

Business Analytics on zEnterprise High Performance Analytics & Integrated Attached Co-processors



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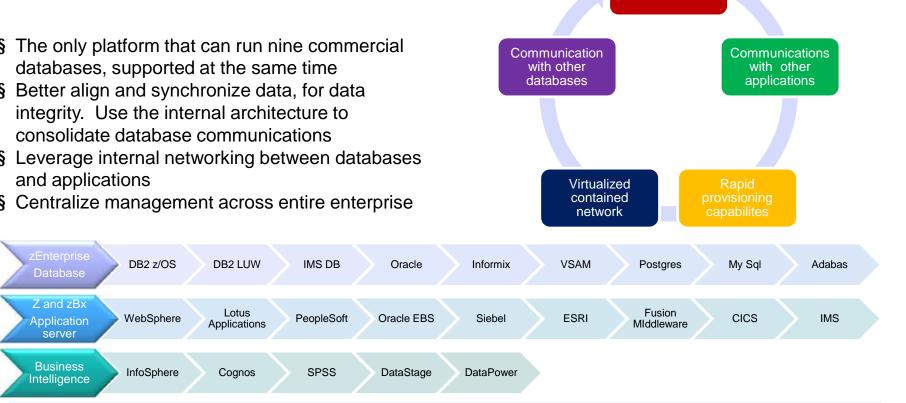
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zEnterprise Solutions Can Support and Integrate Data Like No Other Platform, Providing a Foundation for Other Analytic and Application Capability Centralized Management

- § The only platform that can run nine commercial databases, supported at the same time
- § Better align and synchronize data, for data integrity. Use the internal architecture to consolidate database communications
- **§** Leverage internal networking between databases and applications
- § Centralize management across entire enterprise



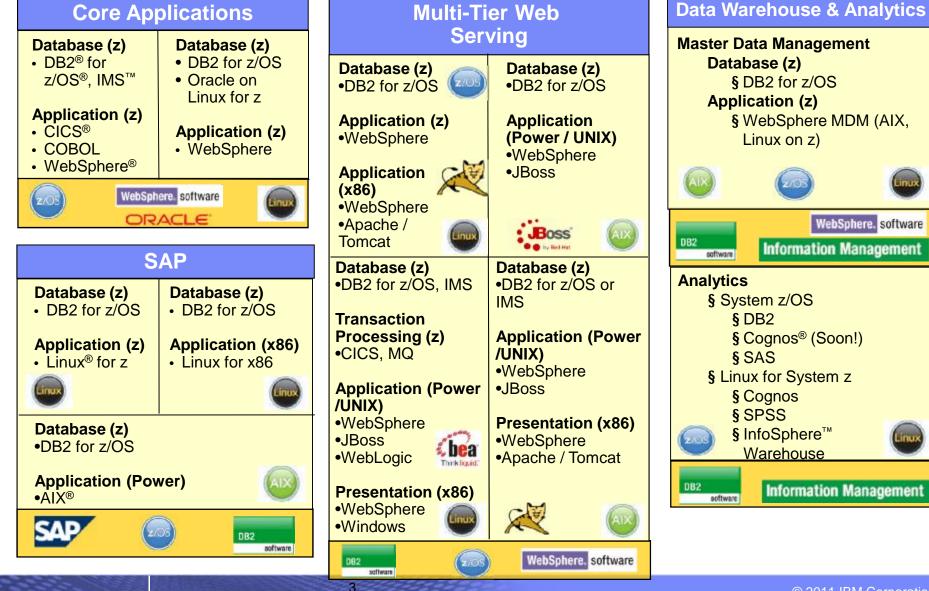
- § Consolidation of databases
- Tighter integration of data to applications §

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§ Business intelligence close to the data

