

# Total addressable market

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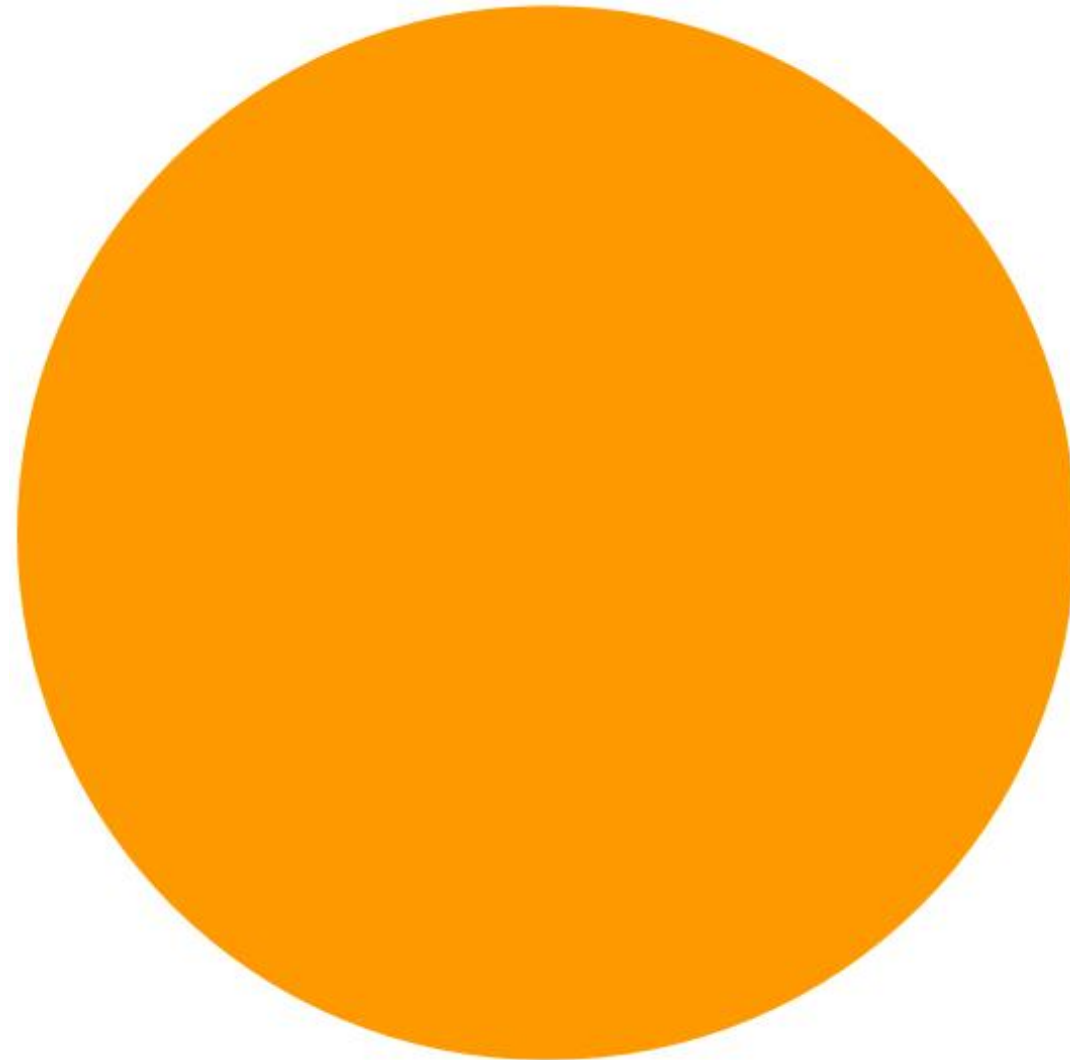
**Ted Kwartler**  
Data Dude

