

Translating Strategic Objectives into Individual Decisions

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The Emerging Landscape of Decision Optimization

Scott Horwitz Fair Isaac lan Brodie Fair Isaac



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AGENDA



- Three Key Insurer Challenges
- How Decision Analytics Can Help
- Pricing Optimization Case Studies
- Insurance Industry Demonstration
- Conclusion

Three Key Insurer Challenges



- 1. How do you translate high-level goals into individual decisions?
 - Trade-off between competing business goals and objectives
- 2. How do you account not just for what a customer is, but what a customer will do?
 - Your decisions affect consumer choices and behaviors
- **3.** How do you create strategies known to be optimal?
 - Your opportunities to experiment in-market are limited

Three Key Insurer Challenges High-Level Goals → Individual Decisions



- Competing goals and decisions
 - Growth
 - "We need to grow business in this region by 5% . . ."
 - "Maybe we should lower our underwriting standards?"
 - Risk
 - "We need to keep the combined ratio below 98% . . ."
 - "Maybe we should raise our underwriting standards?"

Speed

- "We need to manually underwrite less than 30% of policies . . ."
- "Maybe we should simplify our underwriting standards?"

Lifetime Value

- "Our retention rate needs to be above 90% . . ."
- "Maybe we should extend our underwriting standards?"
- Market share, distribution, etc.

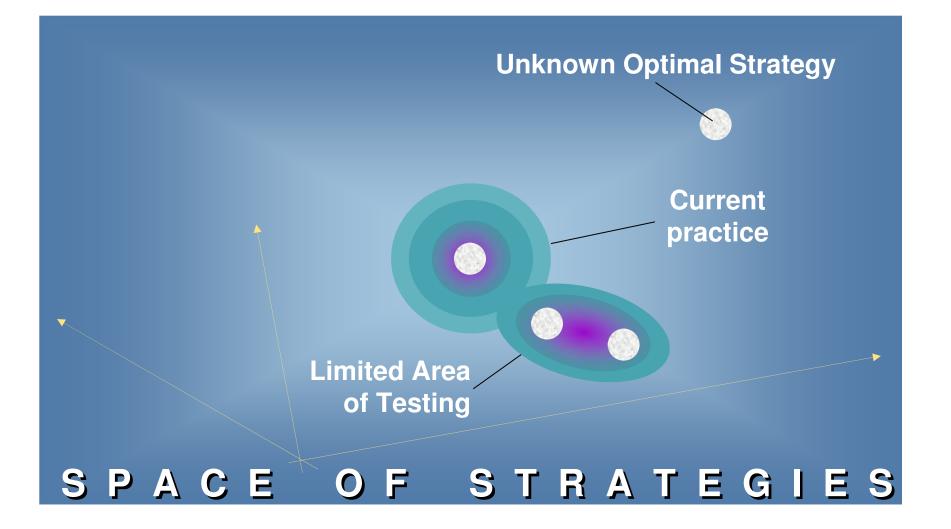
Three Key Insurer Challenges What Customer Is → What Customer Does



- Standard customer models
 - Reflect a static, historical set of circumstances
 - Response and conversion models
 - Limited to specific past offers, channels and interactions
 - Retention models
 - Limited to specific prior customer management behaviors
 - Risk models
 - Limited to specific historical customer product choices
- With different conditions, what would a customer do?
 - Move from customer actions to customer reactions
 - Insurer actions affect consumer reactions

Three Key Insurer Challenges Limited Experiments → Optimal Strategy





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Three Key Insurer Challenges Real-life Questions?

Manage the Risk Portfolio

- How should we manage our distribution system in order to meet strategic objectives?
- How should we underwrite and price in various geographies in order to guarantee a strategic and risk-managed distribution of policies?

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 How should we identify the appropriate offers to make to individual customers to meet both growth and risk objectives?

Improve Customer Profitability

- How can we identify how individual customers will respond to new rates in combination with competitive pricing?
- How can we offer individual customers the "right" package of policies, products and services to increase retention and lifetime value?

