Improving and
Optimizing Your
Offerings with Conjoint
Analysis Market
Simulators



Sawtooth Software

## We Simulators

#### We Love Simulators

- They turn unfamiliar "utilities" into demand predictions, revenue estimates, etc.
- ▶ The show attribute importance with an added competitive context
- They are the gateway to further estimates like price elasticity, willingness to pay, etc.

#### This is not a webinar on simulator basics

(been there, done that)

https://sawtoothsoftware.com/resources/events/webinars/us ing-the-market-simulator

or

https://sawtoothsoftware.com/resources/events/webinars/intro-to-choice-based-conjoint-with-lighthouse-studio-part-2

### Today we're talking about finding improvements through <u>recommendations</u> and <u>optimizations</u>

#### What is a recommendation?

- ► Recommendations are essentially some sorting after running a sensitivity analysis (if we were trying to boost stock price we'd call it AI)
- ► Recommendations are available in our online platform Discover and it's standalone compliment app.sawtoothsoftware.com

> Standard market simulations give choice predictions for a specific scenario

	Product 1 Product 2		Product 3	
Brand	Visa	MasterCard	Discover	
Interest Rate	15% interest	20% interest	15% interest	
<b>Credit Limit</b>	\$2,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

Sensitivity holds the scenario constant and systematically steps through levels of an attribute

	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Interest Rate	15% interest	20% interest	15% interest	
<b>Credit Limit</b>	\$2,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

Sensitivity holds the scenario constant and systematically steps through levels of an attribute

	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Interest Rate	15% interest	20% interest	15% interest	
Credit Limit	\$2,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

Sensitivity holds the scenario constant and systematically shuffles through levels of an attribute

	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$2,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

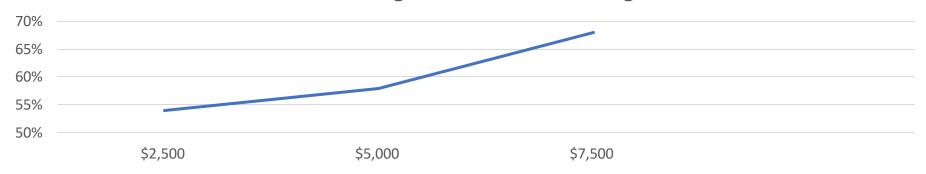
	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$5,000 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$7,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	37%	

Share

58%

Share Change as Credit Limit Changes



	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$2,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	54%	8%	38%	

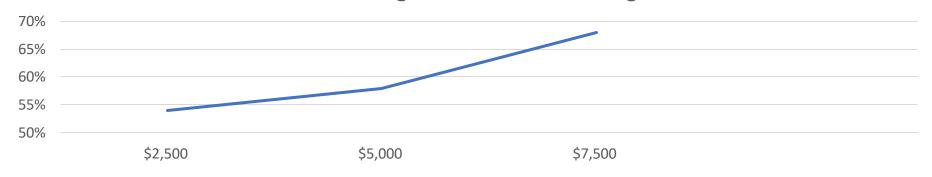
	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$5,000 credit limit	\$5,000 credit limit	\$7,500 credit limit	

6%

36%

	Product 1	Product 2	Product 3	
Brand	Visa	MasterCard	Discover	
Intere st Rate	15% interest	20% interest	15% interest	
Credit Limit	\$7,500 credit limit	\$5,000 credit limit	\$7,500 credit limit	
Share	68%	4%	28%	

Share Change as Credit Limit Changes



- ▶ Recommendation #1: Change credit limit to \$7,500 for a 14% increase in share
- ► Recommendation #2: Change credit limit to \$5,000 for a <u>4% increase</u> in share

#### Software Demo

#### Recommendations Versus Optimization

- ► Recommendations step through each one-way shift from the current configuration, one attribute at a time
- Not guaranteed to find the overall optimal configuration (need to shift around multiple attributes at the same time)

#### Recommendations Versus Optimization

▶ In theory, we could just do exhaustive sensitivities

1-1-1-1

1-1-3-1

1-1-1-2

1-1-3-2

1-1-1-3

1-1-3-3

1-1-2-1

And so on

1-1-2-2

1-1-2-3

#### Algorithms for optimizations

#### Exhaustive

Simple; Examines all possible combinations; Guaranteed to find the global optimal solution; Can conduct Multi-Objective Searches

