

# Predictive Demand Model Strategy

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# Problem Statement

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- 1) Building a Predictive Demand Model – Can be trend/time series based or causal
- 2) Basis  $n+2$  month forecasted , planning the optimized supply need for month for the next 12 months
- 3) Based on the Associated costs constrains creating a simulation showing the Net Loss of business and/or net additional cost

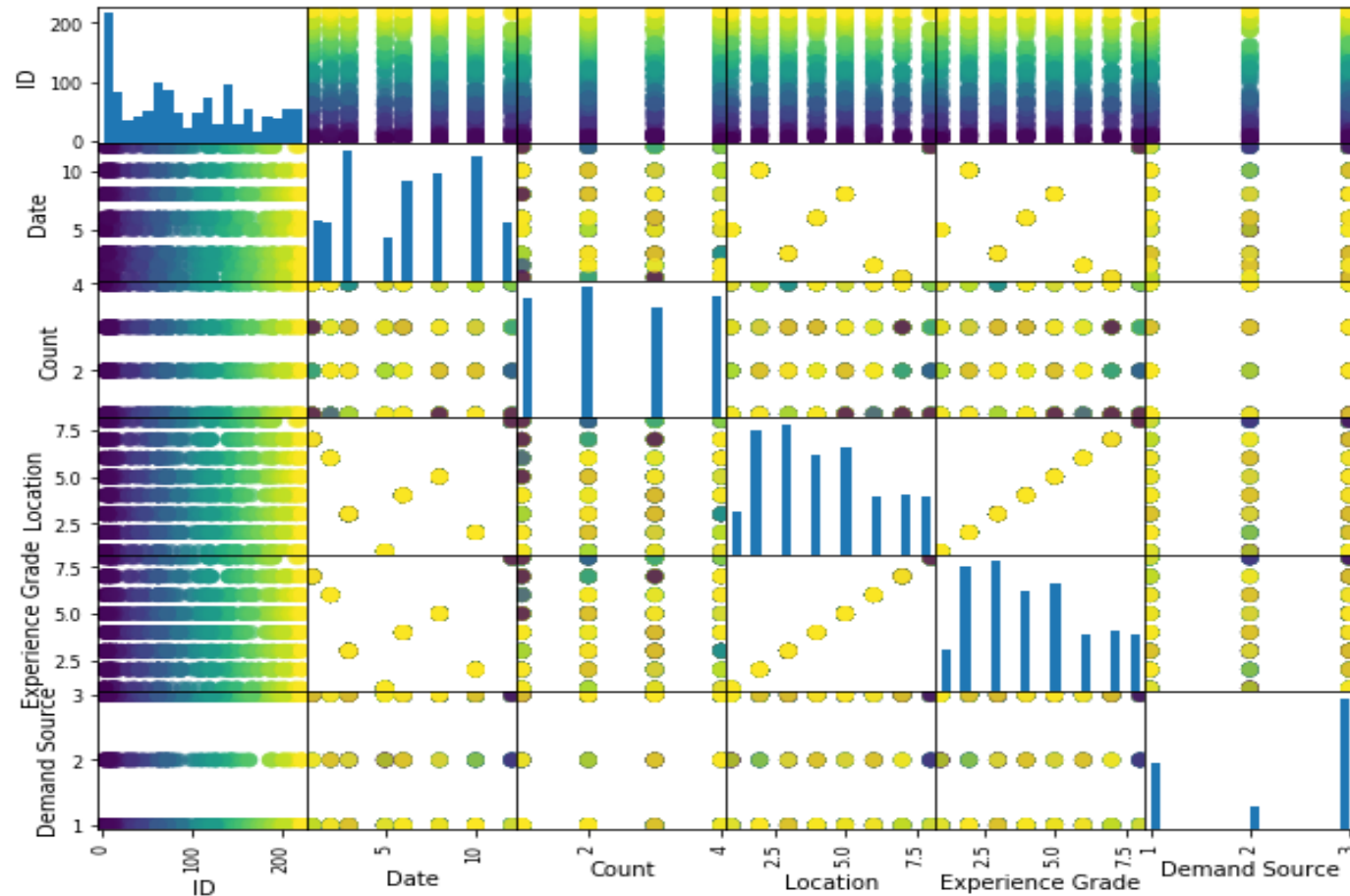
# THE DATA –Demand Trend Last Year

	A	B	C	D	E	F	G	H
1	Month DD	No. of FTE Reque	SkillList	Location	Experience Grade	Practice	Skill Group	Demand Sour
2	May	4	Salesforce (Functional)	Bangalore	A4	BANK	(F) IT Operations	Account
3	October	2	Amazon Web services - Build , Admini	Bhubaneswar	B2	M&FT	(F) IT Operations	Account
4	March	2	Amazon Web services - Build , Admini	Pune	A3	TEST		Support
5	October	1	Test Automation	Bhubaneswar	B2	M&FT	(F) IT Operations	Account
6	June	2	PMO (Account)	Noida	A5	I&D	(D) Banking Sector	Proactive
7	August	4	Salesforce (Functional)	Gurgaon	B1	GP	(T) Data Governance / Master Data Mana	Account
8	June	1	Testing Manual,Banking,Test Automat	Noida	A5	I&D		Support
9	August	4	Amazon Web services - Build , Admini	Gurgaon	B1	INS		Proactive
10	March	3	Angular.js	Pune	A3	TEST	(F) IT Operations	Support
11	March	2	Angular.js	Pune	A3	TEST	(F) IT Operations	Proactive
12	June	2	POSTGRESQL	Noida	A5	ADM		Support
13	February	4	Informatica MDM (formerly Syperian)	Hyderabad	A2	CFS	(T) Card packages	Account
14	October	2	Informatica MDM (formerly Syperian)	Bhubaneswar	B2	M&FT	(T) Office / Desktop Solutions	Proactive
15	June	1	Informatica MDM (formerly Syperian)	Noida	A5	I&D	(F) Business Management / Management	Support
16	January	2	Angular.js	Chennai	A1	DCX	(T) Enterprise Content Management (EC	Proactive
17	August	3	Angular.js	Gurgaon	B1	INS	(D) Banking Sector	Proactive
18	March	3	Java	Pune	A3	TEST	(T) Enterprise Content Management (EC	Support
19	August	1	Java	Gurgaon	B1	GP	(T) Mainframe Development Technologi	Proactive