Respond To Competition

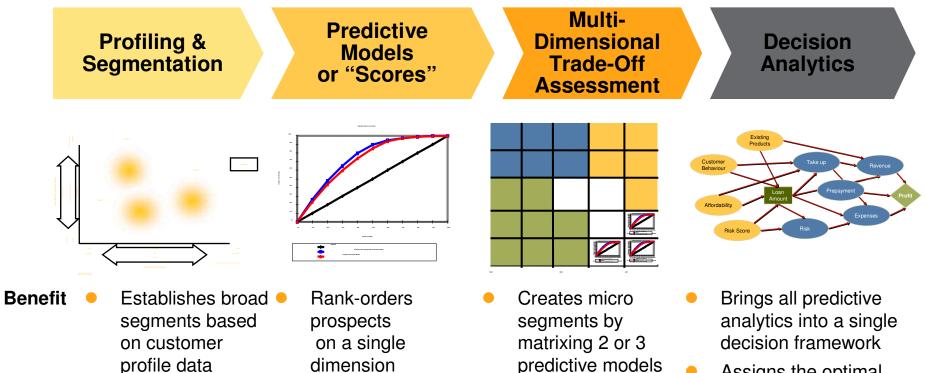
- How can we resist "following" the competition and focus on profitable customers who are tempted?
- How can we communicate to the organization the actions required to offset a competitive lower price?

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Analytical Approaches Have Evolved To Better Meet Insurers' Needs



Assigns the optimal action for each prospect/account given specific business constraints

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Taking predictive analytics and business rules beyond the calculation of a score to the optimization of a decision

Role of Analytics Current State: Prediction of Risk

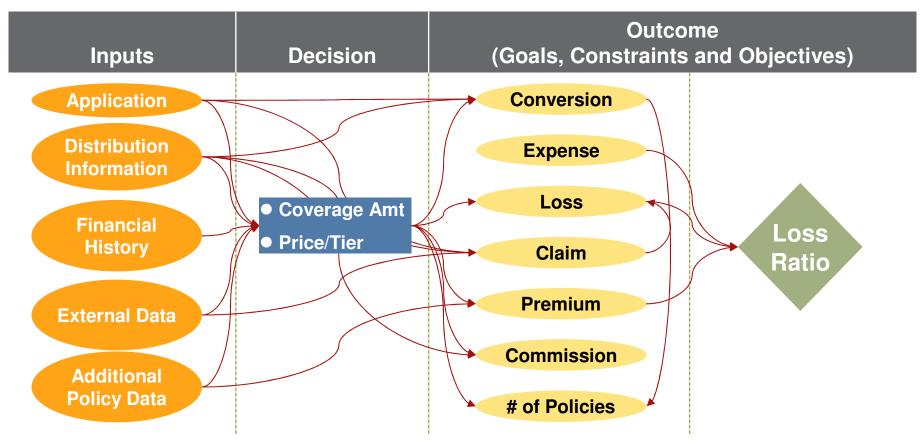




- Limited set of predicted outcomes
- Few degrees of freedom in your decision making

