

AUCTIONING EXPERTS IN CREDIT MODELING

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Opportunities

- ◆ Anticipate default
 - Who are most likely to default in the near future?
- ◆ Detect fraudulent applications
 - Which loan applications are made up?
- ◆ Segment corporate bond market
 - Which companies are most risky?
- ◆ Other domains...
 - Employee evaluation: Who should we hire?
 - Disease prognosis: Who are most at risk?
 - Document classification: Can you find one like this?

Different contexts, but common characteristics...

- ◆ Rare events
 - Few cases dominate costs.
 - Millions of accounts, thousands of defaults.
- ◆ Synergies
 - Linear models find little. Interactions work.
 - Many combinations seem plausible.
- ◆ Wide data: more features than cases
 - Interactions, transformations, categories, missing data...
 - Too many to find the best at each stage.

Common Objective

- ◆ Regardless of the context
 - Credit default
 - Detecting fraudulent loan applications
 - Segmenting corporate bond market
- ◆ Pragmatic goal remains *prediction*.
- ◆ Best model generates highest revenue
 - Asymmetry of costs, presence of rare events
- ◆ Many schemes for building a predictive model
 - Algorithms, features, criteria...

Which model to use?

Every domain has experts...



But which offer good advice?



Automated Methods

- ◆ Expense of custom modeling hard to justify

- ◆ Automate process

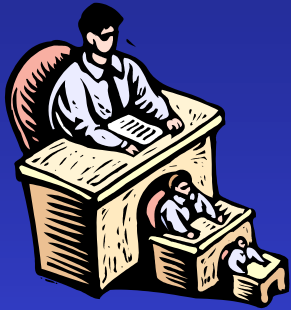
- Higher productivity
- “Objective”
- “Rigorous”
- Convenient



- ◆ But what about expert know-how?

- Is the loss of their insight worthwhile?

Comparison



Substantive

Pick model “by hand”

- ♦ Advantages
 - Leverage domain knowledge
 - Can “interpret” for regulator
- ♦ Disadvantages
 - Did we miss something?
 - Time consuming to
 - Construct
 - Maintain



Automatic

Computer search

- ♦ Advantages
 - Scans entire data warehouse
 - Hands-off, fast
 - Construction
 - Maintenance
- ♦ Disadvantages
 - Lost domain expertise
 - Hard to explain or interpret