20	January	1	Java	Chennai	A1	DCX	(T) Card packages	Account
21	March	3	Test Environment Management	Pune	A3	TEST	(T) Operating systems and Virtualization	Account

# THE DATA-Headcount

	A	B	C	D	E	F	G	H	I
1	Region	Employee Code	Last Name	Local Date of Join	Designation	Status	Market Unit	SkillList	Location
2	US	45149	AALLURI	31-10-2016	CONSULTANT	Billable	BANK	Salesforce (Functional)	Bangalore
3	IN	47875	AARATTUKULAM	20-11-2014	SENIOR SOFTWARE	Billable	BANK	Amazon Web services - Build , A	Bhubaneswar
4	IN	90386	AASHIK	15-12-2014	ASSOCIATE CONSUL	Billable	BANK	Amazon Web services - Build , A	Pune
5	IN	41161	Aastha	26-02-2015	ASSOCIATE CONSUL	Billable	BANK	Test Automation	Bhubaneswar
6	IN	92293	ABBAN	02-12-2013	CONSULTANT	Billable	BANK	PMO (Account)	Noida
7	IN	79912	ABBANOLLA	19-01-2012	ASSOCIATE CONSUL	Billable	BANK	Salesforce (Functional)	Gurgaon
8	IN	60970	ABBARAJU	01-10-2009	MANAGER	Billable	BANK	Testing Manual,Banking,Test Au	Noida
9	IN	92528	ABBARAJU	08-01-2015	SENIOR MANAGER	Billable	BANK	Amazon Web services - Build , A	Gurgaon
10	IN	77180	ABBARAJUL	12-08-2013	SENIOR MANAGER	Billable	INS	Angular.js	Pune
11	IN	21384	ABBAS	24-01-2013	SENIOR CONSULTAI	Billable	INS	Angular.js	Pune
12	IN	76211	ABBI	20-12-2012	SENIOR CONSULTAI	Billable	INS	POSTGRESQL	Noida
13	IN	47520	ABBIREDDY	22-09-2011	SENIOR CONSULTAI	Billable	INS	Informatica MDM (formerly Sype	Hyderabad
14	IN	58905	ABBURI	25-02-2013	CONSULTANT	Billable	INS	Informatica MDM (formerly Sype	Bhubaneswar
15	IN	28524	ABBURI	31-10-2013	CONSULTANT	Billable	INS	Informatica MDM (formerly Sype	Noida
16	US	76959	ABDUL HAMEED	24-11-2014	SENIOR CONSULTAI	Billable	INS	Angular.js	Chennai
17	IN	29787	ABDUL NATHAR USS	22-07-2013	SENIOR CONSULTAI	Billable	INS	Angular.js	Gurgaon
18	US	35192	ABDULRASAD	06-10-2014	SENIOR CONSULTAI	Billable	INS	Java	Pune
19	IN	49925	ABHISHEK	09-03-2015	ASSOCIATE CONSUL	Billable	INS	Java	Gurgaon
20	IN	67919	ABHYANKAR	30-12-2005	MANAGER	Billable	INS	Java	Chennai
21	US	66815	ABHYANKAR	16-09-2013	MANAGER	Billable	INS	Test Environment Management	Pune