Role of Analytics Beyond Predictions → Decision Analytics

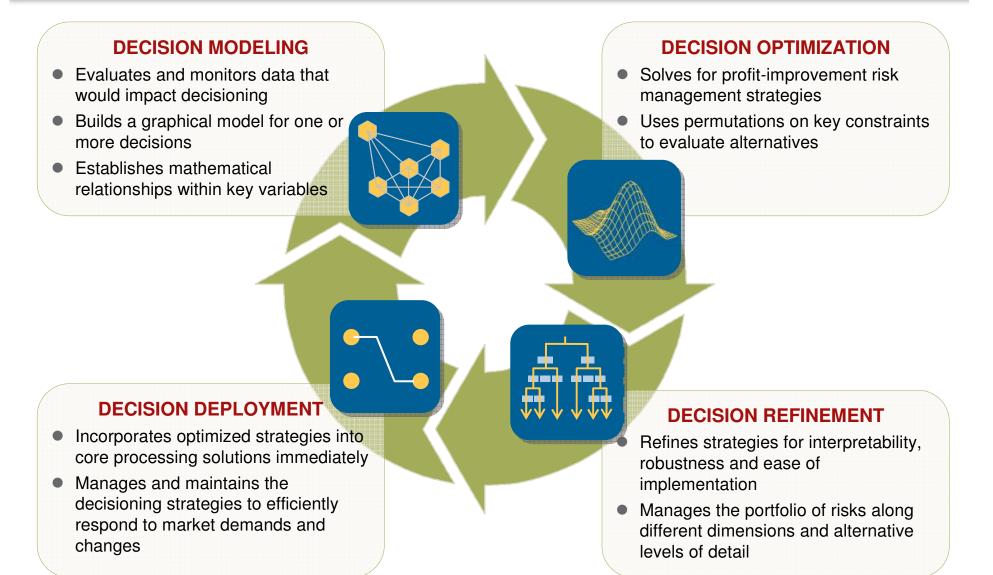




- Explicit modeling of your available decisions
- Broader set of outcomes, constraints and objectives considered

Decision Modeling Four Key Steps





Model The Decision



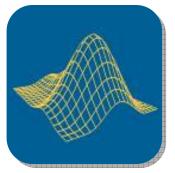


<u>Same</u> customer profile receiving <u>different</u> treatments result in <u>different</u> consequences for your business.

Action		Customer	Reaction
Policy: Coverages A Price Tier 1	•	Insurance Credit Score = 640 # of Drivers = 1 Vehicle type = Sedan Agent #360 1 speeding violation	E(Loss) = \$1000 P(Conversion) = 80% Premium = \$750
-			
■ Policy: Coverages F Price Tier N	•	Insurance Credit Score = 640 # of Drivers = 1 Vehicle type = Sedan Agent #360 1 speeding violation	E(Loss) = \$1200 P(Conversion) = 60% Premium = \$1000

Simulate Decision Strategies Find Optimal Approach



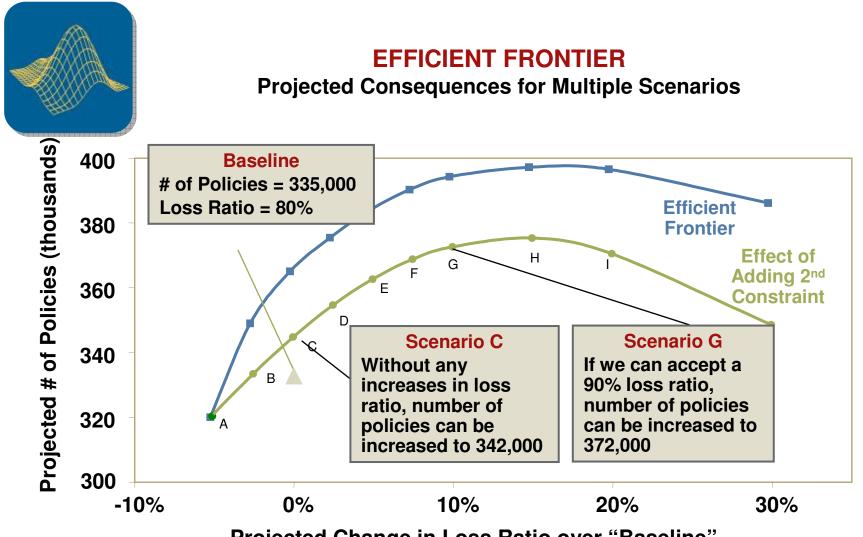


Test business scenarios in your analytic environment <u>before</u> you deploy them in the market.

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		oldLossRatio			Total			56% 9%
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Efficient Frontiers Visibility to Business' Key Metrics

