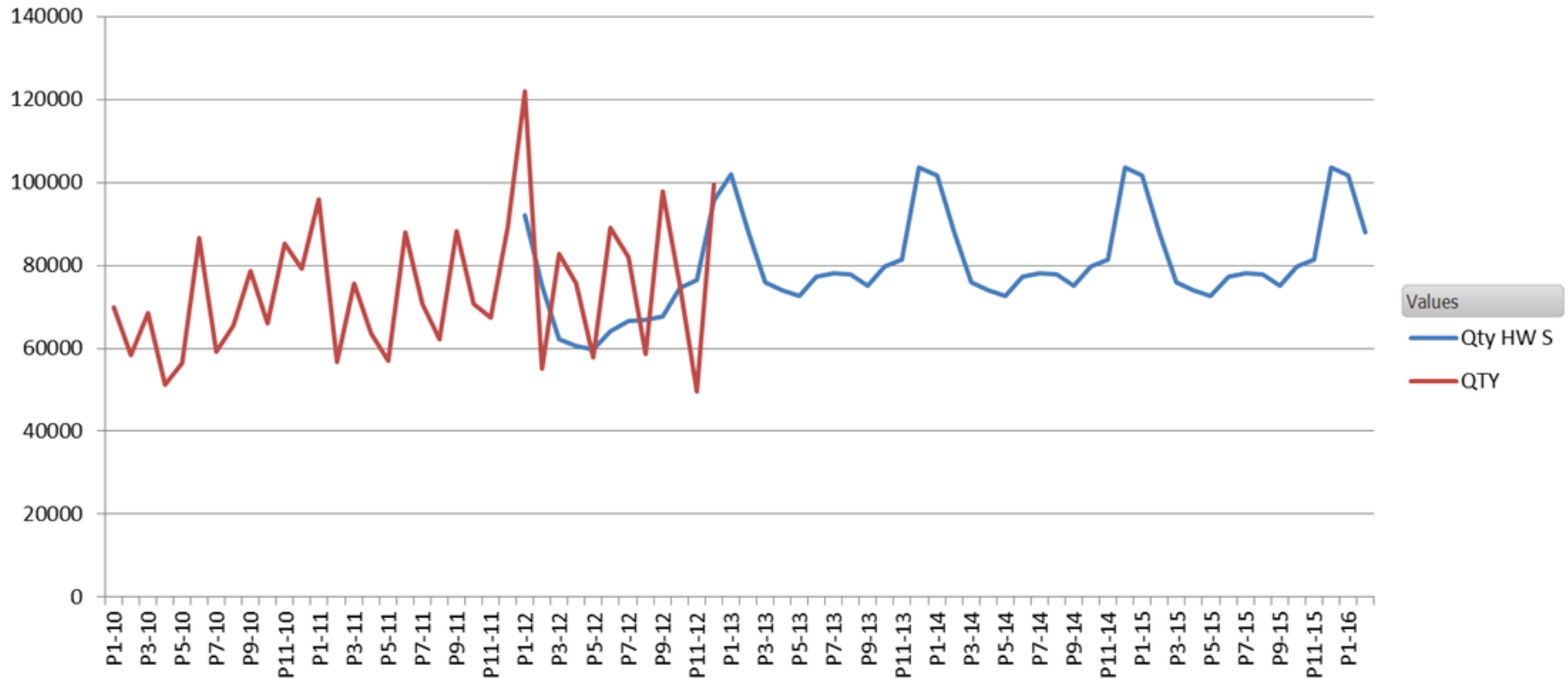


# Forecasts Did Not Make Sense



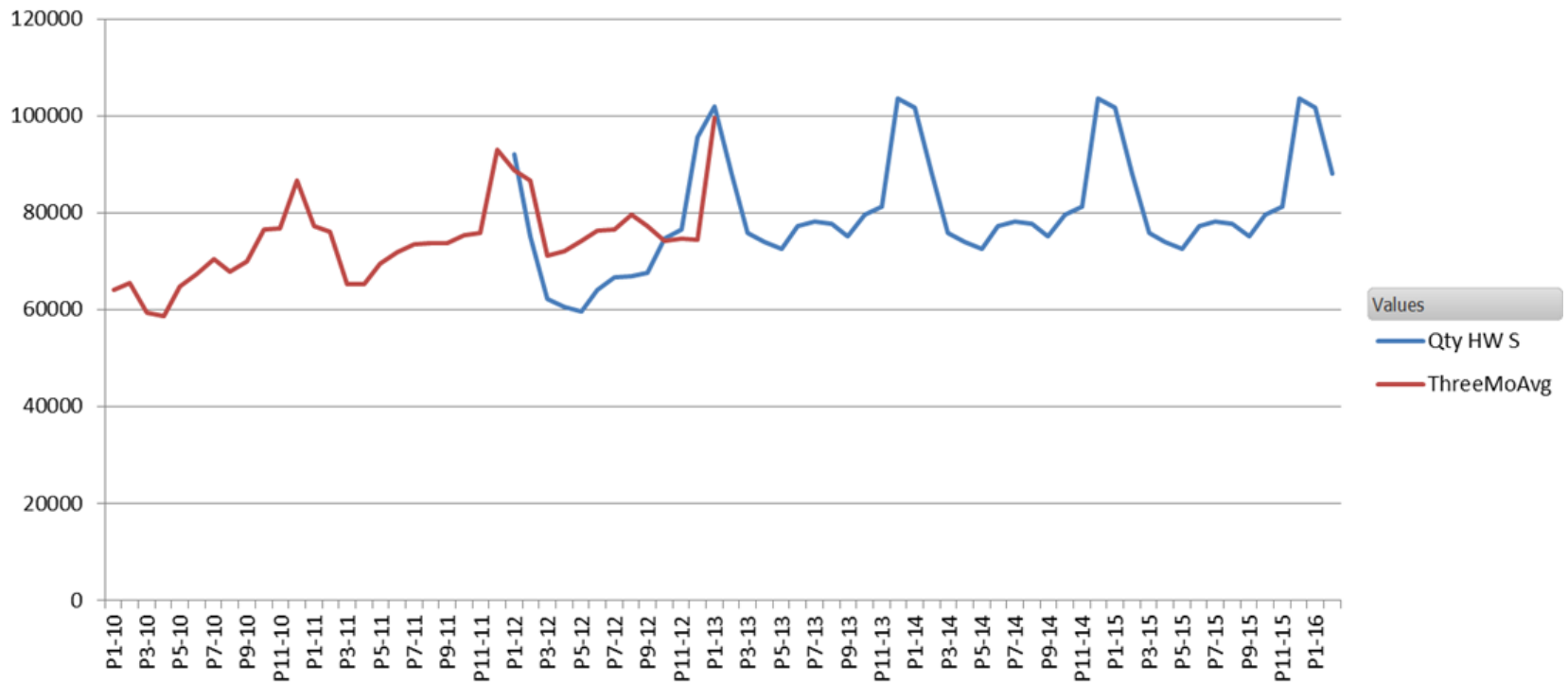


# Holt-Winters Forecast After Fix





# Holt-Winters Vs. 3-Mo Moving Avg





# Essbase @TREND

- Includes single, double, and triple exponential smoothing techniques.
- Includes linear and non-linear regression option.
- Does not include an auto-choice function.
- Non-linear regression transforms must be manually applied.
- Many other transform, calculation, and modeling capabilities in Essbase.



# ARIMA

- Autoregressive Integrated Moving Average
- Powerful algorithm for series analysis and prediction
- Three parameters  $(p, d, q)$ 
  - Auto regression (how reliant series values are on previous series values). AR(0) is white noise.
  - Integrated (degree of AR differencing, Random Walk)
  - Moving average (smoothing function)
- ARIMA (1,0,0) = AR(1)
- ARIMA (1,0,1) = ARMA (1,1)
- Large number of potential models (
- Know the name Rob Hyndman for ARIMA in R



# Stationarity

- Processes with no growth related to time.
- Random walks are stationary.
- Necessary to difference non-stationary series before applying ARMA models. (ARIMA handles this through the “Integrated” term “ $d$ ” of the  $(p, d, q)$  model parameters.)





# Non-Seasonal ARIMA $(p, d, q)$

- $\phi(B)(1 - B^d)\gamma_t = c + \theta(B)\varepsilon_t$
- $\{\varepsilon_t\}$  is a white noise process with 0 mean and variance  $\sigma^2$ .
- $B$  is a backshift operator
- $\phi(z)$  is a polynomial of order  $p$
- $\theta(z)$  is a polynomial of order  $q$



# Seasonal ARIMA $(p, d, q)(P, D, Q)_m$

- $\Phi(B^m)\phi(B)(1 - B^D)(1 - B^d)\gamma_t = c + \Theta(B^m)\theta(B)\varepsilon_t$
- $\{\varepsilon_t\}$  is a white noise process with 0 mean and variance  $\sigma^2$ .
- $B$  is a backshift operator
- $\Phi(z)$  is a polynomial of order  $p$
- $\Theta(z)$  is a polynomial of order  $q$