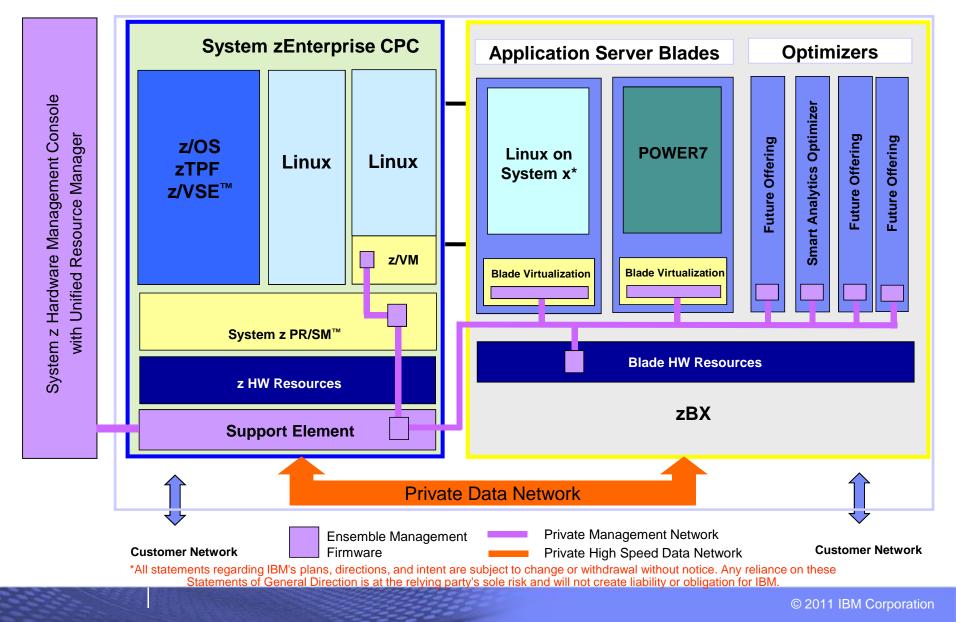


These workloads have recognizable patterns



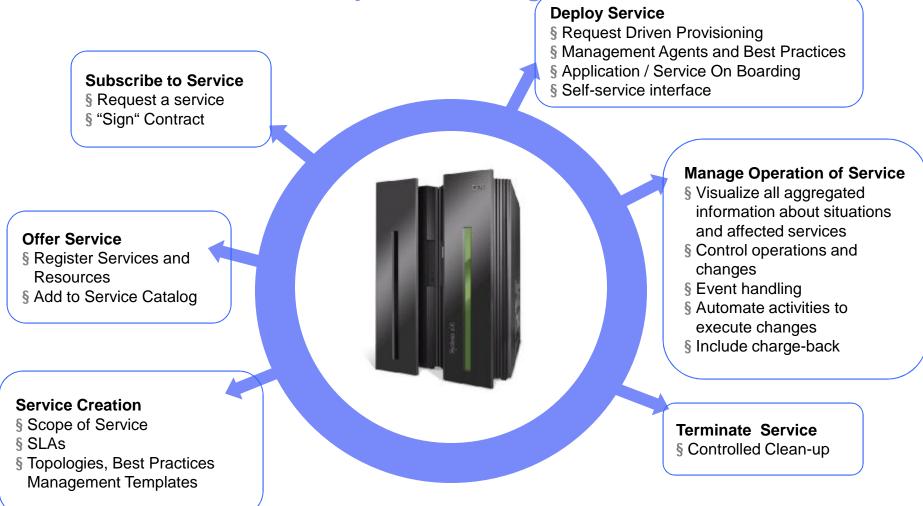
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zEnterprise with a System z Blade Extension (zBX)



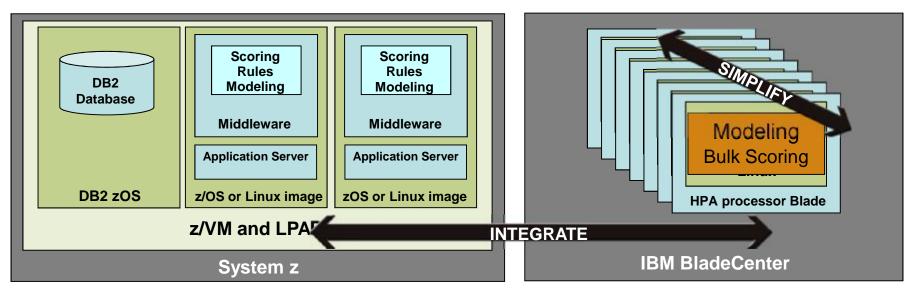


Cloud Service Lifecycle Management



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Hybrid Schema Mainframe and HPA Accelerator



Why Business Analytics on System z

- Highest Frequency compute threads in industry z196
- Very good floating point performance z196
- Large Shared Resource Pool
 - Single point of resource management
 - Single point of operational control
 - Efficient use of underlying compute resources
 - Manage unpredictable loads between instances
 - Easy/fast provisioning

• Integration w/Commercial Business Processing

- Security
- Reliability
- Availability
- Auditing
- Monetary Transactions

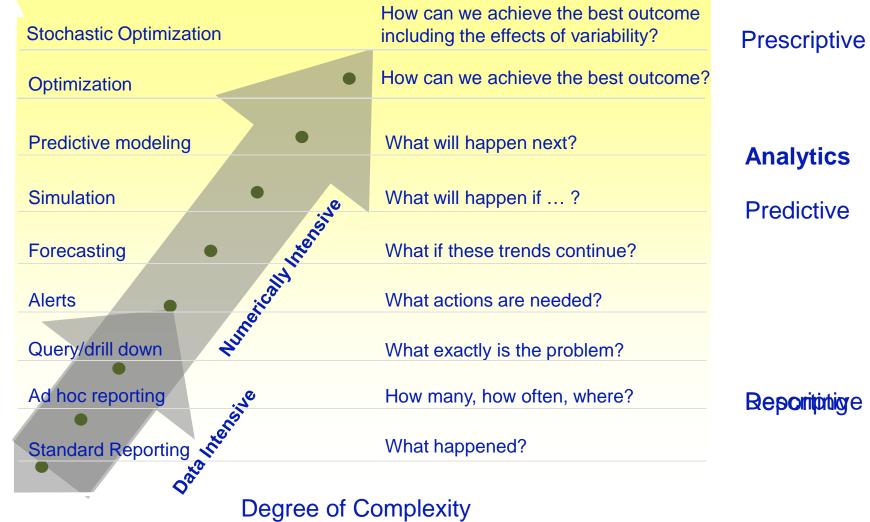
Why Analytics on HPA Blade

- Compute thread rich environment
- State of the art Vector/SIMD architecture

Why Analytics on zGryphon

- HPC enhanced commercial computing
- Single operational domain
 - Avoid standalone distributed cluster
- Extend strengths of System z