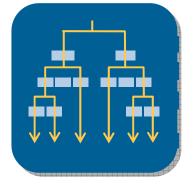


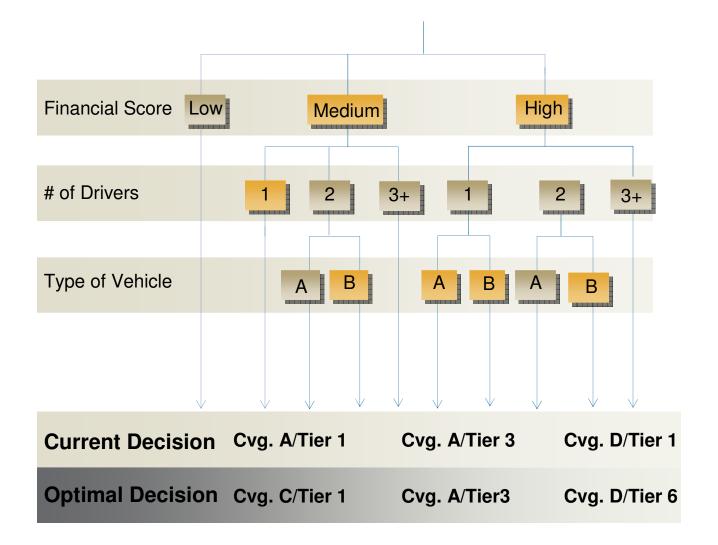


Projected Change in Loss Ratio over "Baseline"

Refine Strategy Implementation Interpretability, Ease and Compliance

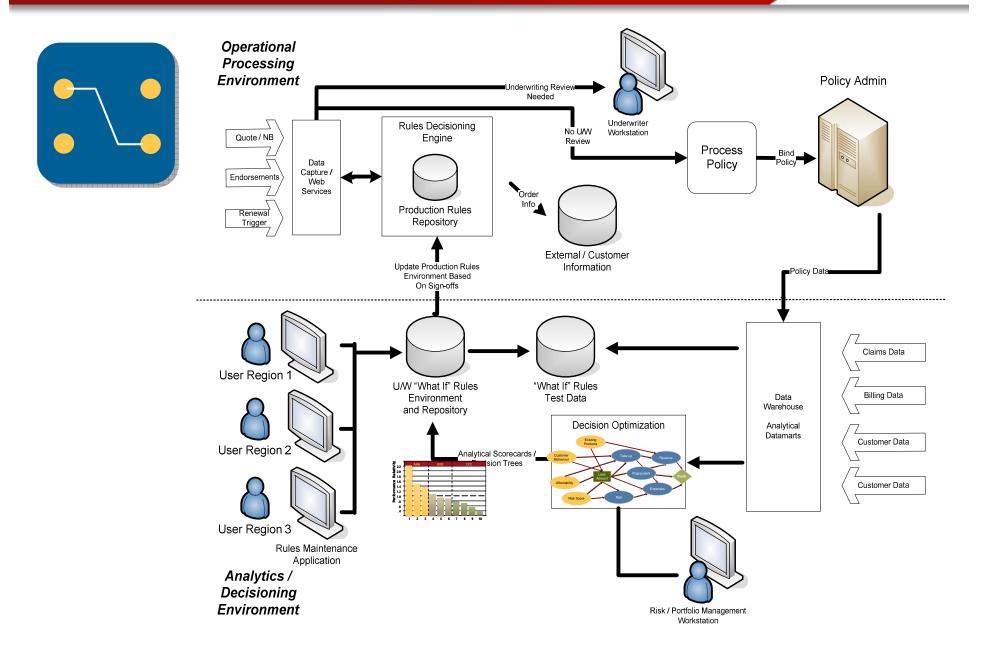






Insurance Decisioning Environment Using Fair Isaac Technologies





Decision Analytics Supporting Technologies



	Technology	Description	Uses
	 Analytics for forecasting future individual behavior 		 Improve risk assessment of customers
	Model Builder• Utilizes analytics to define decision trees that optimize the strategy that has been chosen• Ta op• Im	 Utilizes analytics to define 	 Target marketing opportunities
Modeling, Optimization, and		 Improve use of information in portfolio management 	
Refinement	Decision Optimizer • Simulates offerings to align • Manage the portion	 Design strategies that increase profit, response, other key metrics 	
		decisioning with organization	 Manage the portfolio of risk at a local market or agency level
Deployment	Blaze Advisor	 Software for defining, testing and executing rules, processes and strategies 	 Make instant, consistent decisions in real time, across the enterprise

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Pricing Optimization Address Pricing as Intro Rates Expire



- Client: A top 10 bank and credit card issuer
- Business Problem: Declining balances as intro rates expire Determine optimal marketing offer to retain customers and stimulate balances:
 - Price reduction
 - Line increase
 - Product upsell
- Results:
 - Identify consumers who respond most to price reduction offers while keeping re-priced balances to a minimum
 - 12% increase in purchase activation and 7% increase in profit in 12 months