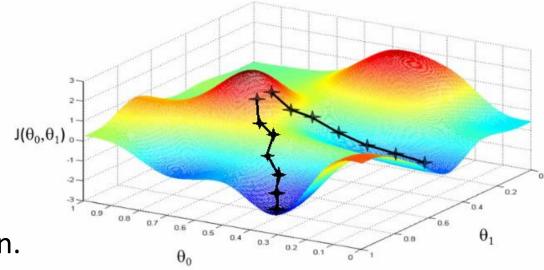
#### **Enormous Search Space?**

- ▶ Suppose we have 10 attributes, each with 5 levels.
- ▶ There are then 5<sup>10</sup> possible product configurations, or almost 10 million!
- ▶ If optimizing multiple products simultaneously, the problem gets even bigger. (9,765,625 \*9,765,624\*...)
- ► For larger optimization problems, even the fastest computers could take hours, days or weeks to evaluate all possible combinations

#### Grid Algorithm

▶ It isn't necessary to try all possible combinations to find really good solutions

► Grid changes one attribute at a time (holding all others constant) and keeps any change that improves the solution. This repeats until no other change results in a better solution.



▶ If the response surface is single-peaked, it is guaranteed to find the global optimum.

#### Algorithms for optimizations

#### Exhaustive

Simple; Examines all possible combinations; Guaranteed to find the global optimal solution; Can conduct Multi-Objective Searches

#### Grid

Extremely fast if search space is large; Accurate if search space is single-peaked; Used to reduce Exhaustive search domain

#### What About a Middle Ground?

- ▶ Do we need to test all possible combinations?
  - Some levels are just plain bad
- ▶ Can we avoid getting stuck on smaller hills?
  - ► Add some elements of randomness

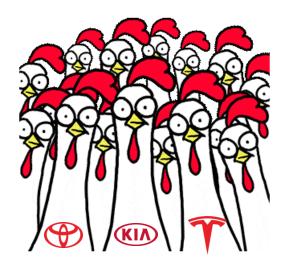
# YASSASSASS



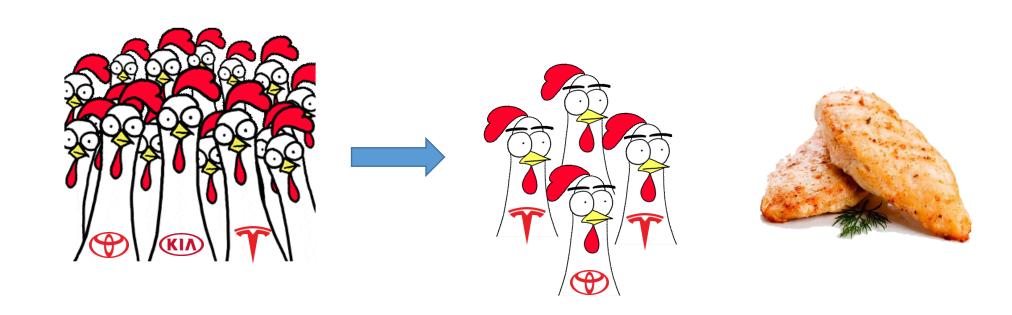


▶ Based on concepts of evolutionary biology and Darwinian theory (survival of the fittest)

- ▶ Step 1: Create the herd
  - ► Generate 300 product configurations (randomly, grid, etc.)



- ▶ Step 2: Release the predators
  - ▶ Top 150 "fittest" members of the herd survive (e.g. highest share)



- ▶ Step 3: Survivors mate and produce next generation
  - ▶ Fittest survivors have a higher probability of producing offspring with other fittest survivors
  - Mutations can occur

	A1	A2	A3	A4	A5
Parent "A"	Tesla	2 door	350 miles	4 year	\$55,000
Parent "B"	Kia	4 door	350 miles	3 year	\$55,000
Offspring "C"	Tesla	4 door	350 miles	4 year	\$65,000
				Muta	<b>7</b> ation
Cross-Over					

▶ Rinse and repeat until several generations produce no significant gains (or a set number of iterations is run)



#### Algorithms for optimizations

#### Exhaustive

Simple; Examines all possible combinations; Guaranteed to find the global optimal solution; Can conduct Multi-Objective Searches

Total search space can be enormous (10 attributes w/ 5 levels each makes for 5^10, or 10 million combinations!)

#### Grid

Extremely fast if search space is large; Accurate if search space is single-peaked; Used to reduce Exhaustive search domain

Not guaranteed to find the global optimal solution if several peaks

#### Genetic

Faster than Exhaustive if search space is large (but longer than Grid); Finds a variety of near-optimal solutions, and most times the single best optimal solution; Can conduct Multi-Objective Searches

Still not guaranteed to find the global optimal solution, but Genetic is less vulnerable than Grid search to finding the local optimum

#### Software Demo



Sawtooth Software

+1 801 477 4700 sawtoothsoftware.com support@sawtoothsoftware.com

THANK YOU

Questions?