zHPC > EdgeHPC > Commercial HPC > Business Analytics (Mathematical) Analytics Landscape



Increasing prevalence of compute and data intensive parallel algorithms in commercial workloads driven by real time decision making requirements and industry wide limitations to increasing thread speed. Based on: Competing on Analytics, Davenport and Harris, 2007

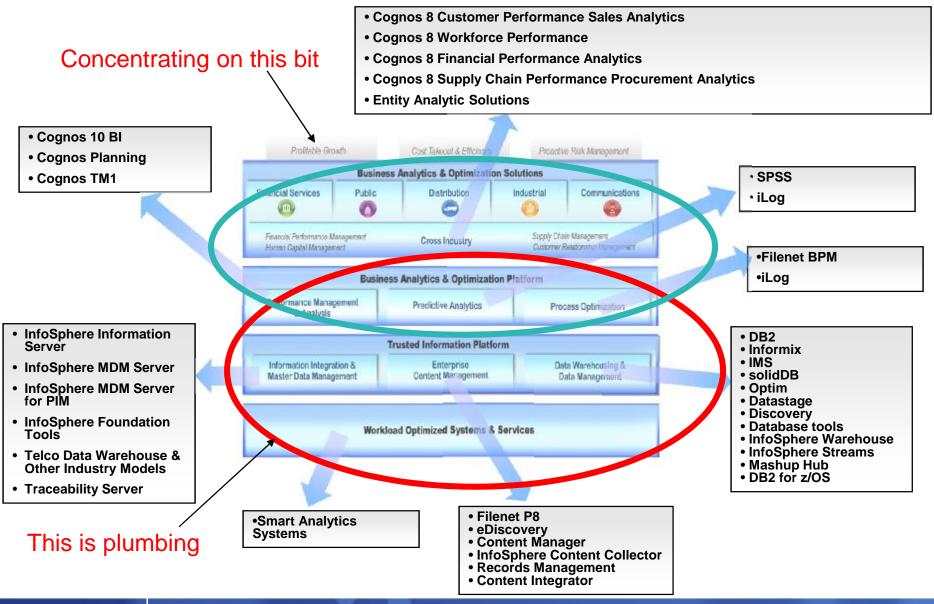


Competitive Advantage

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Market Leading Business Intelligence & Analytics Software



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Surveyed Customer Reqts

Customers want to integrate analytics with Operational processes

New BI trends map well to core strengths of DB2 for z/OS and System z

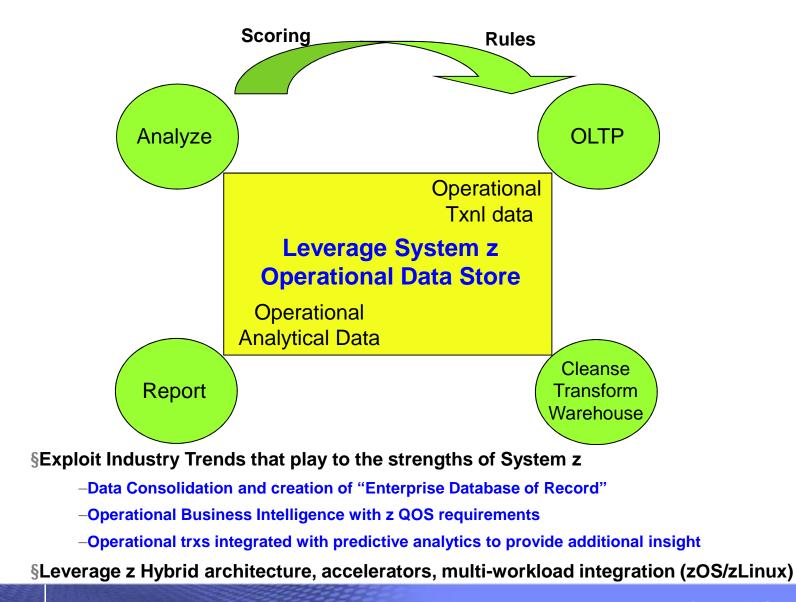
Mixed workload performance becoming single most important performance issue for DW/BI

Moving to a strongly centralized, shared infrastructure to improve economies of scale