# Analysing the data




From the plots, we can see that the classes are not well separated , Therefore it's difficult for a machine learning model to separate them using a classifier


# The Top Skills Required.....

Predicting becomes easy when we have the idea of the in demand skill for the resource.....

	SkillList	No. of FTE Request Raised	Counts
0	Java	179	409
1	Selenium	120	270
2	Ab Initio	72	208
3	Java (Multi Skilled),Java	66	204
4	Guidewire	64	197
5	Angular.js	57	194
6	Testing Manual	55	129
7	Mulesoft ESB	53	129
8	Pega	47	118
9	Functional Testing	47	114
10	Java (Multi Skilled)	46	114



Skills possessed by the most Billed Person  
of the organisation



Count as per Demands

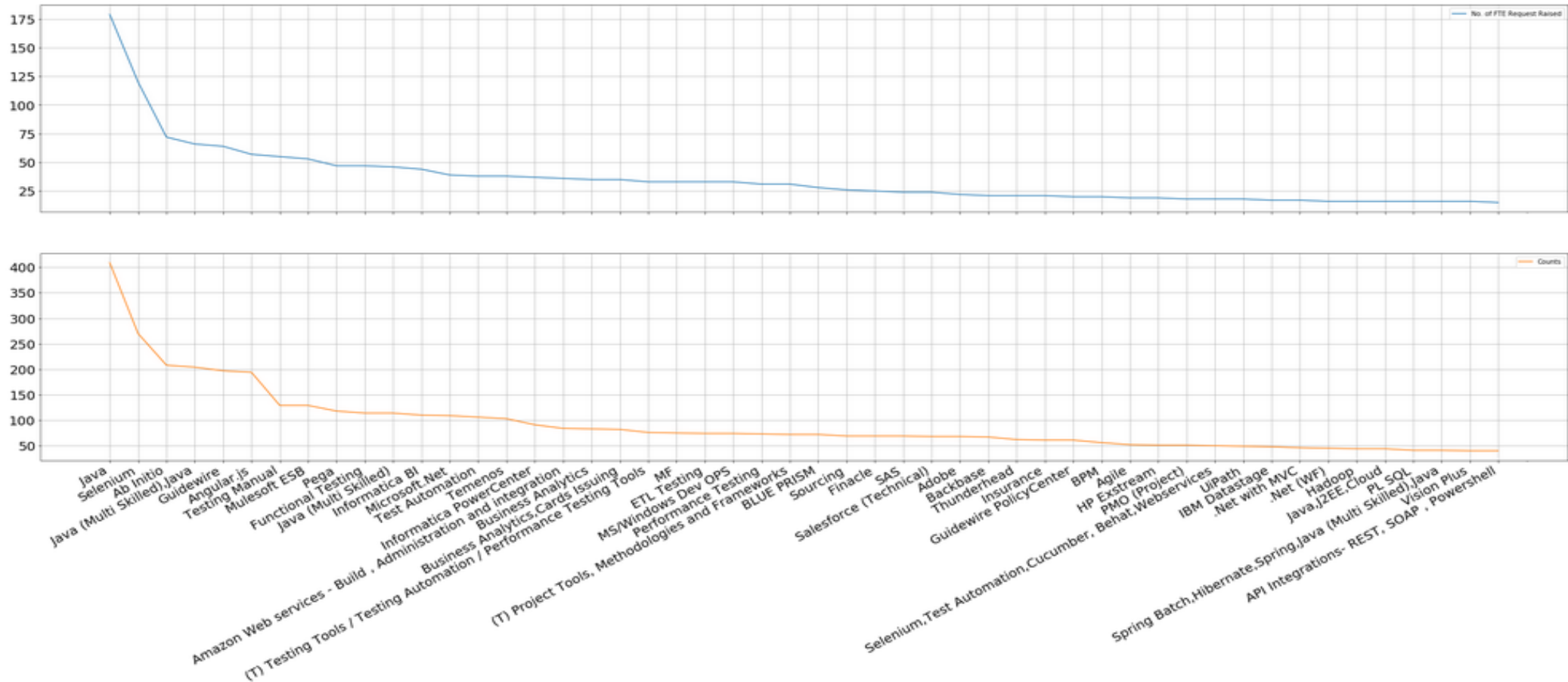