# Forecast() package in R

## Includes methods:

- `ets()`
- `auto.arima()`
- `Arima()`
- `arima()`
- `HoltWinters()`
- `StructTS()`

## Produces

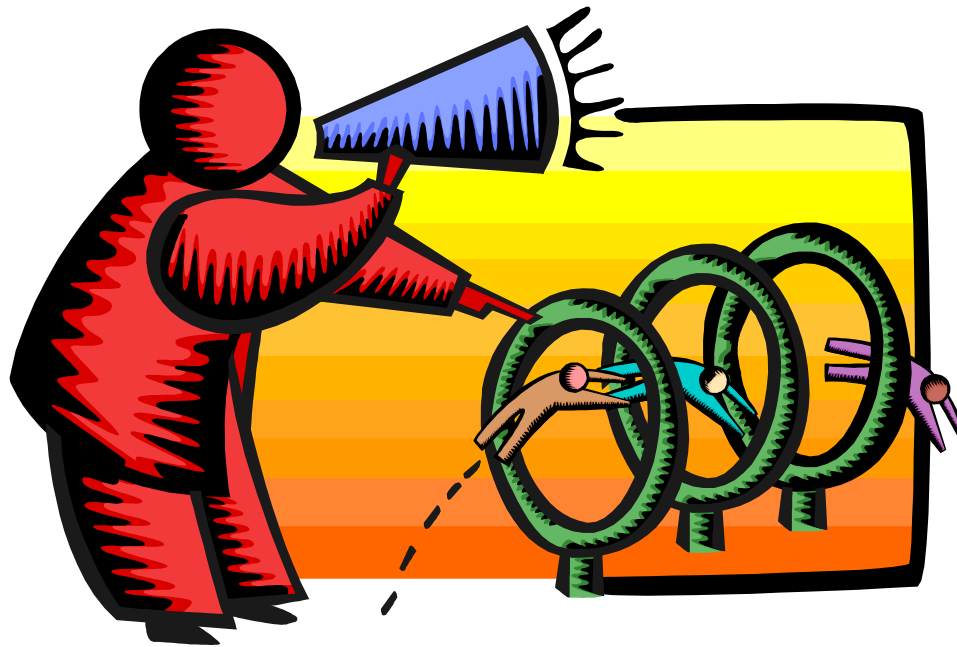
- the original series;
- $\hat{\mu}$  point forecasts;
- prediction intervals of specified coverage;
- the forecasting method used and information about the fitted model;
- $\hat{\mu}$  residuals from the fitted model;
- $\hat{\mu}$  one-step forecasts from the fitted model for the period of the observed data.



# Choosing an ARIMA model

- Auto.arima can be used for model choice.
- Manual model choice requires hypothesis testing and evaluation of results.
- Use minimum AIC to chose best model
  - $AIC = -2\log(L) + 2(p + q + P + Q + k)$
  - Compare AIC values to each other, absolute values carry no meaning

# arima Demo





# ARIMA vs. Holt-Winters

- Holt-Winters can be used for series that are seasonal and have a trend. (require order 2 differencing in ARIMA)
- Model selection can be complex in ARIMA and auto.arima selection may not be well understood.
- ARIMA best for stationary data series.
- ARIMA very powerful, but more to learn.
- Initial values more important in ARIMA (can have a big effect on predictions depending on model selected.)
- ARIMA provides confidence intervals



# Time Series Functions in OBI 11g

- Very powerful, accessible capability
- Time dimension must be designated
- Query results must be exact to pull from cache
- Can be “expensive” in processing
- Make sure that unique keys are defined at each level (“Jan13” rather than “Jan”)



# AGO function

- Defines a time-based offset
- Can nest multiple AGO statements (same level)
- Ago(<<Measure>>, <<Level>>, <<Number of Periods>>)
- Measure is a fact such as sales.
- Level is an optional term, default is set by the grain of the query (BY clause) or is specified in repository for level based measures.
- Number of periods is an integer specifying the offset value.





