

Risk Mitigation at Call Centers

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Introduction

- Since 1980, the United States has experienced **218** weather and climate disasters.
- In 2017, across the U.S. there were **15** weather and climate events that resulted in material and financial losses that exceeded **\$1 billion** each
- The annual average of climate disasters has **doubled** in the last five years.



Image adapted from 2017 Weather and Climate Disasters in the US. (<https://www.ncdc.noaa.gov/billions/>)



Photo: Hurricane Harvey 2017

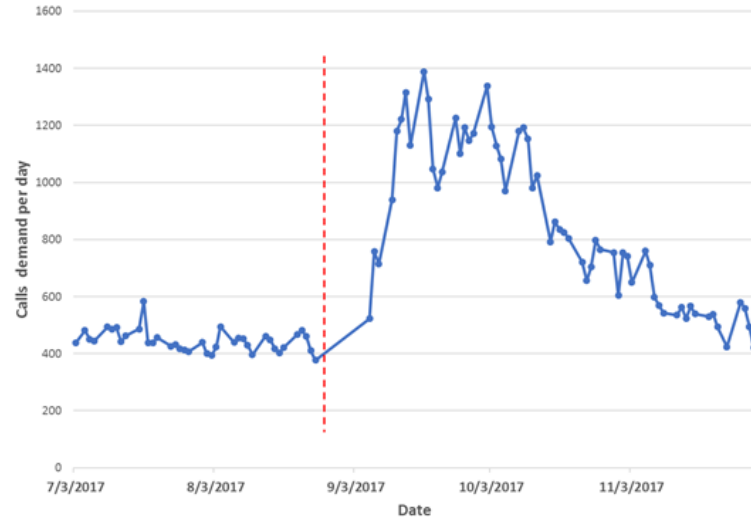
The Company

- Manages an operation that helps sellers connect with buyers of product A through ecommerce site
- Over **150 physical locations** across the US, where Company X conducts **storage, distribution** and **call center** operations relating to the transfer of product A
- Call center operation handles inbound and outbound calls
- Target Service Level Agreement (**SLA**) to respond to **incoming calls** in under **60 seconds**

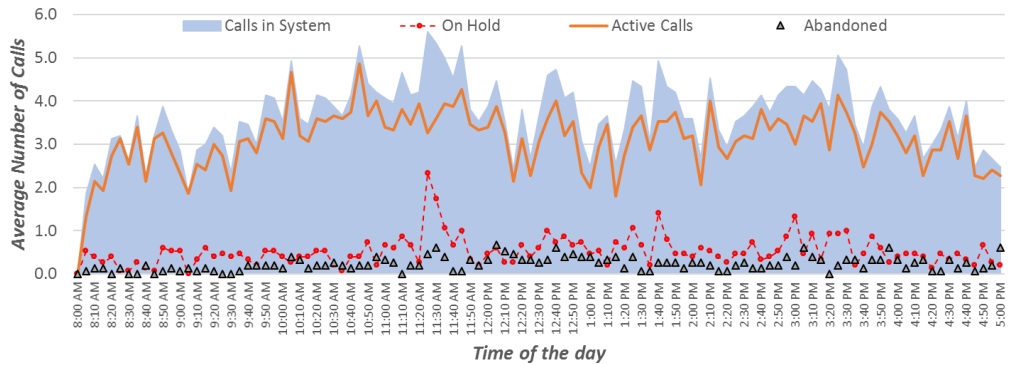
Motivation



How can a company **leverage resources** from a network of **call centers** to accommodate during a disruption, such as a **climate catastrophe event**?