# Noticeable Facts

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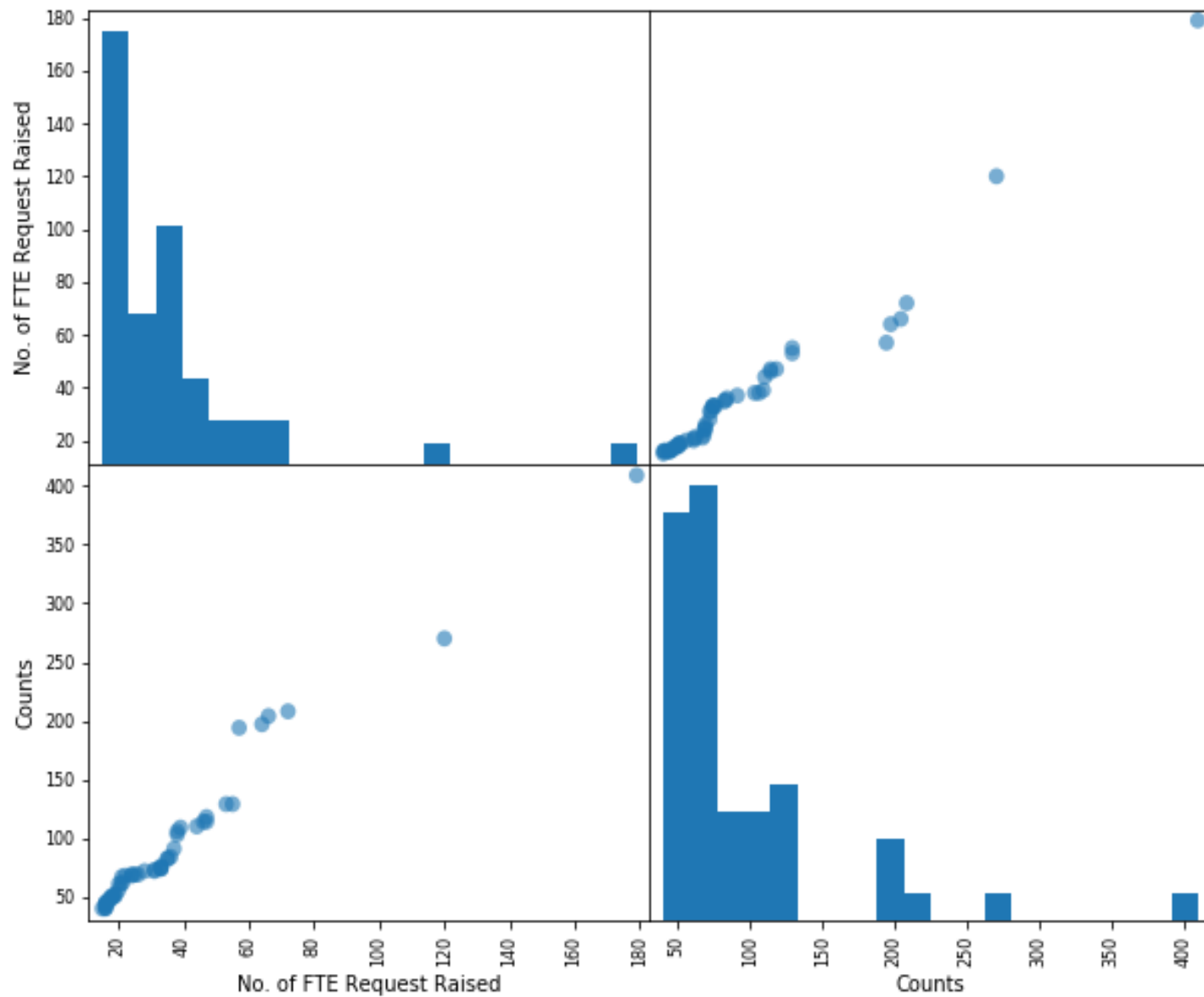
Month DD Raised	Location	Experience Grade	Practice
August	Gurgaon	B1	GP
			INS
December	Hartford	C1	ERP
February	Hyderabad	A2	CFS
January	Chennai	A1	DCX
June	Noida	A5	ADM
			I&D
March	Pune	A3	TEST
May	Bangalore	A4	BANK
October	Bhubaneswar	B2	Cards
			M&FT

➤ According to the data the demand of each month is directly mapped to a specific location & Experience Grade. For Instance : In August the demand location is Gurgaon Experience Grade is B1 & Practice can be either GP or INS. Similarly in December the location is Hartford experience Grade is C1 & practice required is ERP etc.