# TODATE

- Time-based aggregation function.
- Calculates based on starting value to current.
- Can nest with AGO (same level)
- ToDate(<<Measure>>, <<Level>>)
- Measure is a fact such as sales
- Level is the time grain such as year or month

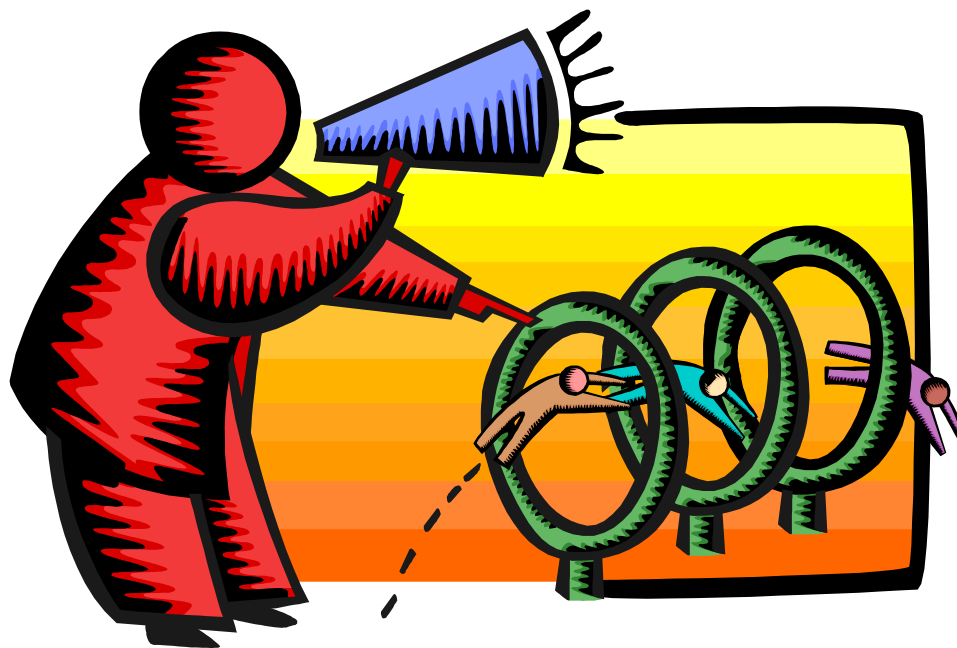


# PERIODROLLING

- Defines a period of time contextually
- Performs an operation across a specified set of query grain periods
- PeriodRolling(<<Measure>>, <<Starting Period Offset>>, <<Ending Period Offset>>, <<[Hierarchy]>>)
- Measure is a fact such as sales
- Starting Period Offset is an integer value, use a minus sign (“-2” means 2 periods ago)
- Ending Period Offset defines the end of the period, use a zero for current period
- Hierarchy is an optional setting to specify which time hierarchy to use such as “fiscal”
- Use “unbound” for starting period offset to calculate total from beginning
- PeriodRolling uses either the query level grain of “measure” or the measure level for “measure” if it has been set in the Admin tool.



# Oracle BI Trend Demo

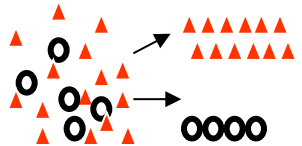
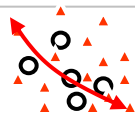
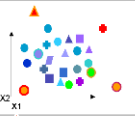
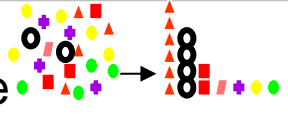
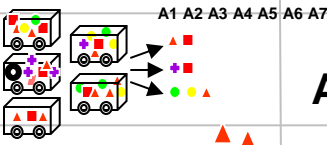
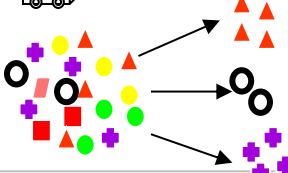
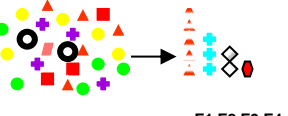




# Oracle Data Mining

- Oracle Data Mining is an option for the Enterprise Edition of the Oracle Database.
- A collection of APIs and specialized SQL functions.
- Includes a large number of specialized algorithms and built-in procedures.
- Makes use of many built-in capabilities of the Oracle Database
- ODM typically refers to “Oracle Data Mining”

# Oracle Data Mining Algorithms

Problem	Algorithm	Applicability
Classification 	<b>Logistic Regression (GLM)</b> <b>Decision Trees</b> <b>Naïve Bayes</b> <b>Support Vector Machine</b>	<b>Classical statistical technique</b> <b>Popular / Rules / transparency</b> <b>Embedded app</b> <b>Wide / narrow data / text</b>
Regression 	<b>Multiple Regression (GLM)</b> <b>Support Vector Machine</b>	<b>Classical statistical technique</b> <b>Wide / narrow data / text</b>
Anomaly Detection 	<b>One Class SVM</b>	<b>Fraud Detection</b>
Attribute Importance 	<b>Minimum Description Length (MDL)</b>	<b>Attribute reduction</b> <b>Identify useful data</b> <b>Reduce data noise</b>
Association Rules 	<b>Apriori</b>	<b>Market basket analysis</b> <b>Link analysis</b>
Clustering 	<b>Hierarchical K-Means</b> <b>Hierarchical O-Cluster</b>	<b>Product grouping</b> <b>Text mining</b> <b>Gene and protein analysis</b>
Feature Extraction 	<b>NMF</b>	<b>Text analysis</b> <b>Feature reduction</b>