- New DB2 features, Cognos/SPSS/ILOG software offerings, new optimizations and improved solution packaging with ISAS/ ISAO
- Single view of enterprise, Continuous availability/DR, Security, Governance, Query prioritization
- Virtualization and WLM enables consolidation of diverse DW and BI environments onto System z - zISAS
- z196 performance w/ integrated zBX + technology providing new ways to integrate analytic solutions while managing costs – iSAO



System z Platform Direction: From Data hub to Analytics hub



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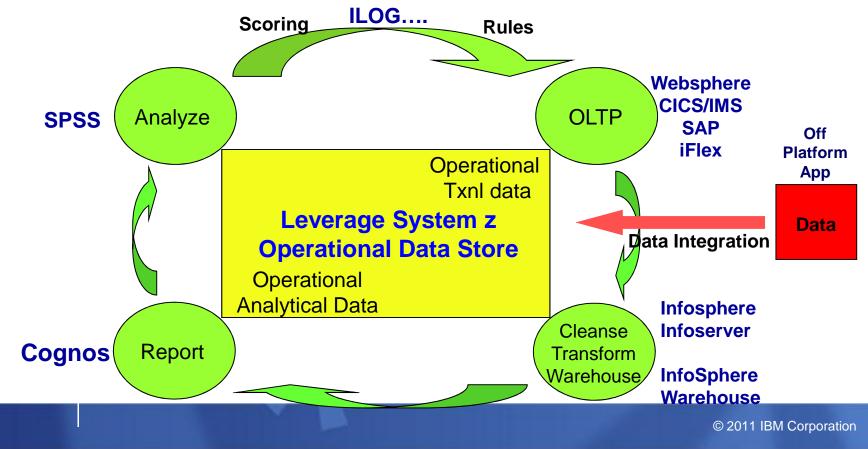


System z Platform Direction: From Data hub to Analytics hub

Sexploit Industry Trends that play to the strengths of System z

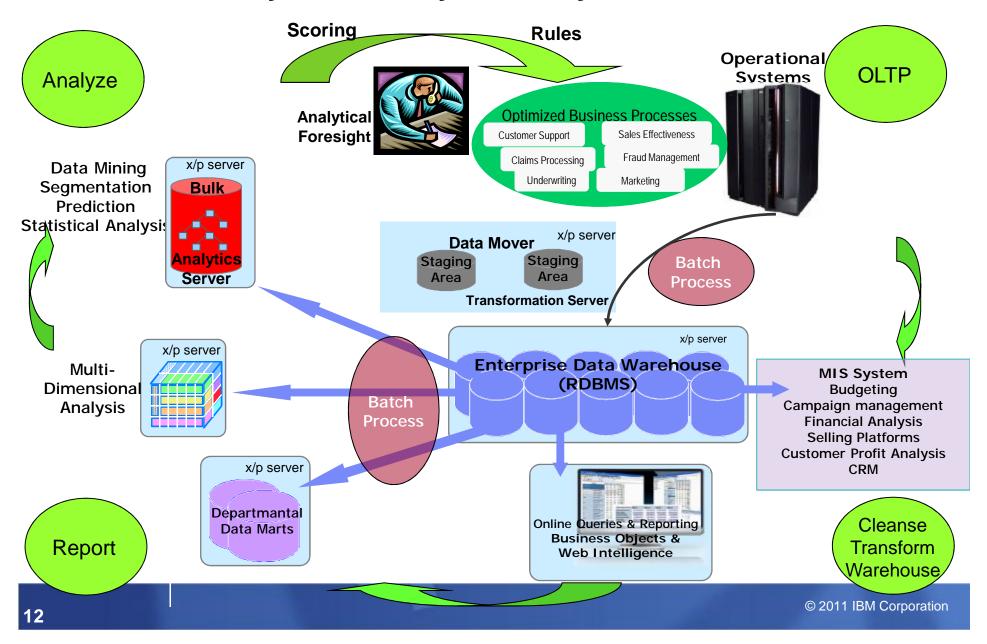
- Data Consolidation and creation of "Enterprise Database of Record"
- BI/Analytics application consolidation and creation of enterprise single version of truth
- Operational Business Intelligence with z QOS requirements
- Operational trxs integrated with predictive analytics to provide additional insight
- Superior end/end analytics life cycle integration
- Analytics as a service in an internal or external cloud

§Leverage z Enterprise architecture, accelerators, multi-workload integration (zOS/zLinux)