Before Catastrophe Event*



Increase in

calls

141%

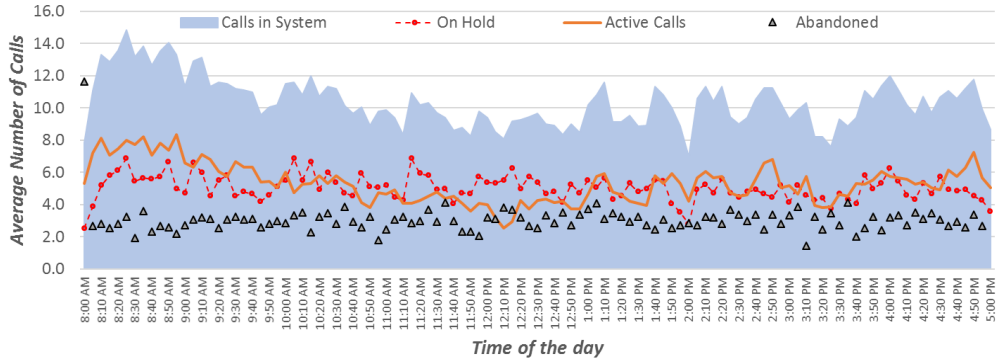
Waiting Time

32 sec → **143 sec**

Drop Calls

4.3% → **28.9%**

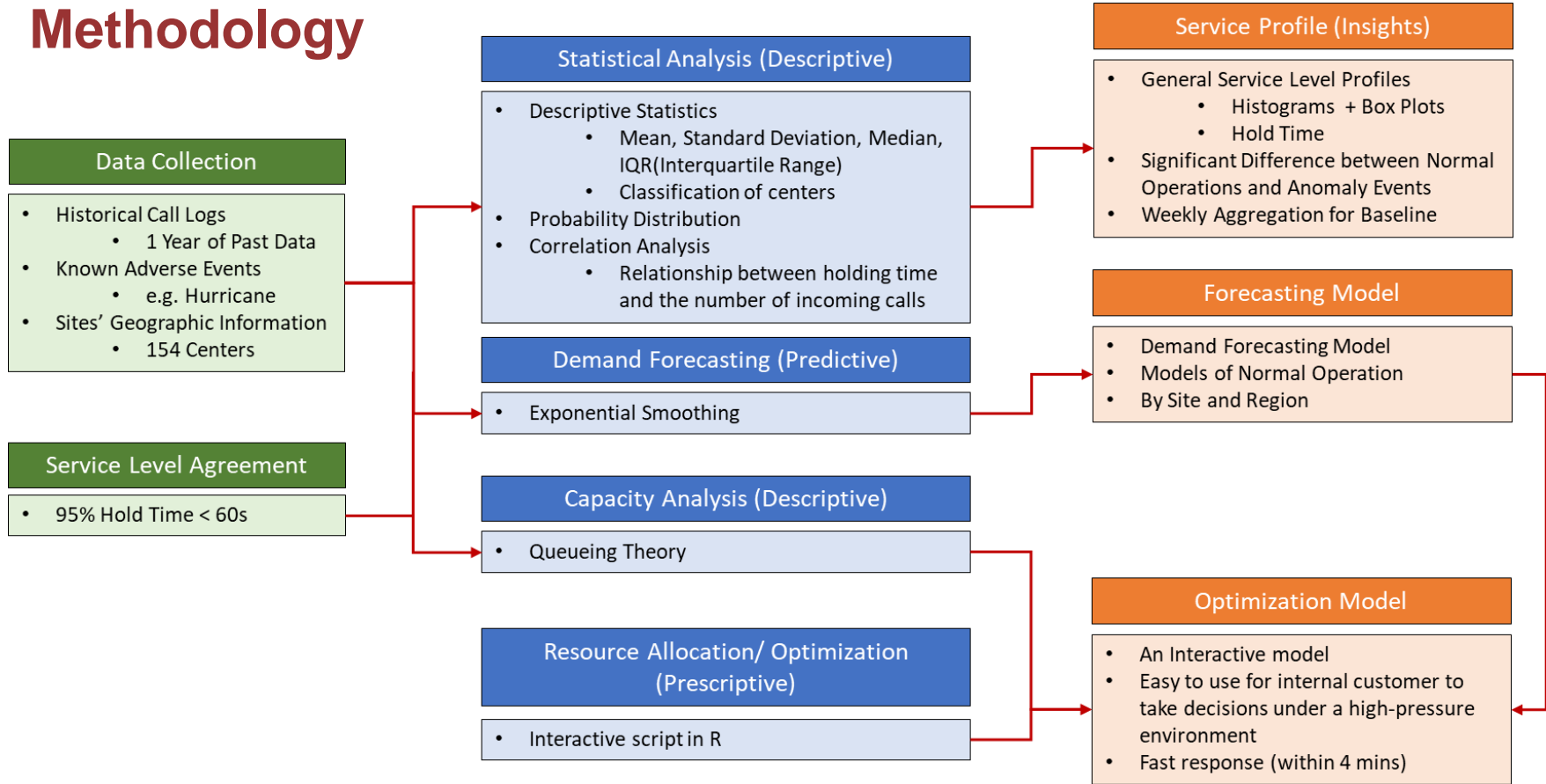
After Catastrophe Event**



*Data Source: Three weeks of data before climate event

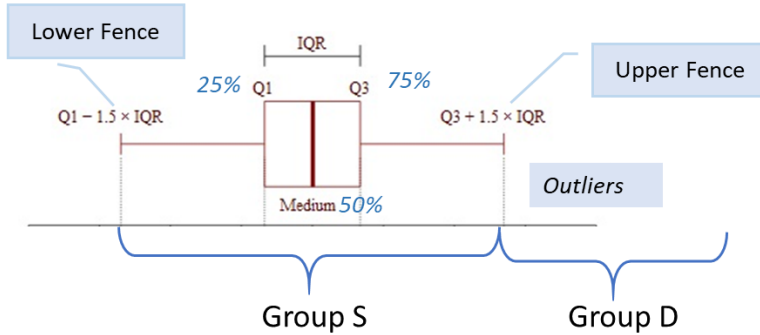
**Data Source: Three weeks of data after climate event