Pricing Optimization Address Pricing Sensitivity in a Retention Call Center



- Client: A top 5 US credit card issuer
- Business Problem: Inbound retention call center Balance customer retention re-pricing with:
 - Balance Build
 - Yield
 - Risk
- Results: Identify different levels of price sensitivity within segments of customers
 - Understand balance build/profit trade-off
 - Identify opportunities to increase wallet share
 - Increase profitability by nearly \$100 per account in segments with high price sensitivity and yield potential

Pricing Optimization Optimize Installment Loan Offers - Price and Amount



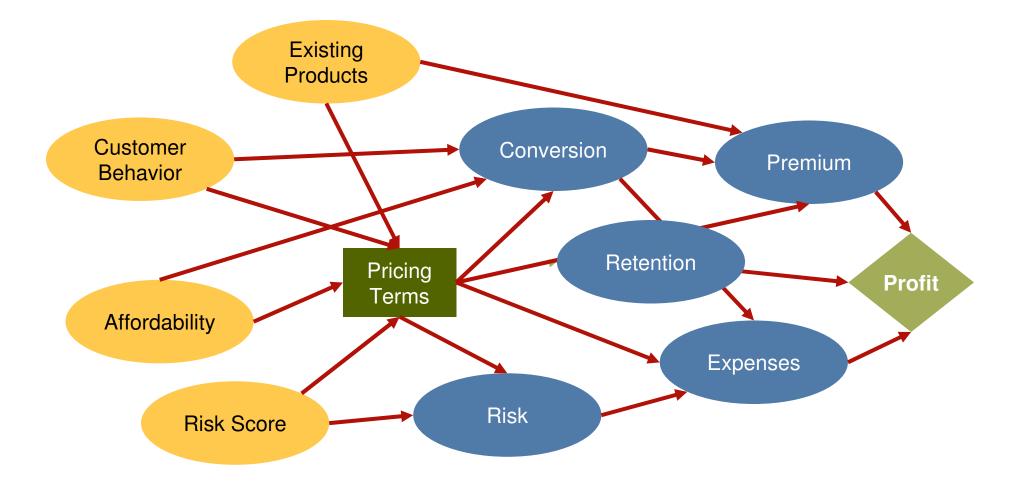
- Client: A full-service retail bank in the UK
- Business Problem: Improve loan profitability by optimizing
 - Who to target for a loan product
 - Loan amount
 - Loan price

While meeting other key business metrics:

- Maintain acceptance rates
- Increase loan take up rates and insurance take up
- Maintain or reduce losses and bad rates
- Meet regulatory requirements on pricing and loan amounts
- Results: 20% profit improvement after 5 months, expected 60% profit improvement over life of loan, while maintaining bad rates, losses and acceptance rates

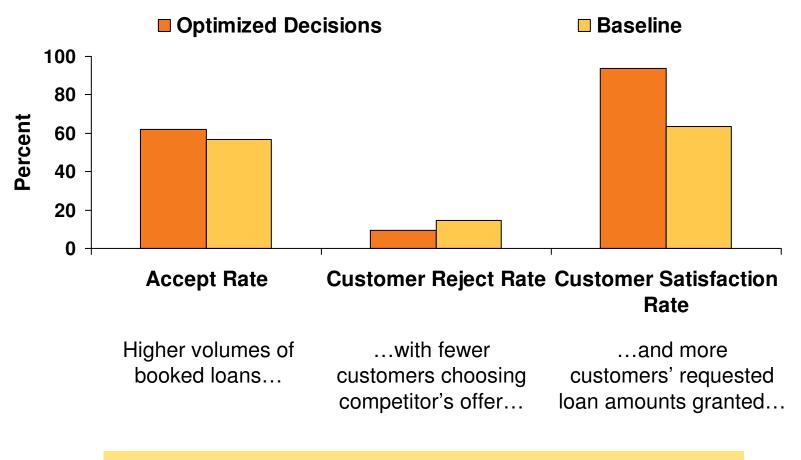
Pricing Optimization Decision Analytics and Optimization