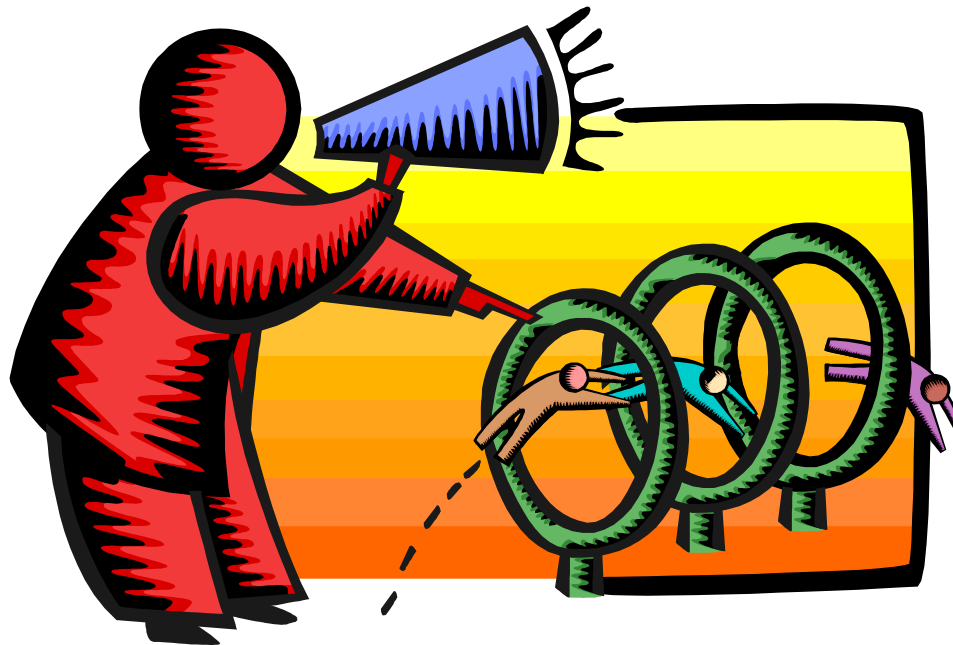


# Classification

- Prediction model for non-continuous information
  - Binary such as yes/no
  - Limited set (low/medium/high)
- Involves “supervised learning”
  - Prediction directed by a previously known dependent variable or “target” variable.
  - Commonly includes three phases:
    - Training
    - Testing
    - Scoring
- Results in predictive models that are applied to new data sets.
- In our example, we predict which prospects are likely to buy insurance.



# Oracle Data Mining Demo





# Oracle Test Drive

- Free to try out Oracle BI
- Go to [www.vlamis.com/testdrive-registration/](http://www.vlamis.com/testdrive-registration/)
- Runs off of Amazon AWS
- Hands-on Labs based on Collaborate 2012 HOLs
- Test Drives for:
  - Oracle BI
  - BI Publisher
  - Microsoft Excel against Oracle OLAP
  - Oracle Data Mining
  - Map Views in OBIEE
- Once sign up, you have private instance for 5 hours
- Available now





# Recent and Future VlamiS Presentations

- [Collaborate 13](#) April 7 - 11, 2013 • Denver, CO
- Advanced Dashboard Design in OBI 11g
- Advanced OLAP: Making the Hard Stuff Easy
- Blazing BI: The Analytic Options to the Oracle Database
- Map Views and Geospatial Analytics in OBI 11g
- VlamiS Process and Maturity Model: BI Project Best Practices
- Data Visualization Best Practices in Oracle Business Intelligence 11g
- Mobile BI: Using When and Where You Need It
- [Encore Presentation of OBIEE 11.1.1.7 TechCast](#) April 18, 2013
- What's New in OBIEE 11.1.1.7 IOUG BIWA TechCast
- [Central States OAUG Spring 2013 Executive Summit](#)  
April 30, 2013 • St. Louis, MO
- [Location Intelligence and Oracle Spatial and Graph User Conference](#)  
May 21 - 22, 2013 • Washington, D.C.
- Using Maps for BI Analysis at the USDA
- [ODTUG KScope 13](#) June 23 - 27, 2013 • New Orleans, LA
- Making Sense of Oracle's Business Intelligence and Analytics Offerings



# Thank You!