Methodology

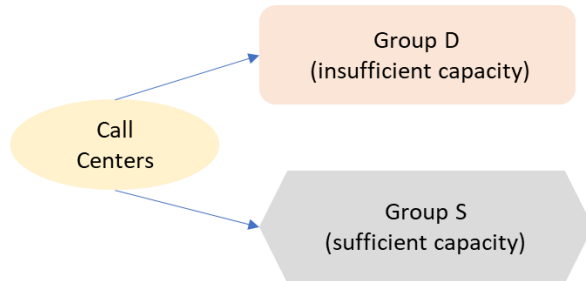


Methodology

Statistical Analysis & Classification



IQR: Interquartile Range



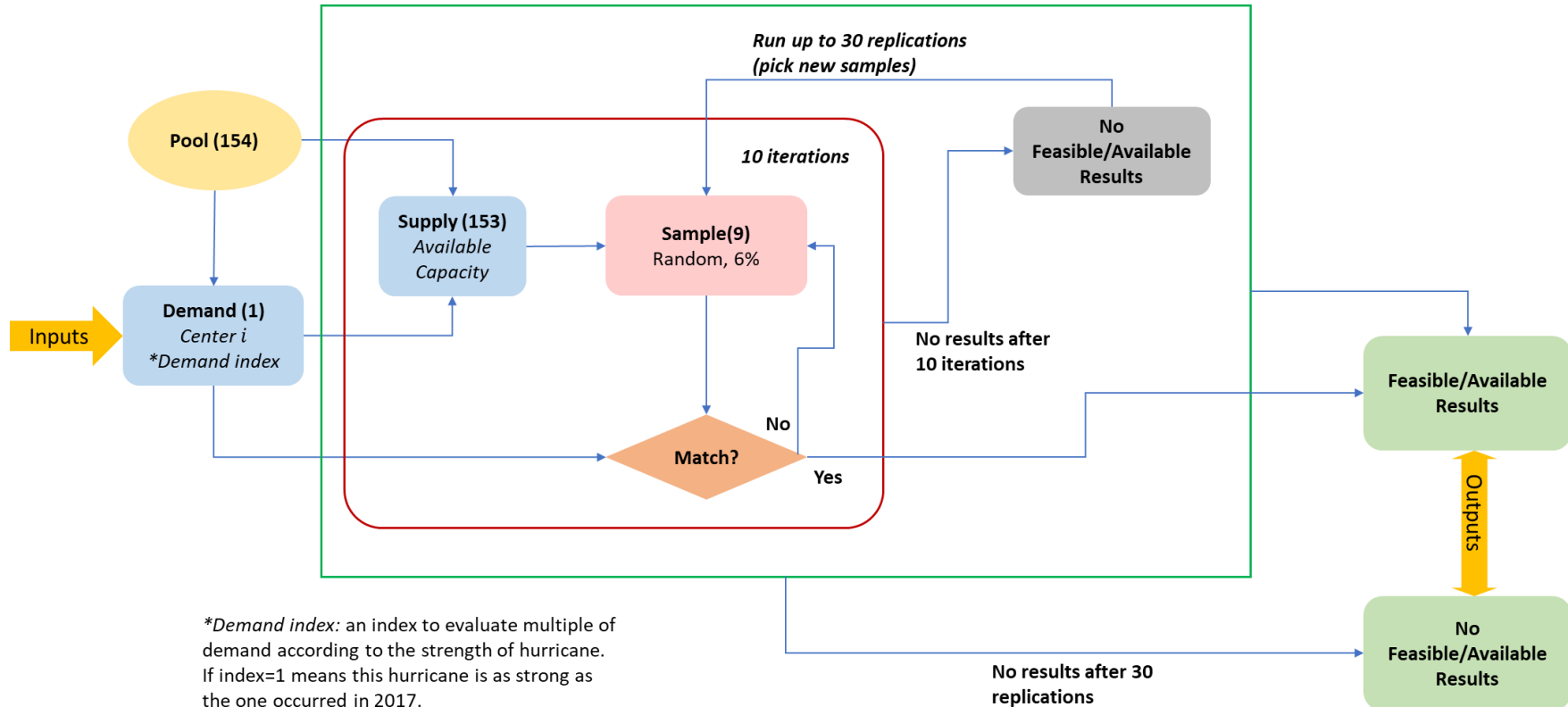
Optimization Model

$$\begin{aligned}
 \text{Min} \quad & \sum_{i=1}^M \sum_{j=1}^N c_{ij} * X_{ij} + \sum_{i=1}^M \sum_{j=1}^N \sum_{k=1}^T s_{ijk} * Y_{ijk} && \leftarrow \text{Objective Function} \\
 \text{s.t.} \quad & Y_{ijk} \leq \frac{C_{ijk}}{D_{jk}} \quad \forall_{ijk} \in I, J, K && \leftarrow \text{Capacity Constraint} \\
 & \sum_{i=1}^M Y_{ijk} \geq 1 \quad \forall_{jk} \in J, K && \leftarrow \text{Demand Constraint} \\
 & \sum_{k=1}^T (-Q * X_{ij} + Y_{ijk}) \leq 0 \quad \forall_{ij} \in I, J && \leftarrow \text{Linking Constraint} \\
 & X_{ij}, Y_{ijk} \in \{0,1\} && \leftarrow \text{Binary Constraint}
 \end{aligned}$$

- i serial number of location $i \in [1,154]$
- j serial number of queue $j \in [1,5]$
- k serial number of time slot $k \in [1,9]$
- X_{ij} be recommended to reroute calls, =1; otherwise, =0 (supply side)
- Y_{ijk} need to be reroute, =1; otherwise, =0 (demand side)
- Q big number