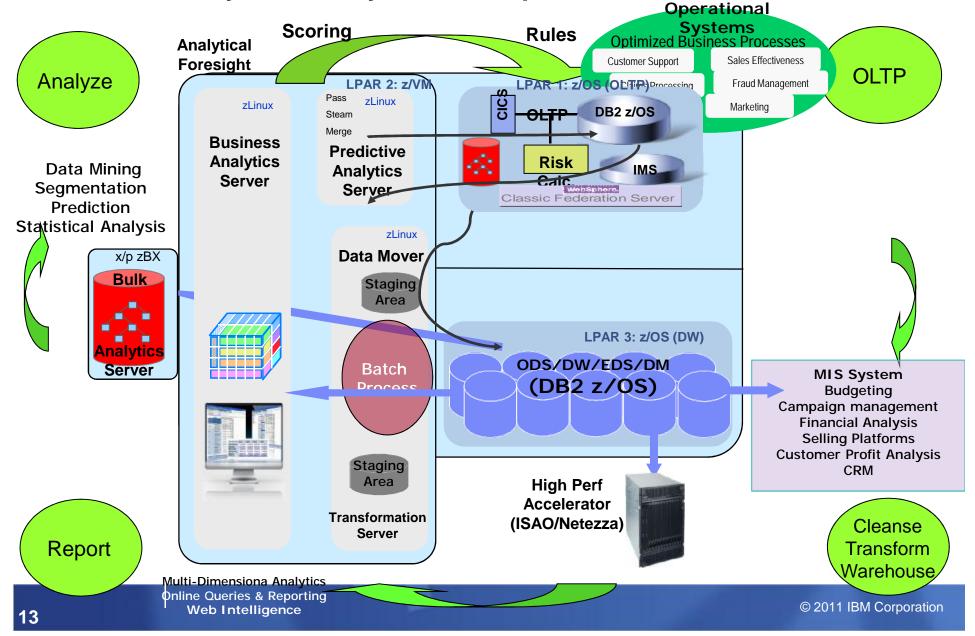
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Business Analytics Life Cycle – Async and Distributed



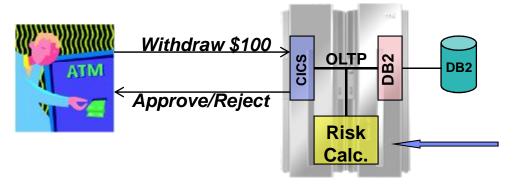
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Business Analytics Life Cycle – zEnteprise (IBM Smart Analytics System)

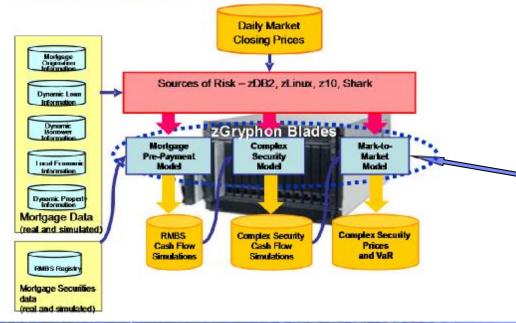




Evolution of OLTP



Bulk Data Analytics



ØReal time 'transactional' analytics

Credit Card Fraud Detection

Ø Compute intensive 'neural network' calculations required offload to alternative hardware

 ${\it 0}$ Batch runs overnight – business imperative for real time response. POC w/ ACI/PRM using z/OS and Cell.

> Latency costs of offload negated compute advantages of Cell

• Optimized on-board floating point architecture would re-host this application on z/OS

Ø Eliminate network latency delays

Ø Add value to OLTP transaction

 $\ensuremath{\ensuremath{\mathcal{O}}}$ Huge savings potential the sooner the act of fraud is detected

ØBatch and near real time •Risk Analysis (IBM Treasury POC)

Multiple repositories of operational data

Sophisticated numerical algorithms

> Bayesian probability algorithms

> Monte Carlo simulation

•Batch and near real time good match for host/accelerator offload

High performance accelerator HW building blockHigh speed bulk data transport

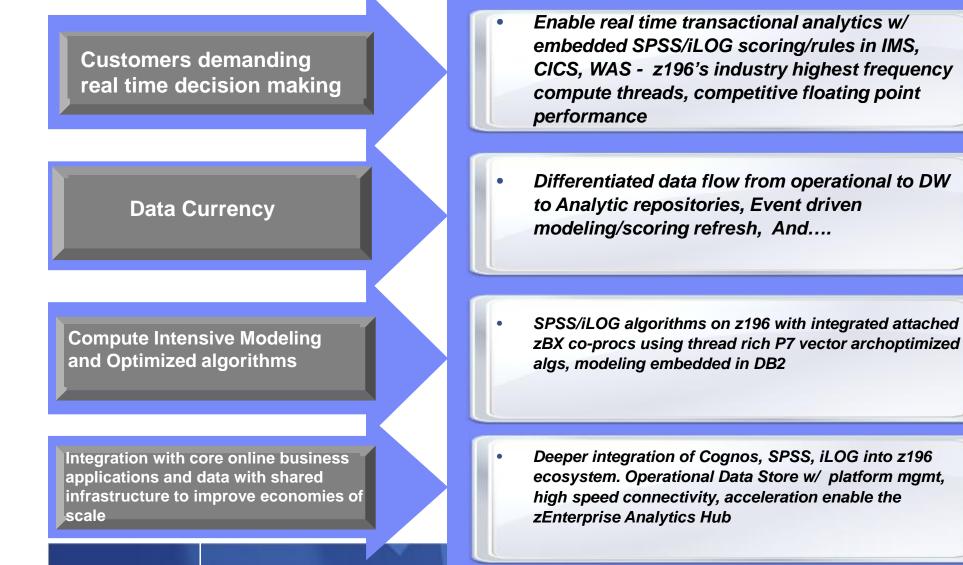
•Efficient data cleansing/transformation engines – ETL