# Total Addressable Market (TAM)



# TAM: What is the entirety of the market?

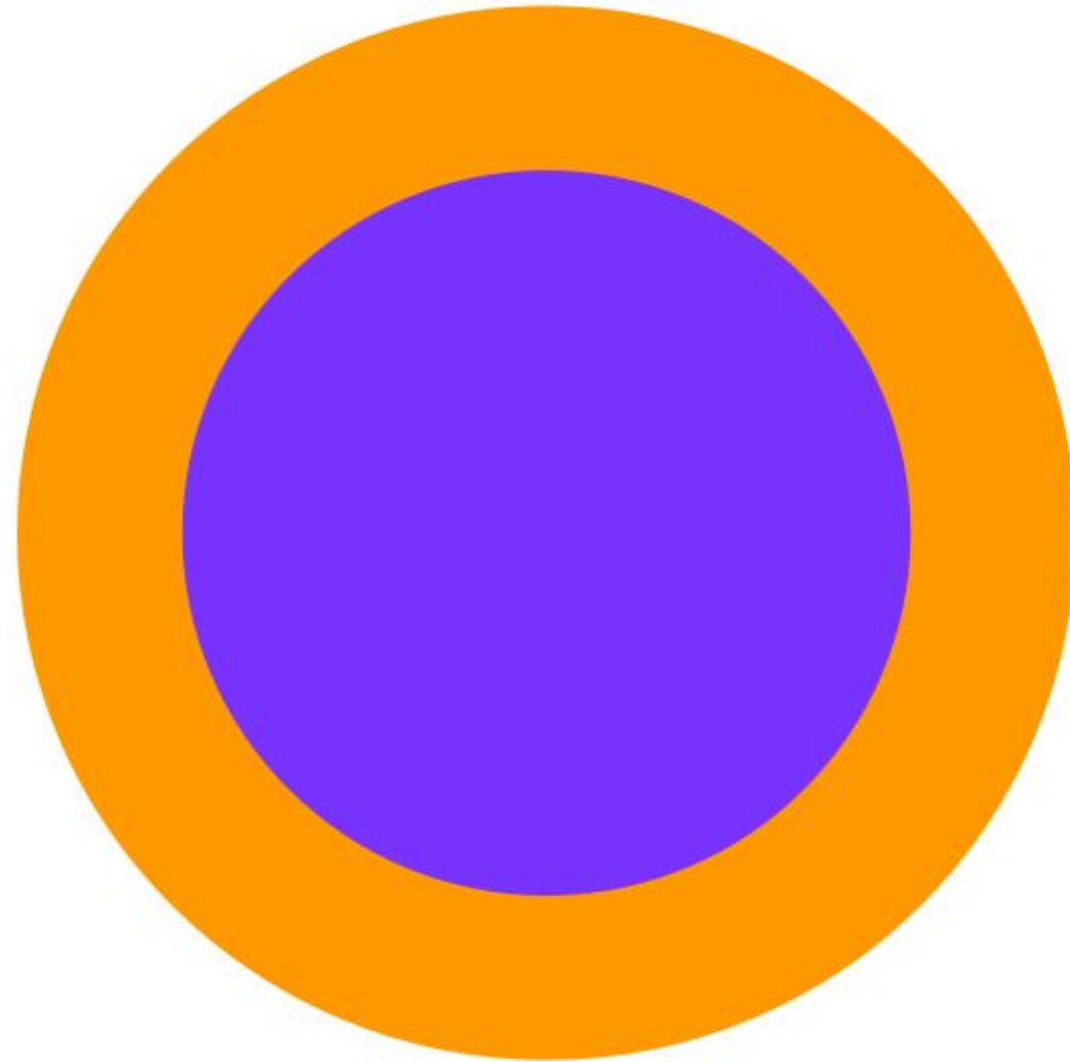
Total Addressable Market (TAM)





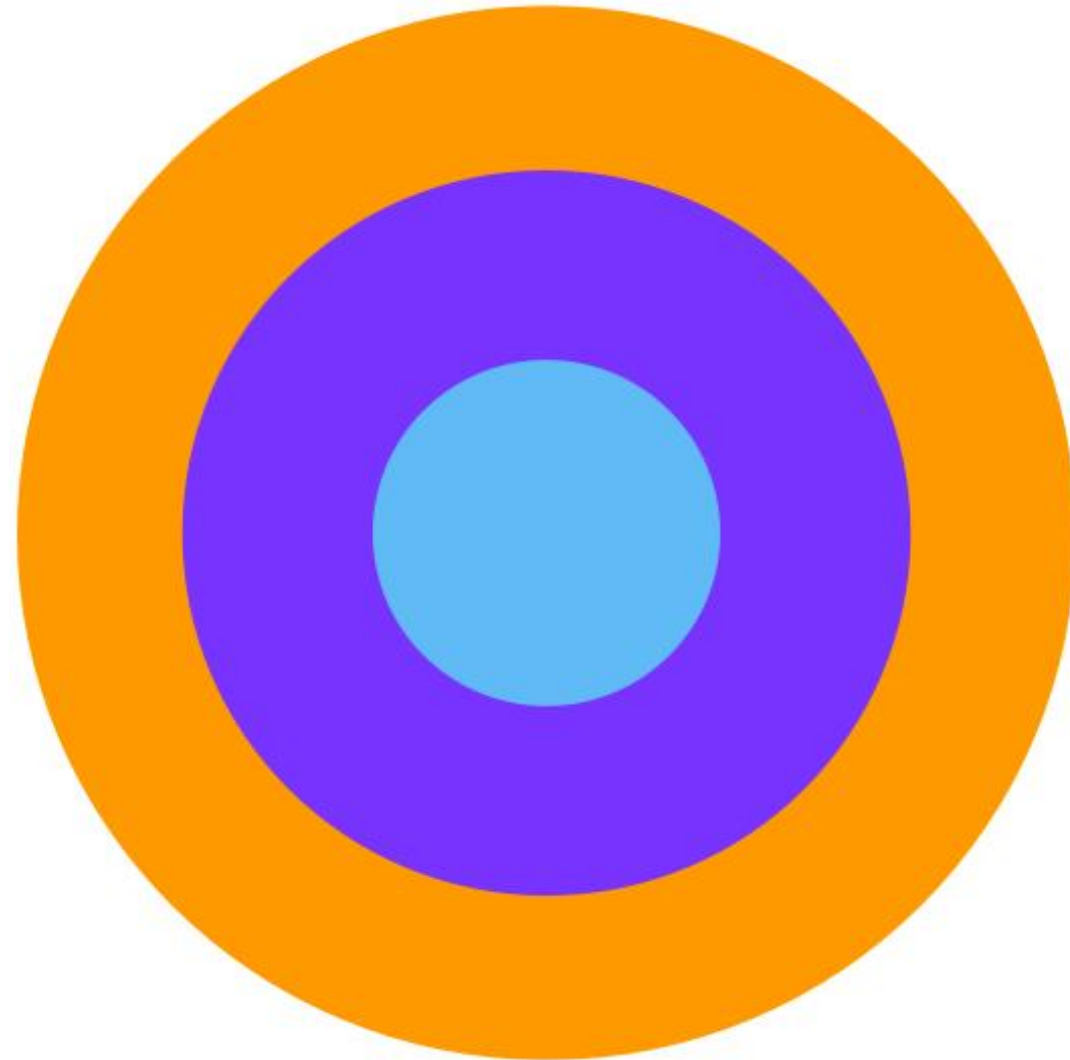
# SAM: What part of the market can your company get?