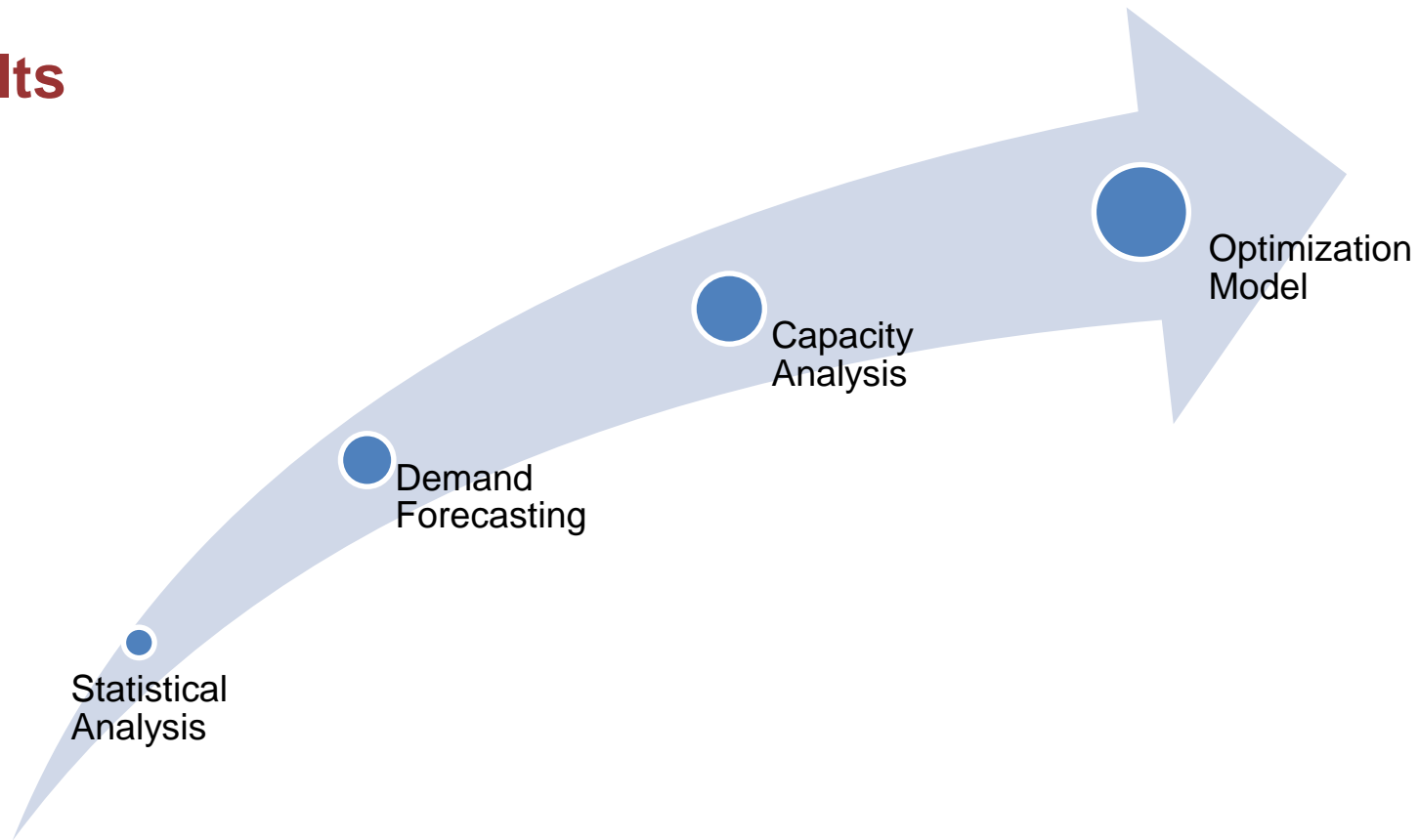
Methodology

Optimization Model

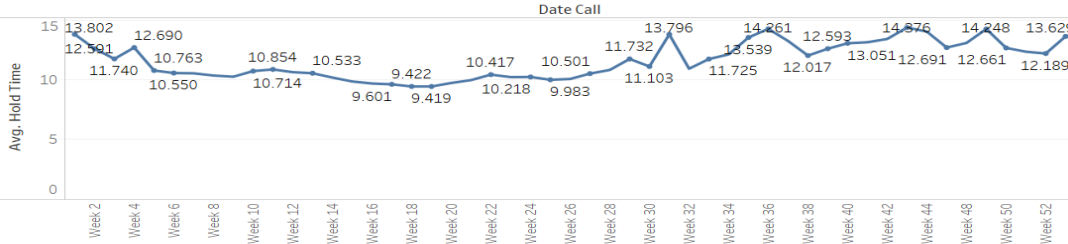


*Demand index: an index to evaluate multiple of demand according to the strength of hurricane. If index=1 means this hurricane is as strong as the one occurred in 2017.