•Value added proprietary data mining algorithms

Open standard host/accelerator programming model



System z and the Predictive Business



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Predictive Analytics Use Case Scenarios – US Credit Union Example

A. Higher withdrawal limits to increase customer satisfaction

- Many Neighborhood Financial Centers, ATMS, Kiosks do not have service personnel to override withdrawal limits.
- Ø Need real time method of scoring member to determine appropriate limit while limiting risk
- Ø Built a scoring model and embedded it in credit union's daily transaction processing system to automatically determine withdrawal limits
- Saved staffing costs, increased customer satisfaction, retention, enabled increased revenue generation with reduced risk

B. Targeted campaigns to improve retention, revenue

- Ø Exported member data from CU's BI system, applied analytic techniques such as regression to create member profiles to predict likelihood members will need additional products/services
 - Ø e.g. Home equity line of credit
- Ø Combined member usage characteristics w/ census information (i.e. local home ownership)
 - Ø Filtered out 30-40% of unlikely candidates. Focused on 60-70% most likely to respond
- Ø Increased 'lifted' revenue generated per marketing \$\$ by 60-100%
- Analysts wrote queries for rules to assist customer service. Recommendations pop-up on monitors during customer calls for relevant offers

C. Grow customer base while risk shrinks

- Ø Attract new customers w/ prior financial problems
- Ø Used scoring models to control deposit loss
- Ø Boosted CU bottom line and benefited customers avoiding check cashing services and payday lenders

D. Identify new branch locations

- Created predictive model to help identify new branch locations, operate existing branches more profitably, close sites
- Factor and regression analysis to identify composite performance based on new customers, deposits, loan distributions

Predictive Analytics enabled getting more mileage of data. Saved over \$1M annually, increased revenue and improved member satisfaction



Analytic Functional Areas

| Cross Sell | Analysis and exploitation of hidden relationships in data about existing customer behavior to predict efficient future activity (purchase of products) |
|----------------------|---|
| Direct Marketing | Analysis of customer characteristics (demographics, responses) to predict the amount of variability and tailoring of a marketing campaign |
| Collection Analytics | Analysis of customer characteristics to predict ability to pay and optimization of resources to facilitate collection. |
| Portfolio Prediction | Analysis of a portfolio of items (patients, products, financials, stores, etc.) to predict (score) a future outcome (survivability, placement, profitability, etc.) |
| Customer Retention | Analysis of a customers past characteristics to predict the likelihood of a customer's future action. |
| Risk Analysis | Quantitative analysis to numerically determine the probabilities of various adverse events and the likely extent of losses if the event occurs |
| Fraud Detection | Analysis of transactions to predict the likelihood of fraud usually based on a score or probability. |

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Mapping industry requirements to analytic functions Example: FSS (Banking and Insurance)

| FSS Analytics Trends | Industry Requirements | Relevant Functional Areas |
|----------------------|--------------------------------|--|
| Core Banking | Customer Insight | Customer Retention, Cross-Sell, Direct Marketing |
| | Product Recommendations | Customer Retention, Cross-Sell, Direct Marketing |
| | Fraud Detection and Prevention | Fraud Detection, Risk Analysis, Collection Analytics |
| | Underwriting | Risk Analysis |
| Payments | Fraud Detection and Prevention | Fraud Detection, Risk Analysis, Collection Analytics |
| | Anti Money Laundering | Fraud Detection |
| | Underwriting | Risk Analysis |
| Financial Markets | Fraud Detection and Prevention | Fraud Detection, Risk Analysis, Collection Analytics |
| | Portfolio Analysis | Portfolio Prediction, Risk Analysis |
| | Product Recommendations | Customer Retention, Cross-Sell, Direct Marketing |
| Insurance | Cause and Effect Analysis | Portfolio Prediction, Risk Analysis |
| | Underwriting | Risk Analysis |
| | Fraud Detection and Prevention | Fraud Detection, Risk Analysis, Collection Analytics |

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Mapping Trends and Requirements to Analytical Function Retail Sector

| Retail Trends in Analytics | Industry Requirements |
|---|--|
| Product Optimization and Shelf Assortment | Merchandise Performance |
| Customer Driven Marketing | Customer Insight/Customer Churn |
| Fraud Detection and Prevention | Fraud Detection and Prevention |
| Integrated Forecasting | Merchandise Performance/Customer Insight |
| Localization and Clustering | Store and Channel Performance |
| Market Mix Modeling | Promotion Planning |
| Price Optimization | Merchandise Performance |
| Product Recommendation | Promotion Planning |
| Real Estate Optimization | Store and Channel Performance |
| Supply Chain Analytics | Supply Chain Optimizations |
| Workforce Efficiency Optimization | Store and Channel Performance |