Serviceable Available Market (SAM)



# SOM: What proportion of market can you reach?

Serviceable Obtainable Market (SOM)



# Calculating TAM

- "Swag" math
  - Inputs: best estimates
  - Data-driven decision to enter the market
- Two popular methods:
  - Top-down
  - Bottom-up

# TAM: Top-down market sizing

## Top-down TAM

1. Start with macro data set
2. Logically apply interactions

### *Example*

```
330 US Citizens
* 10.5% diabetics
* .75 bars per person
* 100% margin ($1)
= $25.25M TAM
```

## Data points

- [U.S. and World Population Clock](#)
- [The Most Popular Candy Bars in America](#)
- [Typical Retail Markups on Candy](#)

# TAM: Bottom-up market sizing

## Bottom-up TAM

1. Start with granular data
2. Logically extrapolate upwards

### *Example*

\$8B annual revenue  
/ 63 candy bars = \$127m

\$127m \* 10% of revenue is sugar free \*  
100% margin(\$1)  
= \$12.7M TAM

## Data points

- **The Hershey Company**
- **List of products manufactured by The Hershey Company**
- **Price increases, sugar-free innovation on tap from Hershey**



# TAM to the moon!

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# Demand curve: meals and drinks

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# Understanding demand, helps you ensure supply