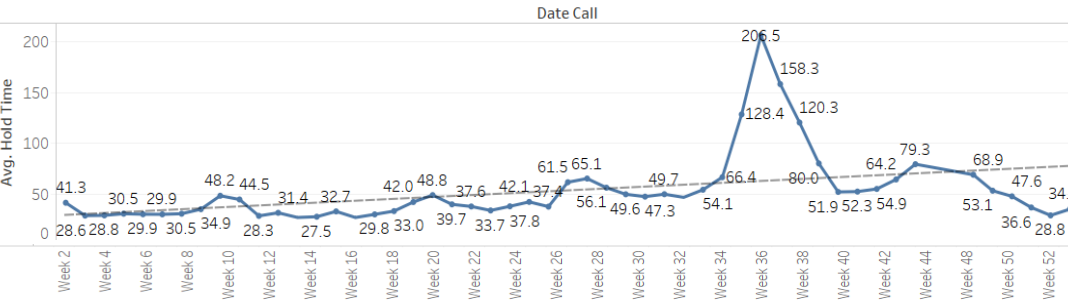
Results



Results – Statistical Analysis



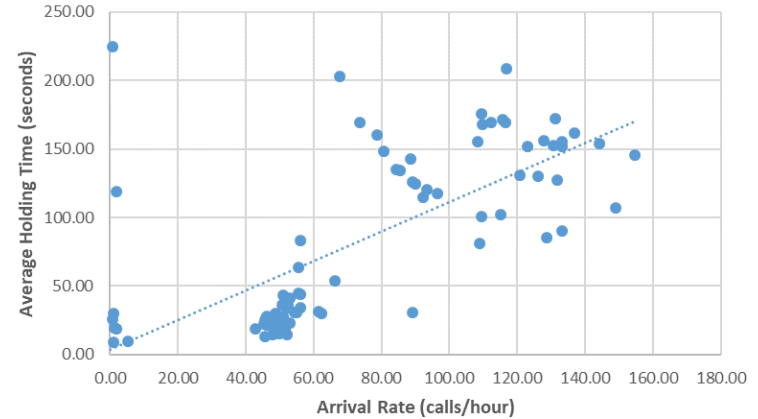
Call centers analysis with holding time lower than 24.5 seconds (Group S)



Call centers analysis with holding time greater than 24.5 seconds (Group D)

	Q1 (25%)	Q2 (50%)	Q3 (75%)	IQR (Q3-Q1)	Lower Fence (Q1-1.5*IQR)	Upper Fence (Q3+1.5*IQR)
Holding time	4	7	14	7	-6.5	24.5
Duration time	60	116	221	105	97.5	378.5
Total time	75	135	247	112	-93	415

Interquartile Range (IQR) analysis for all call centers



Relationship between call arrival rate and holding time (July to October 2017)

Results – Demand Forecasting

- Simple exponential smoothing
- Parameter alpha (α) that minimizes Root Mean Squared Error (RMSE) of call demand.

$$\alpha = 0.21$$

<i>Alpha (α)</i>	<i>RMSE</i>
0.21	4.335632
0.2	4.336268
0.18	4.337819
0.19	4.339523
0.17	4.343181
0.16	4.344477
0.15	4.347776
0.22	4.34808
0.23	4.350566

Results – Capacity Analysis

Inputs: Waiting time (t_q^{ijk}), coefficient of variation for interarrivals (CV_a^{ijk}), coefficient of variation of process time (CV_p^{ijk}), process time (t_p^{ijk}) and, number of parallel agents (m^{ijk})

Output: Maximum capacity for call arrivals (r_a^{ijk})

Location	Queue	Timeslot	Demand	Max. Capacity	Capacity Bandwidth
312	Queue 1	1	6	54	48
312	Queue 1	2	6	58	52
312	Queue 2	3	4	58	54
312	Queue 2	4	3	46	43
312	Queue 3	5	3	30	27
312	Queue 3	6	3	23	20
312	Queue 5	7	3	24	21
312	Queue 5	8	3	4	1
312	Queue 4	9	3	42	39

Results – Optimization Model

Inputs:

