Business Analytics Providing Actionable Insights

With Application to Inventory Optimization

Overview

- About Business Analytics
- Current Situation in US and the World
- Statistical Techniques used in Advanced Business Analytics
- How to web-publish and run your analytics via a web-interface
- Examples of Analytics providing Actionable Insights:

About Advanced Analytics

- Analytics are tools that allows running of an analytical procedure
- Advanced Analytics
 - Use advanced Statistical and OR techniques
 - Provide Actionable Insights
- Analytics are best utilized if web-published enabling users can run them remotely via the Web
 - without having any data analytical software (e.g. SAS) on desktop
 - only with business knowledge and with no expertise in Statistics
- Developed working with experts in an organization

Almost any business can benefit from web-based analytics

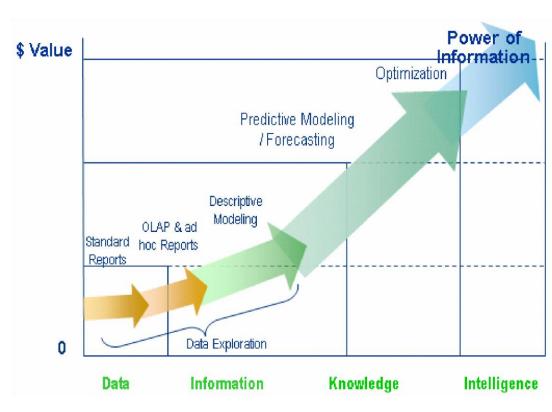
Advanced Business Analytics: Leveraging the power of analytics

- To get full benefit of your models, go beyond ROI estimation
- Optimize ROI
- Provide Business Insights to take specific actions

- Possible when model parameters are estimated at lowest level of decision making
- Use calibrated model to Optimize

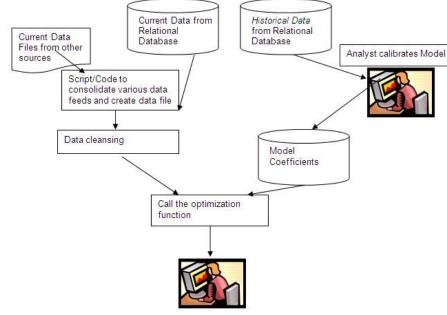
Promotional Tactics and Business Operations

- Web-publish analytics so that
 Marketing/Sales can run models
- No need to know Stats
 Techniques or Programming



Current Situation in US and the World

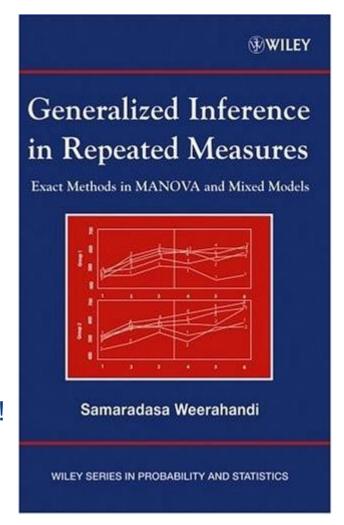
- US Companies are leading in Analytics
- Mostly Technology companies (e.g. Google, Yahoo, Amazon, NetFlix) used to lead in Analytics
- Now most companies use Analytics, but many lacks analytical value
- In-depth Business Analytics require
 - Statisticians
 - BI Developers with analytical knowledge
- Many companies perform simple
 Business analyses such as Trend,
 ROI calculations, Market Share



- Such analyses are of limited use
- Use Analytics providing Actionable Insights



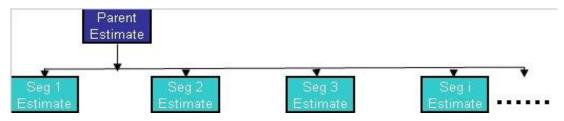
- Mixed models are now heavily used
 - in Business Analytics
 - In analysis of data from Clinical Trials
- In Corporate America performing advanced business analytics Hierarchical Mixed Models are heavily used
- You can read more about Mixed Models from freely distributed Book www.xtechniques.com
- The most widely used Statistical Techniques in Business Analytics was used to be LSE (Least Squares Estimator)!
- What is the Most widely used business analytical technique today?