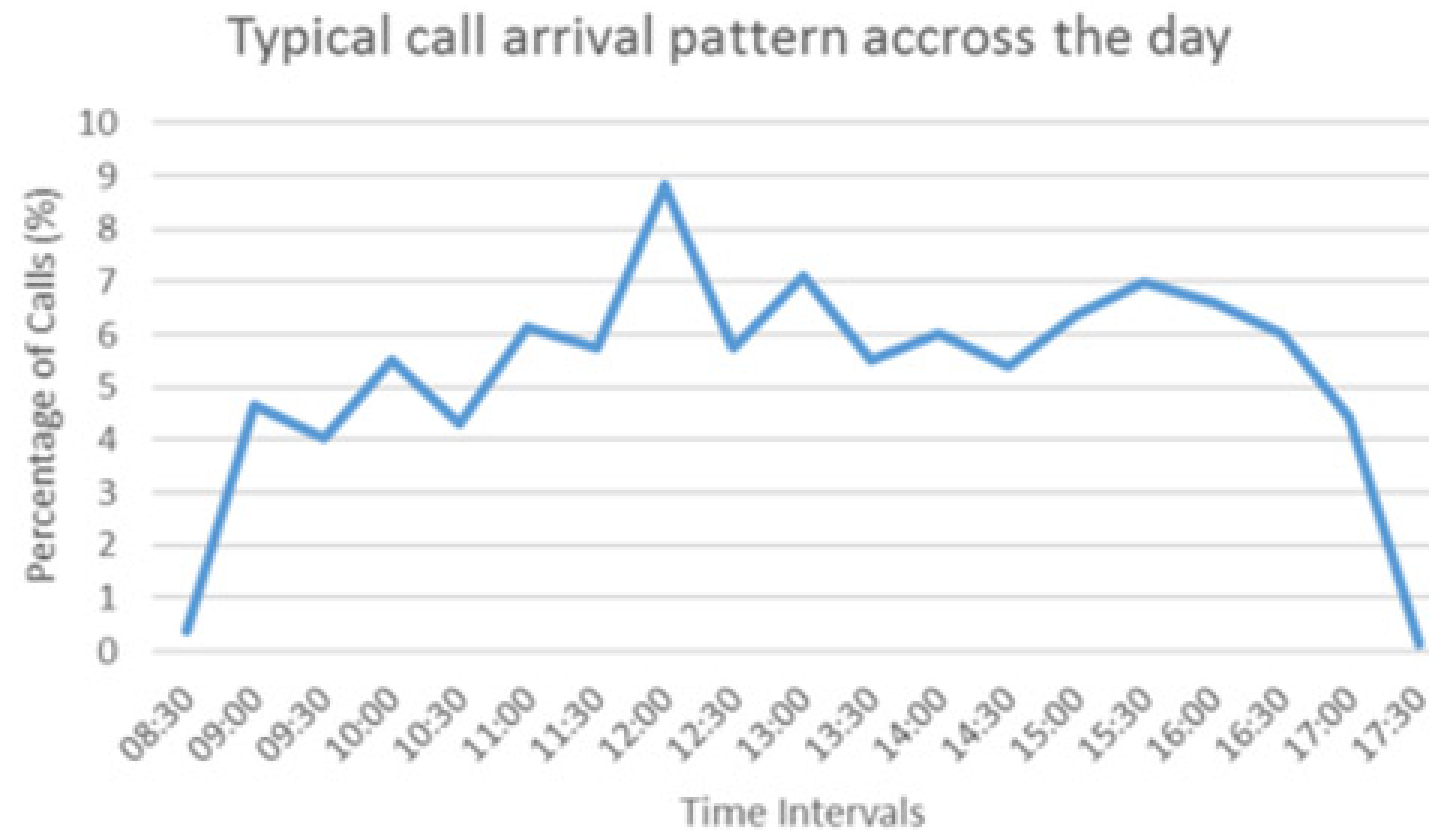
Resource supply



Resource demand

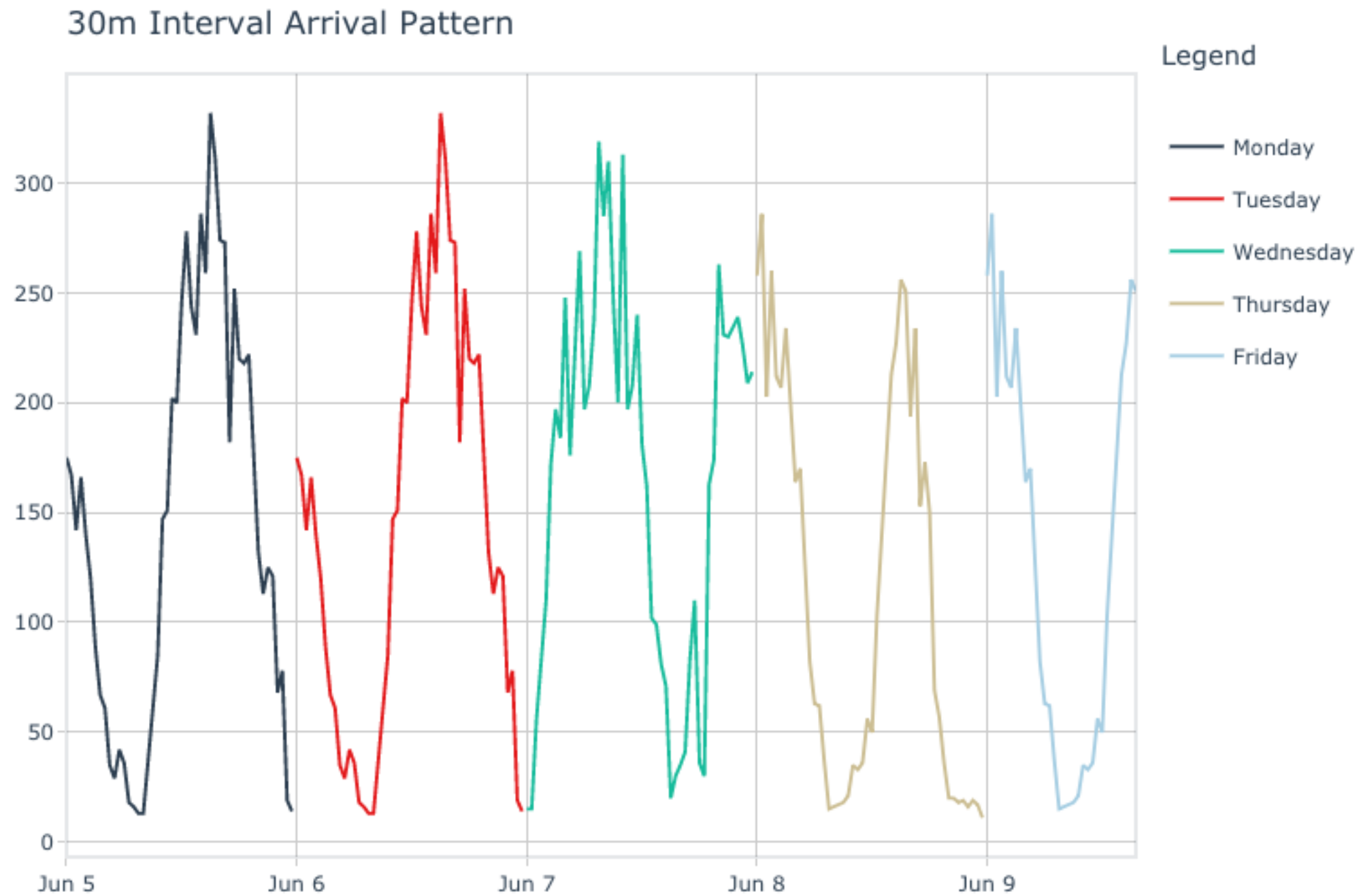


# Time series arrival patterns



- x-axis is the time periodicity
- y-axis is a volume to be measured

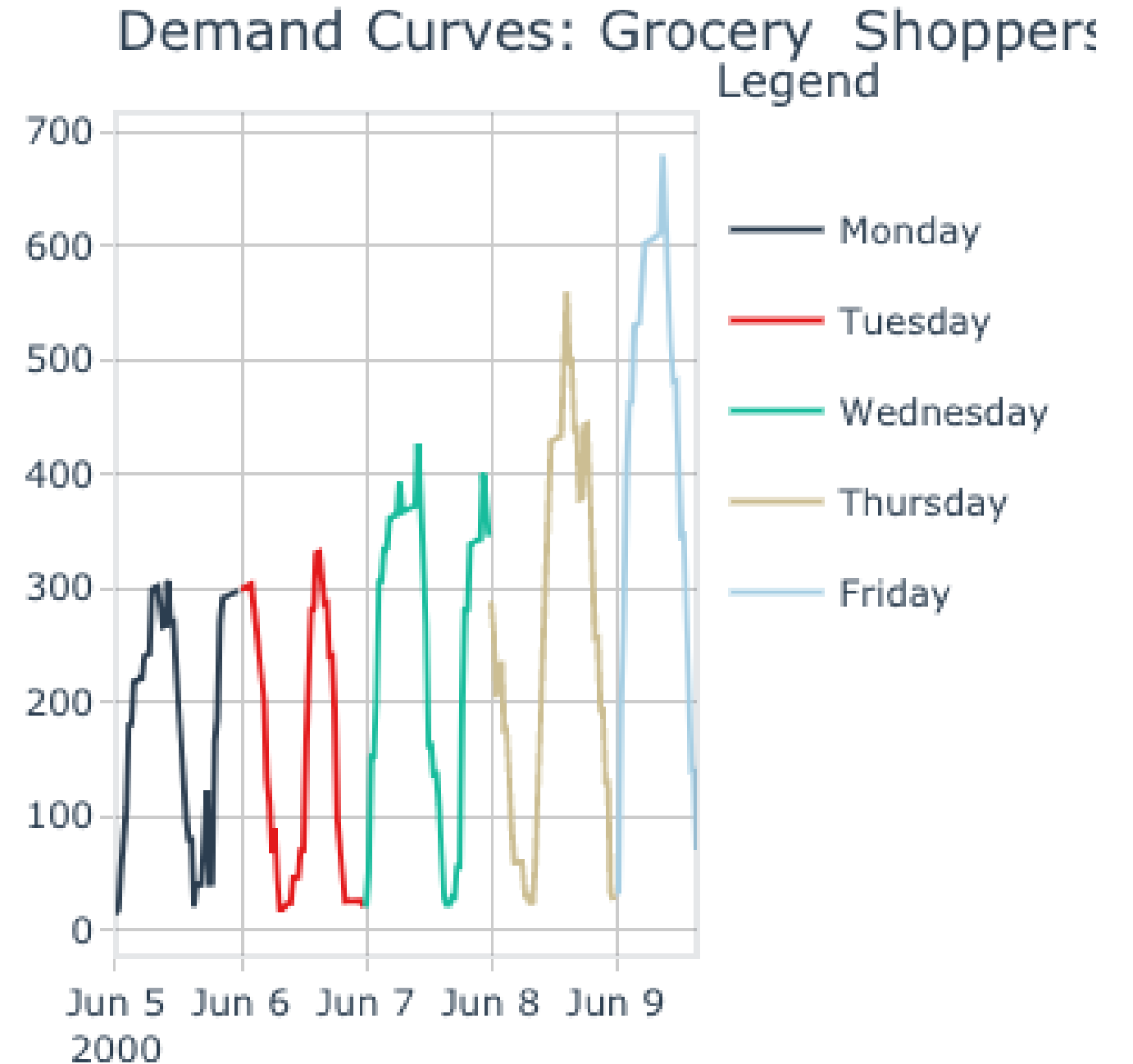
# A time series chart





# Supply and demand

- Line chart represents the demand
- Create a corresponding supply



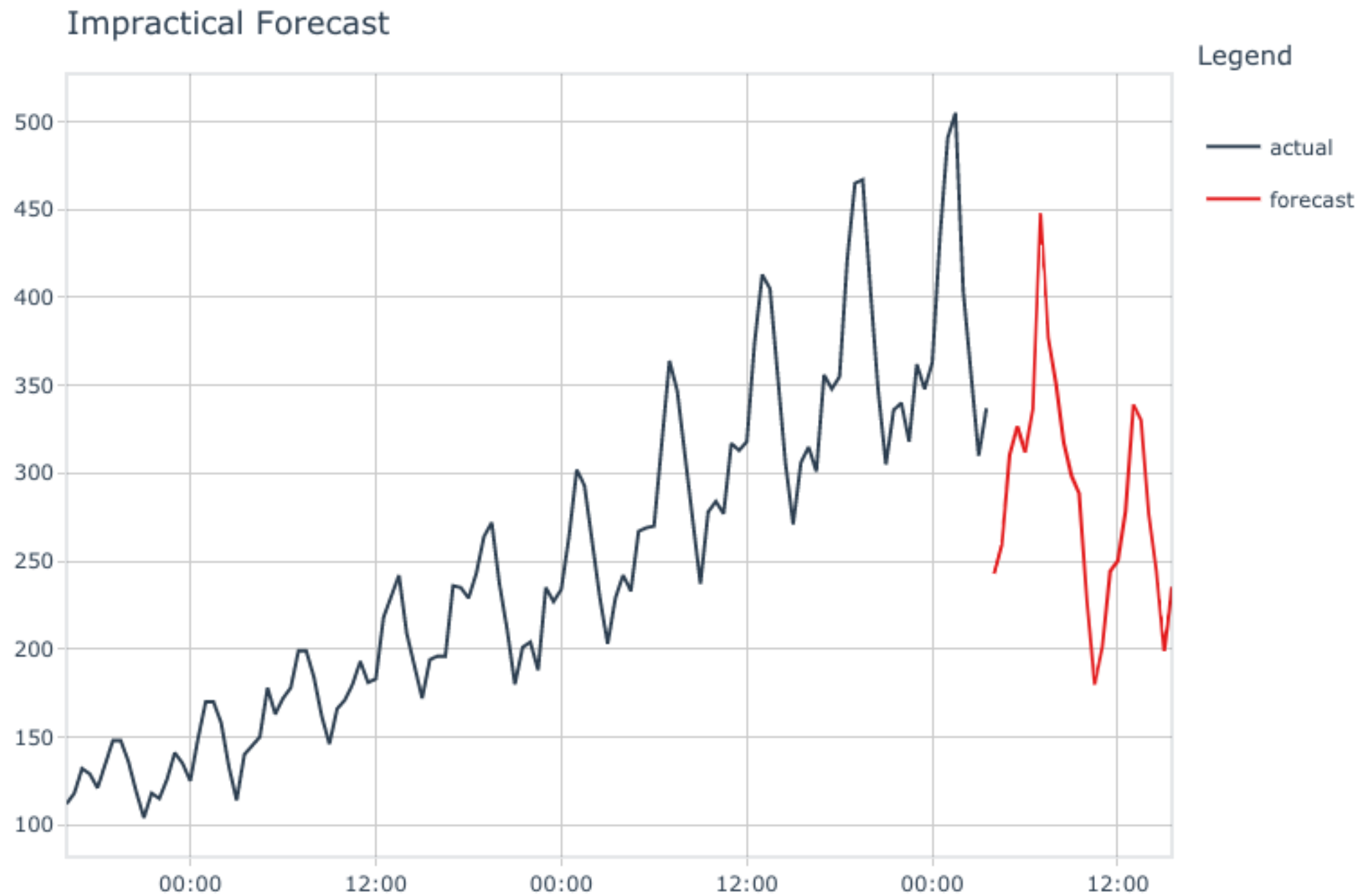
# Use common sense

- Data scientists use the arrival pattern to predict future demand
- Mistakes can occur

## Common pitfalls

- Inconsistent trend compared to business knowledge
- Decimal points when integers are needed
- Negative numbers that don't apply in the real world

# A questionable forecast



# Let's practice!

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# Supply curve: servers for your restaurant