Best of Both Approaches



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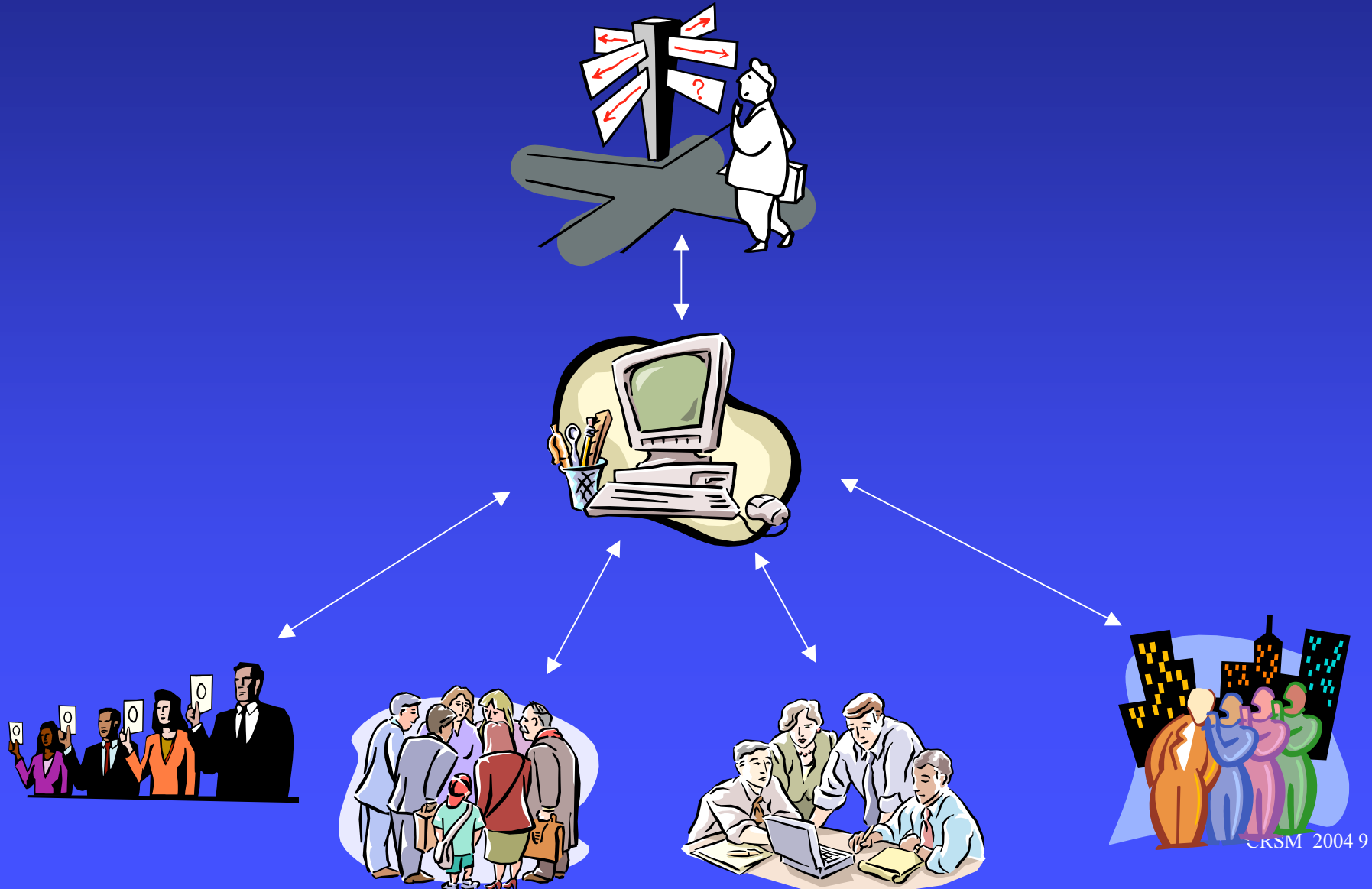