- It is BLUP, the Best Linear Unbiased Predictor
 Statistical Techniques Used in Analytics (ctd.)
- In Corporate America performing advanced business analytics
 - BLUP has replaced LSE as the most widely used statistical technique
- Why? In estimating by large number of segments, estimates you get using
 - BLUP provides more accurate estimates than by LSE
 - BLUP yields shorter prediction intervals
 - Chance of getting wrong sign diminishes
- <u>Example:</u> Suppose you were asked to estimate consumer Response to Car Ad by Market (DMA). Then
 - Ad-stock Advertisement GRPs
 - Model sales by market as a function of ad-stocked GRP and other drivers of sales
 - If you model all covariates as "fixed effects" and use LSE you will not even yield the right sign for some markets

- So, model response to Car Advertisement as a random effect around the national average
- USE BLUP instead of LSE

Overview of the BLUP



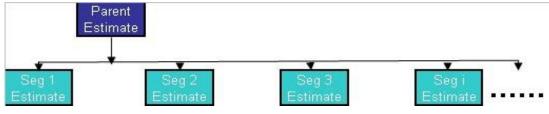
- Suppose certain groups/segments distributed around their parent
- Assumption in Mixed Models: Random effects are Normally distributed around the mean, the parent estimate, say M
- Suppose Regression By Groups yield estimate M_i for Segment in
- Let V_s be the between segment variance and V_e be the error variance, which are known as Variance Components

It can be shown that the BLUP of Segment i effect is

 $V_eM \square kV_sM_i$ $V_e \square kV_s$

- a weighted average of the two estimates, and k is a known constant that depends on sample size and group data
- The above is a shrinkage estimate that move extreme estimates towards the parent estimate

The BLUP (ctd): Some Inference Issues



Recall that BLUP of Segment i effect is

 $\frac{V_e M \square k V_s M_i}{V_e \square k V_s}$

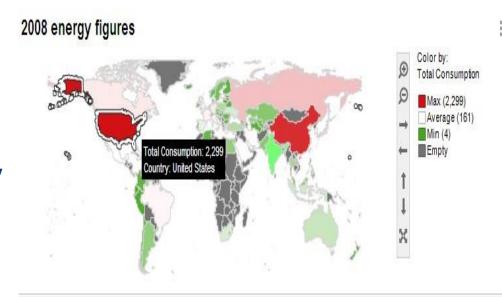
is a function of variance components (two or more)

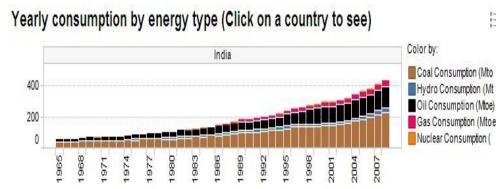
- MLE based methods frequently yield 0 (or negative) variance estimates, a drawback of
- Generalized Estimator (GE) and Bayesian Estimator (BE) do not suffer from such drawbacks
- Areas requiring further research:
 - Extending GE/BE to more complicated mixed models
 - Improving upon GE/BE using Stein type approach

How to Web-Publish Your Desktop Analytics: BI Software

- Use BI Software to web publish your analytics
- They provide Drilldown capabilities, Interactive capabilities, etc.

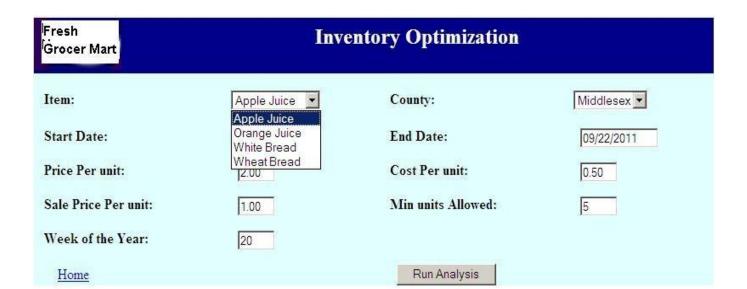
- BI software for reporting:
 Cognos, Business Objects,
 MicroStrategy, SAP, etc.
- SpotFire (see Image) and Microsoft Power BI allow any analysis with R
- Few other BI software allow programming in SAS and SPSS





Advanced Analytics Example: Inventory Optimization

- Analysts estimate Model parameters periodically; e.g. Demand Model for consumer products of a Supermarket Chain; e.g. Interface below
- Other parameters and scenarios are specified by users
- Model is applied real-time with latest data



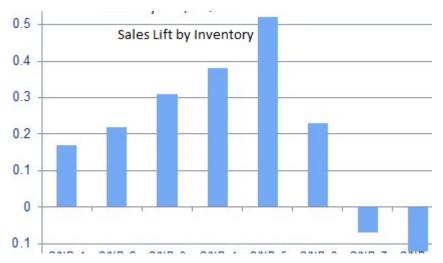
- Results are displayed or exported in desired format
- Optimization is done real-time:
 - Note: Average Demand is NOT the Optimum Inventory
 - Optimum Inventory is a Quanitile, a function of the Mean and Variance Both: Opt Inventory = $m + \sigma F^{-1}(1-c/p)$, where c is the unit cost and p is the unit price

Inventory Optimization: Leveraging to non-flat Demand models

- Preceding Inventory model for "Flat Demand" is called "Newsboy Model"
- Not appropriate even for Magazine Inventory Management, because demand may depend on Display.

Newsboy model is not appropriate also

- when the shape of Demand function, a piece-wise curve as shown in figure
- cannibalize brand sales due to over-inventory



Develop customized Demand models and Optimize for any type of logistics problems