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Mapping Trends and Requirements to Analytical Function Telco Sector

| <u>Telco Trends</u> | Industry Requirements (from Sector Team) |
|---------------------|--|
| Market Optimization | Customer Churn |
| | Customer Retention |
| | Product Cross Sell |
| | Integrating Telco with retail sales |
| | Social Networking Models |
| | Behavioural Analytics |
| Network Analytics | Cell Tower Energy Management |
| | Network Traffic Optimization |
| | Capacity Planning |
| Revenue Assurance | Circuit Consolidation |
| | Budget Forecasting |

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Mapping Trends and Requirements to Analytical Function Healthcare Sector

| Healthcare Trends | Industry Requirements (from Sector Team) |
|---------------------|--|
| Life Sciences | Gene Pool Analysis |
| | Drug Discovery |
| | BioInformatics |
| Healthcare Payer | Insurance Fraud |
| | Clinical Cause and Effect |
| | Medical Record Management analytics |
| | Network Management analytics |
| | Employer Group Analytics |
| Healthcare Provider | Executive Analytics |
| | Patient Access |
| | Clinical Resource |
| | Patient Throughput |
| | Quality & Compliance |

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Mapping Functional Areas to Tasks

Function

Task

| Cross Sell | Association |
|----------------------|---|
| Direct Marketing | Classification, Clustering, Association |
| Collection Analytics | Clustering, Association |
| Portfolio Prediction | Prediction |
| Customer Retention | Classification, Estimation |
| Risk Analysis | Classification, Clustering, Prediction |
| Fraud Detection | Anomaly Detection |

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Mapping Tasks to Techniques/Algorithms

Task

Technique/Algorithm

| Association | Association Rules(Apriori), Decision Trees, Minimum Description Length |
|-------------------|---|
| Classification | Decision Trees, Neural Net, Naïve Bayes, Support Vector Machines |
| Clustering | Clustering, Attribute Analysis, K-Nearest Neighbor |
| Estimation | Logistic, Regression, Discrete Choice Models |
| Prediction | Linear Time Series, Non-Iinear Time Series, Exponential Smoothing |
| Anomaly Detection | Support Vector Machine |



SPSS Analytic Components – 1 of 4 Charts

| Procedure Family | Procedure | Computation Model Fit | |
|--------------------------------|---------------|--|--|
| LINEAR | ALM | Automatic linear modeling | |
| LINEAR | ANOVA | Analysis of variance | |
| LINEAR | DISCRIMINANT | Classify cases into groups based on predictor variables | |
| LINEAR | MEANS | Group means and statistics for target variables within categories of predictor variables | |
| LINEAR | ONEWAY | One-way analysis of variance | |
| LINEAR | REGRESSION | Regression | |
| LINEAR | T-TEST | T-tests for one sample, independent samples and pair samples | |
| LINEAR | UNIANOVA | Univariate analysis of variance | |
| LINEAR | GLM | General linear model | |
| LINEAR | 2SLS | Two-stage least-squares | |
| LINEAR | WLS | Weighted least-squares | |
| LINEAR | CSGLM | Linear regression for complex samples | |
| | | | |
| NON-LINEAR | GLMM | Generalized Linear Mixed Model | |
| NON-LINEAR | PLUM | Multinomial model for an ordinal target with 5 links | |
| NON-LINEAR | PLS | Partial least squares | |
| NON-LINEAR | COXREG | Cox proportional hazards regression to analysis of survival times | |
| NON-LINEAR | GENLIN | Generalized Linear Model | |
| NON-LINEAR | GENLOG | multinomial & Poisson general loglinear analysis & multinomial logit analysis | |
| NON-LINEAR | HILOGLINEAR | Multinomail hierarchical loglinear models | |
| NON-LINEAR | LOGLINEAR | multinomial & Poisson general loglinear analysis & multinomial logit analysis | |
| NON-LINEAR | MIXED | Linear Mixed Model | |
| NON-LINEAR | VARCOMP | estimates for variances of random effects under a general linear model | |
| NON-LINEAR | CNLR | Constrained nonlinear regression | |
| NON-LINEAR LOGISTIC REGRESSION | | | |
| NON-LINEAR NLR | | Nonlinear regression | |
| NON-LINEAR | NOMREG | Multinomial logit model for a polytomous nominal target | |
| NON-LINEAR | PROBIT | Logistic and Probit (binary) | |
| NON-LINEAR | CSCOXREG | Cox proportional hazards regression for complex samples | |
| NON-LINEAR CSLOGISTIC | | Nominal multinomial logistic regression for complex samples | |
| NON-LINEAR CSORDINAL | | Ordinal multinomial regression with 5 links for complex samples | |
| | COOLDINAL | | |
| /INING | Bayes Network | Bayes Network | |

| DATA MINING | NaiveBayes | Self Learning |
|-------------|------------|------------------------------|
| DATA MINING | SVM | SVM (Support Vector Machine) |
| DATA MINING | MLP | Neural networks |
| | RBF | Neural networks |