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# Determine an appropriate supply to meet the demand

## *Approach 1*

### Inputs

- 100 calls in 30 minutes
- 5-minute transaction time

### Extreme outputs

- 100 agents
  - no customers wait
  - 5 minutes of work per employee
  - 25 minutes of idle employees

# Determine an appropriate supply to meet the demand

## *Approach 2*

### Inputs

- 100 calls in 30 minutes
- 5 minute transaction time

### Extreme outputs

- 1 agent
  - 94 customers wait
  - 30 minutes of work per employee, non-stop work
  - Some customers would just hang up!  
Missed opportunity!

# Erlang-C formula to the rescue

- Four inputs
- Returns the number of agents/servers/workers needed

Input	Definition	Example
Rate	How much demand is occurring?	100 calls per 30min
Duration	How long to complete 1 unit?	5 minutes per call
Target	How long can work wait?	People can wait <i>up to</i> 20 seconds
Service goal	What percentage of work units need to be serviced under target?	80% of calls will be serviced under 20 seconds

# Math? Ain't nobody got time for that

$$P_w = \frac{\frac{A^N}{N!} \frac{N}{N-A}}{\left( \sum_{i=0}^{N-1} \frac{A^i}{i!} \right) + \frac{A^N}{N!} \frac{N}{N-A}}$$

# Balancing incoming calls, wait time, and agent occupancy

## Inputs

- 100 calls in 30 minutes
- 5 minute transaction time
- People can wait up to 20 seconds
- 80% of calls need to be answered within 20 seconds

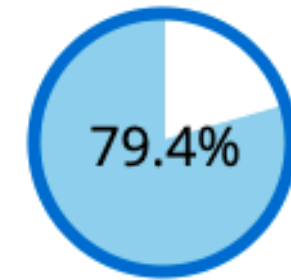
## Results



Agents



20 Seconds



Occupancy



Calls



# Hurry up and wait

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# Customer input to improve your operation