Automatic

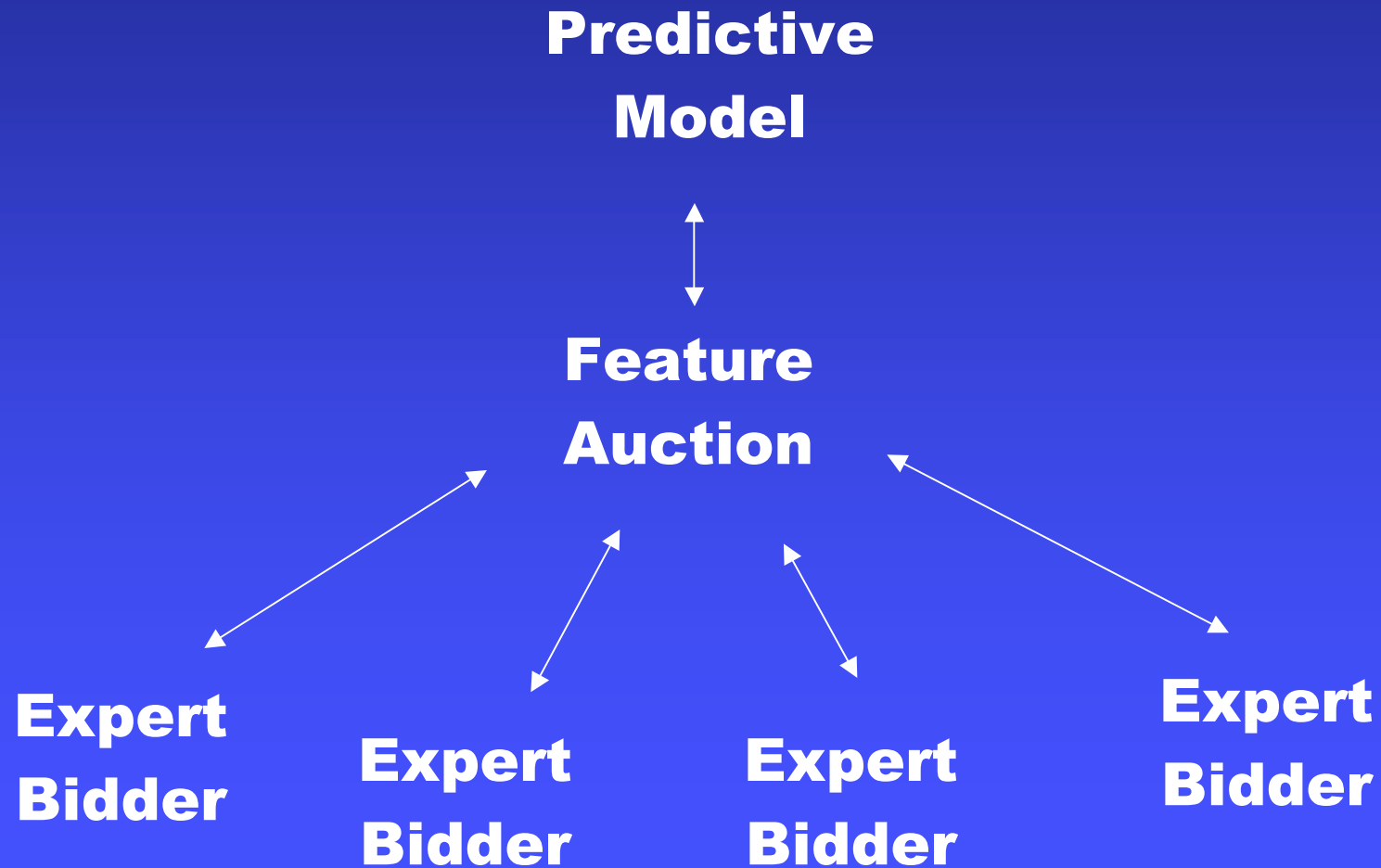
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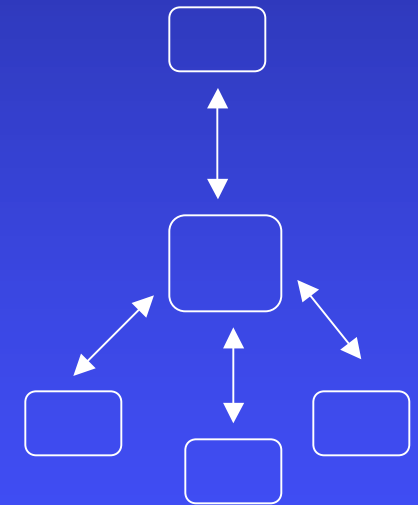


Auction = Experts + Model



AWKTION Modeling

- ◆ *Experts* recommend features.
 - Bid reflects strength of “conviction” (Bayes prior)
- ◆ *Auction* identifies feature with highest bid.
- ◆ *Statistical model* tests feature.
 - Bid determines p-value threshold
 - Accepts significant predictors, rejects others
- ◆ *Auction* passes results back to experts.
 - Winning bids earn wealth for expert.
 - Losing bids reduce wealth.
- ◆ *Information* flows both ways.



- ◆ Experts recommend predictive features
- ◆ *Substantive* experts order features
 - Domain knowledge of specific area
 - Prior models in similar problems
- ◆ *Automatic* experts
 - Interactions based on other experts
 - Transformations
 - Segments, nearest-neighbor, principal components
 - Nonlinearity
 - Feedback adjustments for calibration

Underlying Theory

- ◆ Streaming feature selection
 - Sequential, not all at once
 - “Depth-first” rather than “breadth-first”
 - Overcomes width constraints
 - Ordering captures prior information
- ◆ Universal bidding strategies
- ◆ Multiple testing without overfitting
 - False discovery rate (FDR) for infinite sequence of tests.
- ◆ Calibration
 - Ensures predictions track reality.
 - Adaptive link function