- Location to reroute calls for
- Demand Forecast
- Capacity Bandwidth

Output: Call rerouting assignments

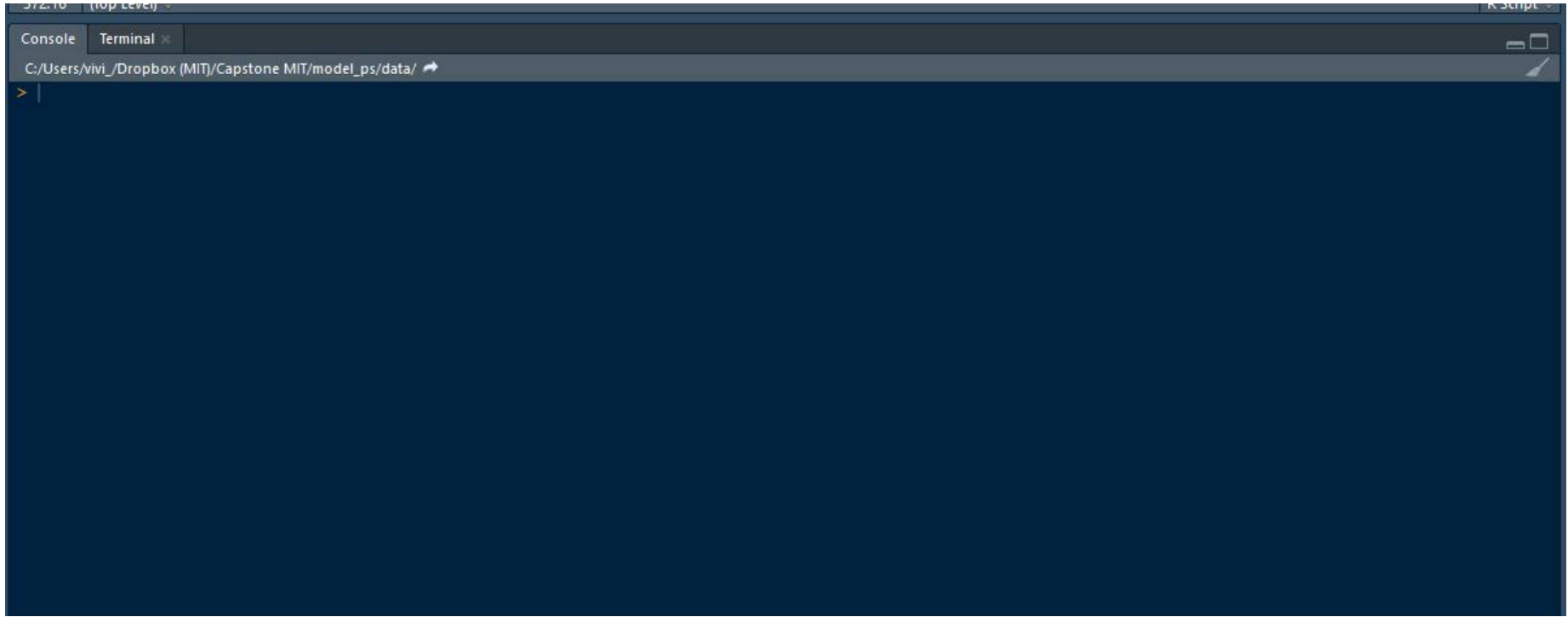
Solver: GLPK

Software: R

Queue Location "A"	Location to	Timeslots "A"
1	349	c("1", "3", "4", "6", "8", "9")
1	437	c("2", "5", "7")
2	365	c("1", "2", "3", "4", "5", "6", "7", "8", "9")
3	429	c("1", "3", "7", "8")
3	391	c("2", "4", "5", "6", "9")
4	447	c("1", "2", "3", "4", "5", "6", "7", "8", "9")
5	349	c("1", "3", "4", "6", "7", "9")
5	447	c("2", "5", "8")

Example of optimization output

Interactive Script in R



Discussion

- As the **number of locations** in the optimization **increases**, the **running time** of the model **increases** exponentially
- Random selection of locations with multiple iterations can help minimize the running time of the Mixed Integer Linear Programming (MILP) model
- Call rerouting framework can be applied in other scenarios such as outages and call center closures

Conclusion

- A sudden increase of demand affects the service level the company has with its customers
- Optimization model helps on **minimizing** the **risk** of losing a customer due to bad service during a catastrophe event
- Implementing the proposed framework will lead to quicker **response times**, better **customer service** and higher **customer satisfaction**

Future Work

- Interactive dashboard using “Shiny Apps” package or Matplotlib
- Utility development
- Integration with ERP system



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Appendix A - Queueing Theory Equations

$$t_q = \left(\frac{CV_a^2 + CV_p^2}{2} \right) \left(\frac{u\sqrt{2(m+1)-1}}{m(1-u)} \right) t_p$$

$$u = \frac{r_a * t_p}{m}$$

Notation	Definition	Unit
r_a^{ijk}	Rate of call arrival at location i in queue j for timeslot k	calls/ time
t_a^{ijk}	Mean time between arrivals at location i in queue j for timeslot k	time/call
CV_a^{ijk}	Coefficient of variation of interarrivals at location i in queue j for timeslot k	
m^{ijk}	Number of parallel agents at location i in queue j for timeslot k	
r_p^{ijk}	Rate or capacity at location i in queue j for timeslot k	calls/time
t_p^{ijk}	Mean effective process time at location i in queue j for timeslot k	time/call
CV_p^{ijk}	Coefficient of variation of process time at location i in queue j for timeslot k	

Notation	Definition	Unit
t_q^{ijk}	Expected waiting time at location i in queue j for timeslot k	time
CT^{ijk}	Expected time in system ($t_q^{ijk} + t_p^{ijk}$) for a call at location i in queue j for timeslot k	time
WIP^{ijk}	Average calls in process at location i in queue j for timeslot k	calls
WID_q^{ijk}	Average work in process in queue at location i in queue j for timeslot k	calls
u^{ijk}	Utilization of the server ($r_a^{ijk} + r_p^{ijk}$) at location i in queue j for timeslot k	calls/time