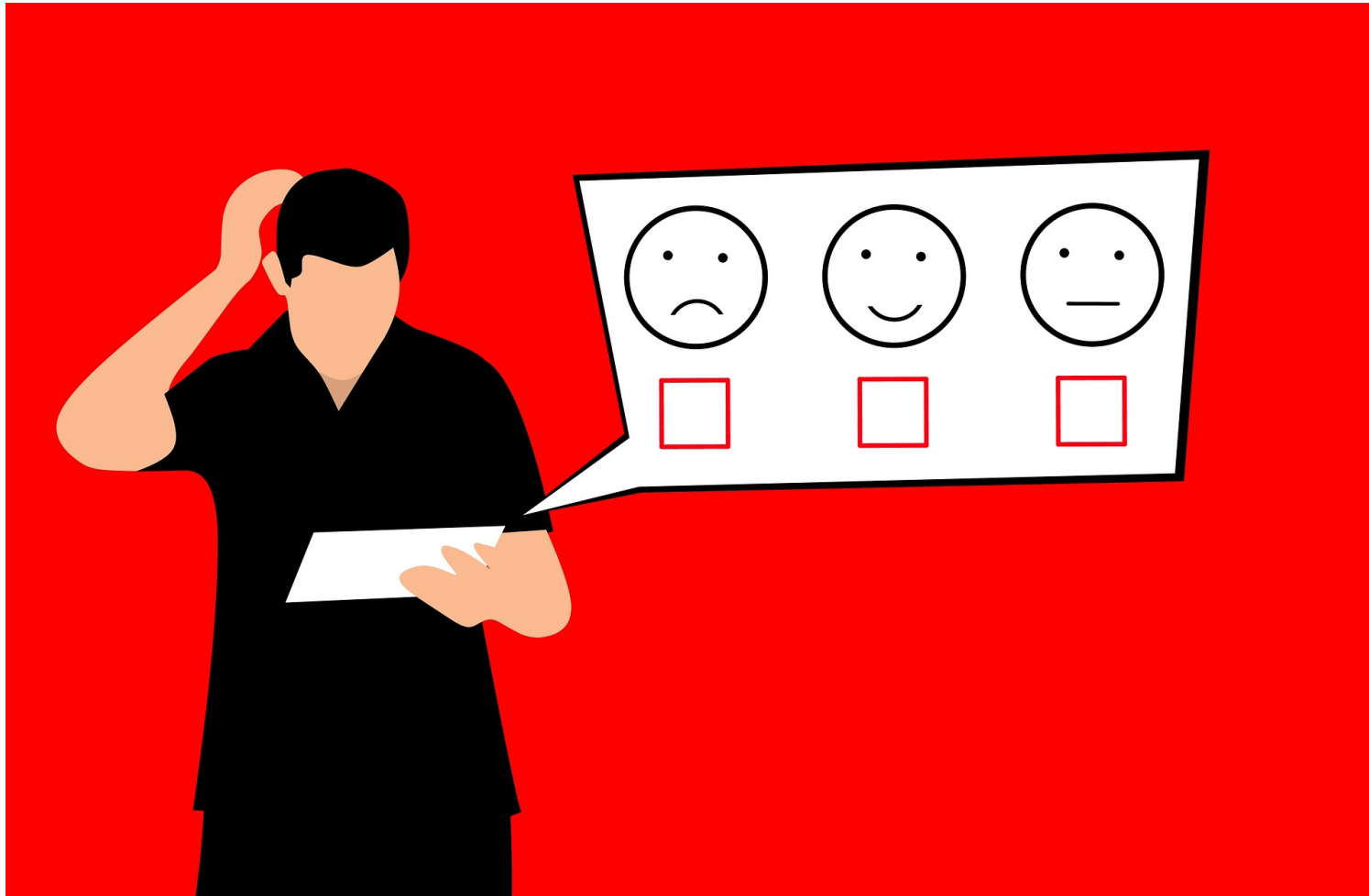
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# Examine customer survey data

- Data collection: qualitative interviews, online reviews, transactional data
- Use model to explain customer survey data



# Survey inputs as a model

- Q1: *How satisfied are you overall [1-5]?*
- Q2: *How do you rate the quality of the product/service [1-5]?*
- Q3: *How do you rate the product/service options [1-5]?*
- Q4: *Do you agree that the offering is fairly priced [1-5]?*

	Q1 (overall satisfaction)	Q2	Q3	Q4
Customer-1	1	5	5	5
Customer-2	1	4	5	5
Customer-3	0	1	3	1
Customer-N	1	1	4	4
...	...	...	...	...

# Survey inputs as a model

	Q1 (overall satisfaction)	Q2	Q3	Q4
Customer-1	1	5	5	5
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Customer-3	0	1	3	1
Customer-N	1	1	4	4
...	...	...	...	...

- Target variable
  - Q1
- Explanatory variables
  - Q2, Q3, Q4



# Explanatory models from customer data

	Q1 (overall satisfaction)	Q2	Q3	Q4
Customer-1	1	5	5	5
Customer-2	1	4	5	5
Customer-3	0	1	3	1
Customer-N	1	1	4	4
...	...	...	...	...