Sequential vs. Batch Selection

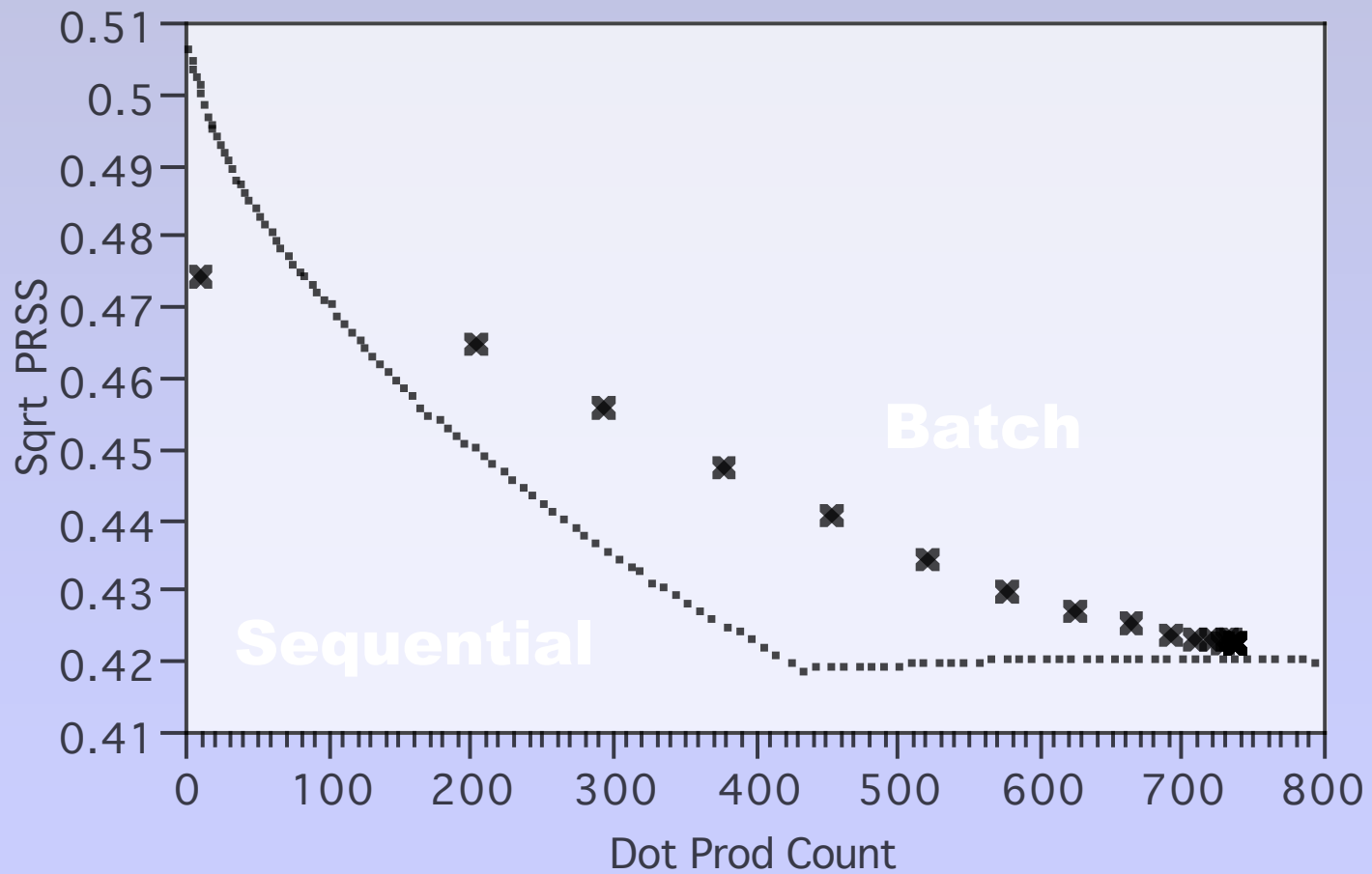
Sequential

- ◆ Search features in order identified by domain expert
- ◆ Allows an infinite stream of features.
- ◆ Adapts search to successful domains.
- ◆ Reduces calculations to a sequence of simple fits.