Categories of Optimization Problems Covered by ILOG Technology

| Mathematical Programming | Continuous Optimization (NP-complete) | linear programming (LP)•linear objective function•linear constraintsquadratic programming (QP)•quadratic objective functionquadratically constrained programming (QCP)•quadratic constraints |
|--|---|---|
| | Discrete Optimization (NP-hard) | mixed integer programming (MIP) •one or more non-continuous variables •includes MILP, MIQP, and MIQCP |
| Constraint Programming (Combinatorial Optimization) | | Vehicle Routing Job Scheduling Custom Search |

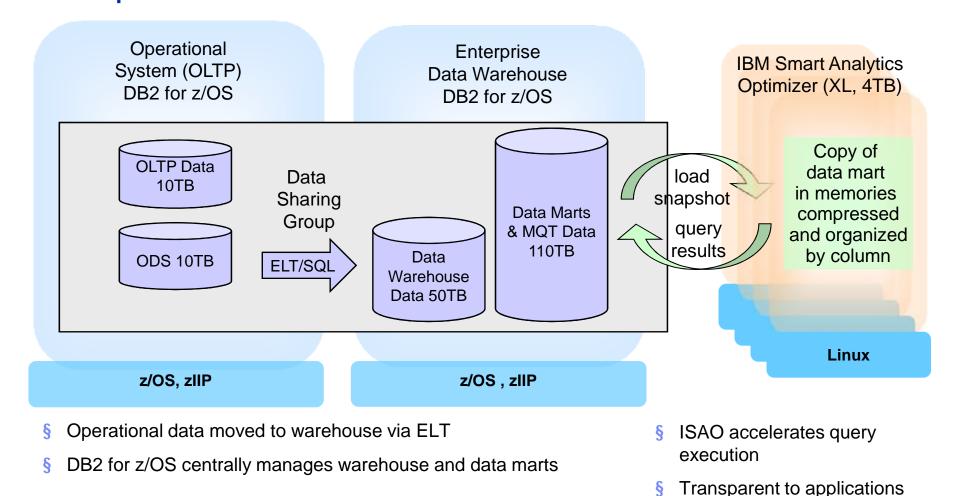


Major iLOG Algorithms of Mathematical Optimization

- § Optimisers
 - Simplex
 - Dual and primal simplex
 - Dual simplex is often the best choice
 - Problems where both dual and primal simplex perform poorly are rare
 - Research literature of running simplex on GPUs exists
- Barrier
 - Suitable for large, sparse problems
 - The only optimizer for QCP problems
 - Parallel version available
 - Network
 - Suitable for network-flow problems
 - Sifting
 - Suitable for problems with large column/row ratios
 - Extension of simplex
 - § Search strategies
 - Branch and cut
 - Search tree with nodes being subproblems
 - Parallel version available
 - Dynamic search
 - A variation of branch and cut



Data Warehousing And OLTP Co-Located On zEnterprise



ODS - Operational Data Store



Summary

- § Business Analytics exploits operational data to try to operate your business better.
- **§** Fully integrated solution: HPC + algorithms + transactions + data => insight
- Cognos, SPSS, ILOG, Infosphere WH with DB2/zOS provide the base for powerful new integrated Business Analytics Solutions with real time OLTP applications



Emerging host/accelerator programming models will facilitate the ease of exploiting co-processors without specific accelerator architecture knowledge with cross-vendor portability

ZEnterprise with integrated attached co-processors provides a unified combination of scalability, aggressive single thread performance and Power based throughput computing threads and vector processing



Questions



SPSS Predictive Analytics Models Available on System z

- § SPSS on Linux for System z supports over 30 models,
 - The 8 popular models support database push back for scoring in DB2 z/OS.
 - 5 popular models now available listed below:
- 1. Logistic regression, Trees (Algorithm names Include CHAID, Quest, C&R Tree)
 - Finance-Used in banking to predict which customers are credit worthy. Which customers should I make a loan to?
 - Finance, Retail, Insurance, Entertainment-Used in marketing departments to determine which customers are going to respond to an offer
 - Insurance-Used in insurance to determine which claims are legit vs. Fraudulent
 - Telecommunication -Predicting customer churn
- 2. Cluster Analysis (Algorithm names Include K Means, Kohonen, Two Step)
 - Finance, Banking, Insurance -Used in marketing departments across industries to better understand customer segments
 - Customer attrition analysis
- 3. Market Basket Analysis (Algorithm name" Apriori)
 - Retail -Product assortment planning
- 4. Time series analysis/forecasting
 - Retail -forecasting catalog sales, forecasting demand, sales planning
- 5. Cox Regression
 - Retail, Telecommunications -Predicting the time for customer churn
 - Healthcare -determining the efficacy of a drug