## Logistic regression model

$$f(\text{overall satisfaction}) = \beta_1 * Q2 + \beta_2 * Q3 + \beta_3 * Q4$$

# Sum of betas to understand impact

## Logistic regression model

$$f(\text{overall satisfaction}) = \beta_1 * Q2 + \beta_2 * Q3 + \beta_3 * Q4$$

## Model output

$$f(\text{overall satisfaction}) = 0.25 * Q2 + 0.25 * Q3 + 1 * Q4$$

## Sum of betas

	Beta	Sum of beta	Proportion
Q2	.25	0.6	.42
Q3	.25	0.6	.42
Q4	.1	0.6	.16

# Adding context with frequency

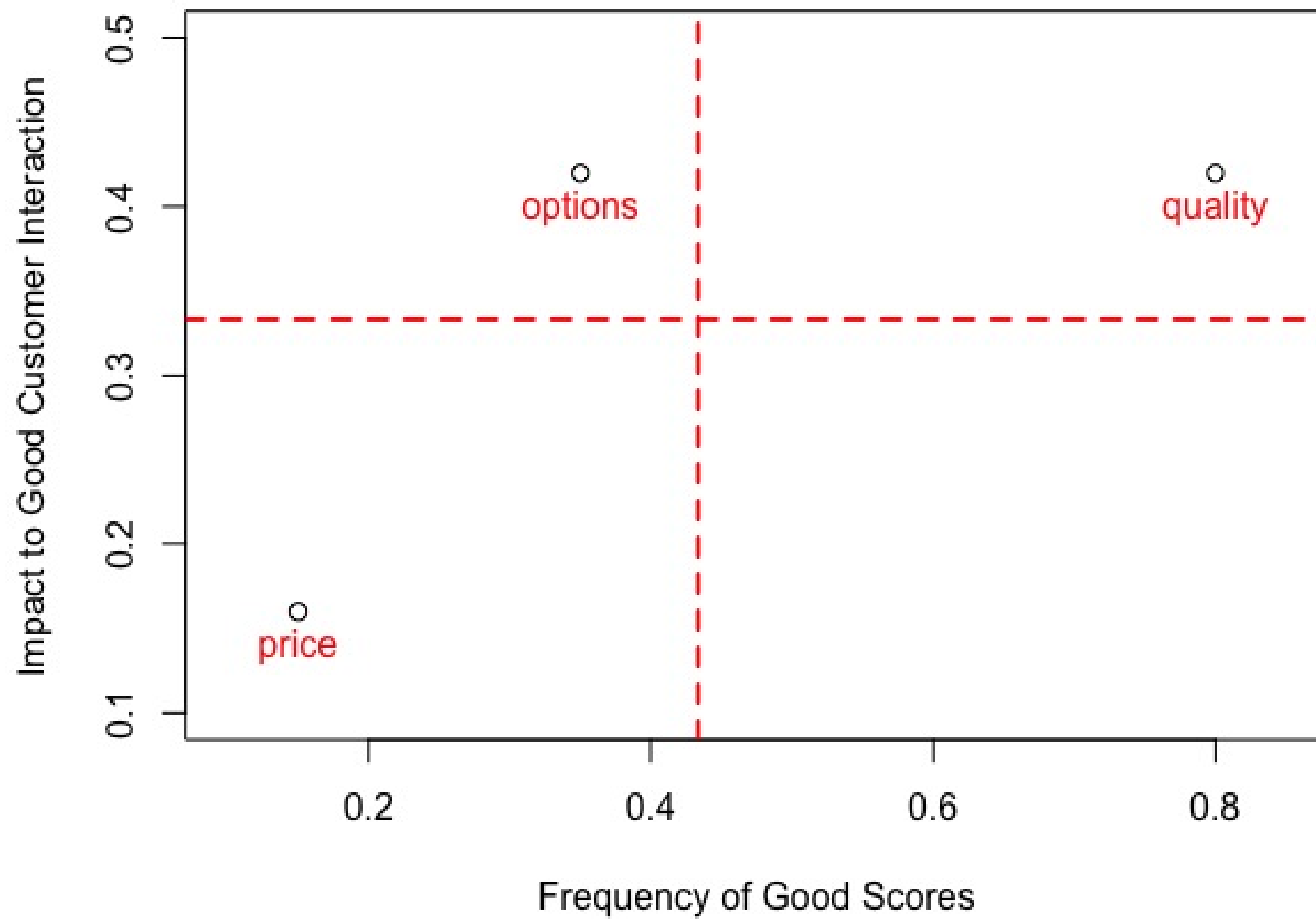
	Beta	Sum of beta	Proportion
Q2	.25	0.6	.42
Q3	.25	0.6	.42
Q4	.1	0.6	.16

# Adding context with frequency

Adding the context of how often the organization does well in a category

	Beta	Sum of beta	Proportion	Frequency of a high score
Q2	.25	0.6	.42	.8
Q3	.25	0.6	.42	.35
Q4	.1	0.6	.16	.15

Sum of Betas Plot



# Off to it!

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# Wrap up

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# Framework for data-driven decisions.

	Cost vs benefit	Risk vs reward	Supply vs demand
Exploratory			
Explanatory			
Predictive			



# Apply your new found knowledge!

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