Batch

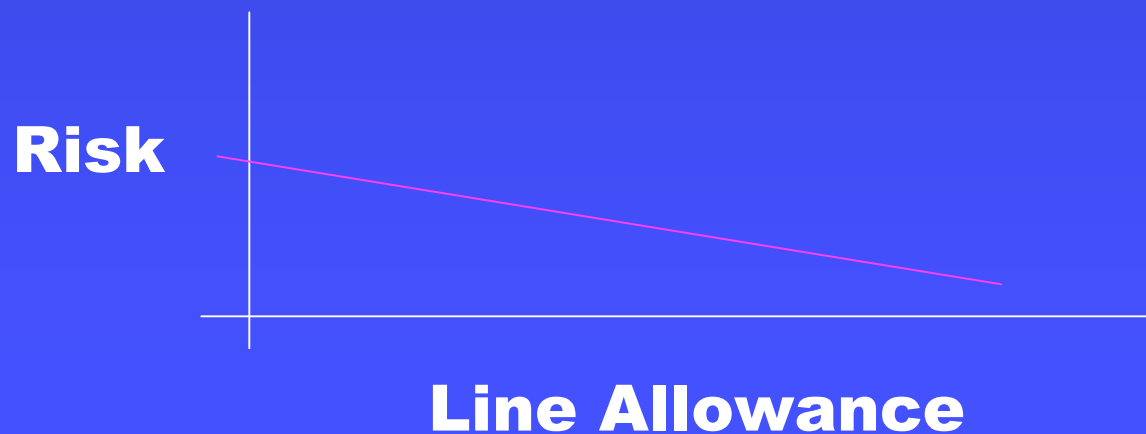
- ◆ Search “all possible” features to find the best one.
- ◆ Needs all possible features before starts.
- ◆ Constrains search to those available at start.
- ◆ Requires onerous array manipulations.

Sequential works...



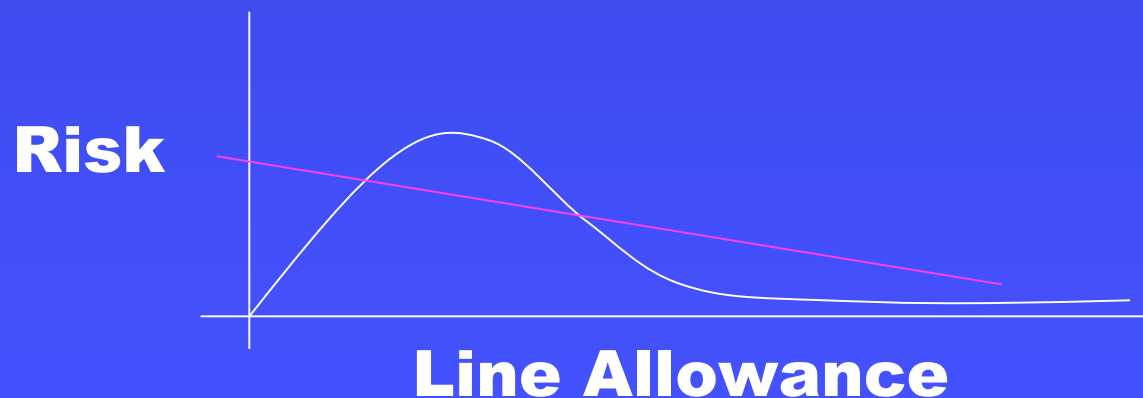
Example

- ◆ Predicting default
 - Logistic regression model
 - 15,000 cases, 67,000 possible features (most interactions).
- ◆ Standard model finds linear predictor
 - Higher risk with lower line allowance.
 - Statistically significant



Example: Nonlinear pattern

- ◆ Auction model
 - Experts recommendations based on state of model.
 - Look for combinations of extant predictors.
- ◆ Discovers nonlinear effect
 - Nonlinear effect for size of credit line
 - Statistically significant “bump” in risk



Example: Synergies

- ◆ Feedback expert
 - Builds interactions among predictors in current model.
 - Limited search does not obscure simple predictors.



Feature	Found in Model
Behavioral score	Marginally linear
Missing data	Behavior score affects these differently
Non-linear	Larger for high scores
Synergies	Changes with payment

Summary

- ◆ Auction modeling combines
 - Domain knowledge
 - Automatic search procedures
- ◆ Offers
 - Fast, guided search over complex domains
 - Strategies for constructing features in parallel.
 - Flexible statistical models
- ◆ More information...

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