Pricing Optimization Improved Customer Experience





...while, lowering risk

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Demonstration: Develop & Deploy Strategies for Quoting Auto Policies



Situation

- Annual growth in Policies in Force is declining below the 5% goal
- The primary driver of this is low conversion rates for agents in regions that are facing strong competition

Action Plan

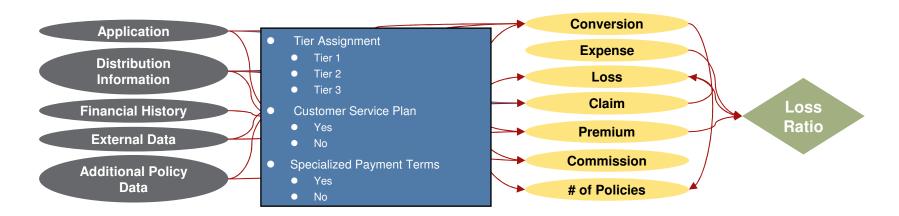
- Focus on new auto policy quotes
- Develop new strategies for assigning prospects to tiers while accounting for local market conditions
- Efficiently allocate a limited number of offers for flexible payment terms and customer service terms to differentiate products and improve conversion

More Competitive Strategies Linking Decisions to Business Goals



Business Goals

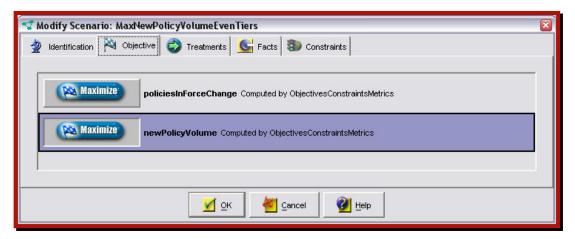
- Grow the number of Policies in Force
 - Keep annual growth above 5%
- Maintain profitability
 - Keep the Combined Ratio within strategic guidelines
- Understand the trade-off between growth in policies and combined ratio
- Decision Alternatives
 - Tier Assignment: Preferred (Tier 1), Standard (Tier 2), Non-Standard (Tier 3)
 - Specialized customer service plan (offer/no offer)
 - Specialized payment terms (offer/no offer)



Demonstration Step 1: Configure Optimization Scenario



Maximize New Policy Volume



Maintain the Combined Ratio

😴 Modify Scenario: MaxNewPolicyVolumeEvenTiers	
🙅 Identification 🏁 Objective 🥏 Treatments 🕵 Facts 🐌 Constraints	
	🕘 Global
	E Local
CombinedRatio <= 0.68	
🗹 QK 🛛 🖉 Cancel 🛛 🚱 Help	

Demonstration Step 2: Optimize the Actions



- For each prospect in the historical data
 - Simulate the effect of all potential actions on the new policy volume and combined ratio
 - Pick the best decision

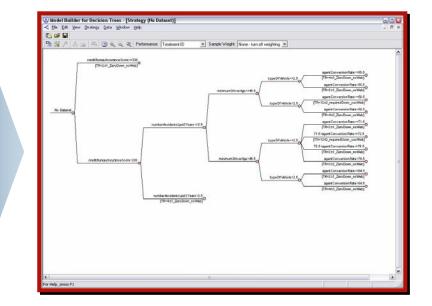
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Demonstration Step 3: Implement Rules into Operation



- Finalize Decision Management Optimization
- Deploy Optimized Decisions
 Through Blaze Advisor Rules

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Conclusions



- In order to compete, insurers must be able assess individual decisions in relation to larger strategic goals
 - Multiple objectives, constraints and metrics
- Decision Analytics extends the evolution of analytic techniques
 - Predictive analytics allows for strong control of metrics
 - Decision analytics allows for making tradeoffs across multiple goals and constraints in order to create value
- Make more informed decisions around strategic goals
 - Brings greater visibility, understanding and control to the decision
 - Use the Efficient Frontier to make clear business trade-offs
 - Communicate those goals and what they mean to the organization
- Other potential decision areas
 - Marketing, Policy Management, Channel Distribution, etc.



THANK YOU

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