# Skills Demanded vs Skilled Possessed by Most Billed People



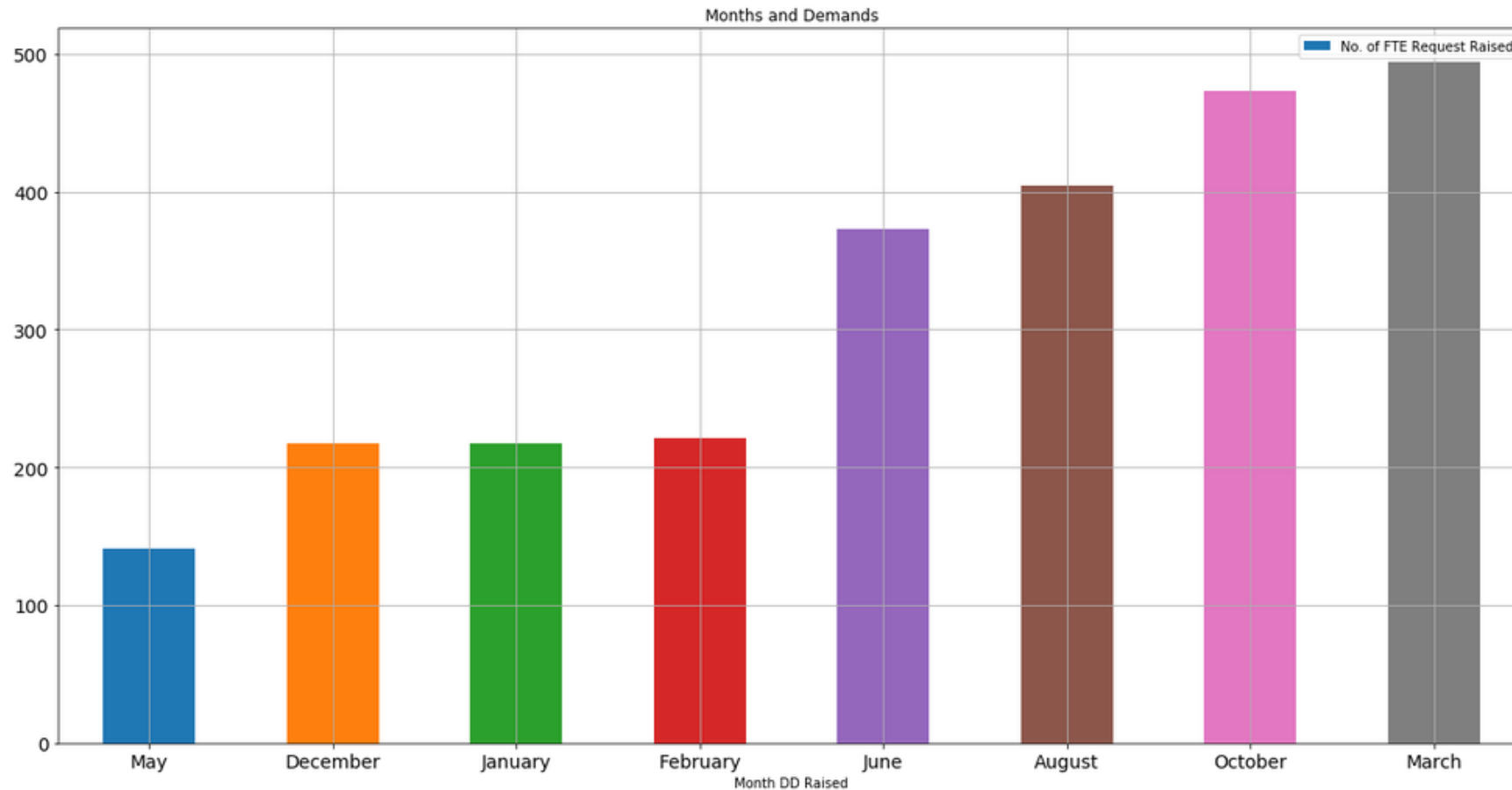




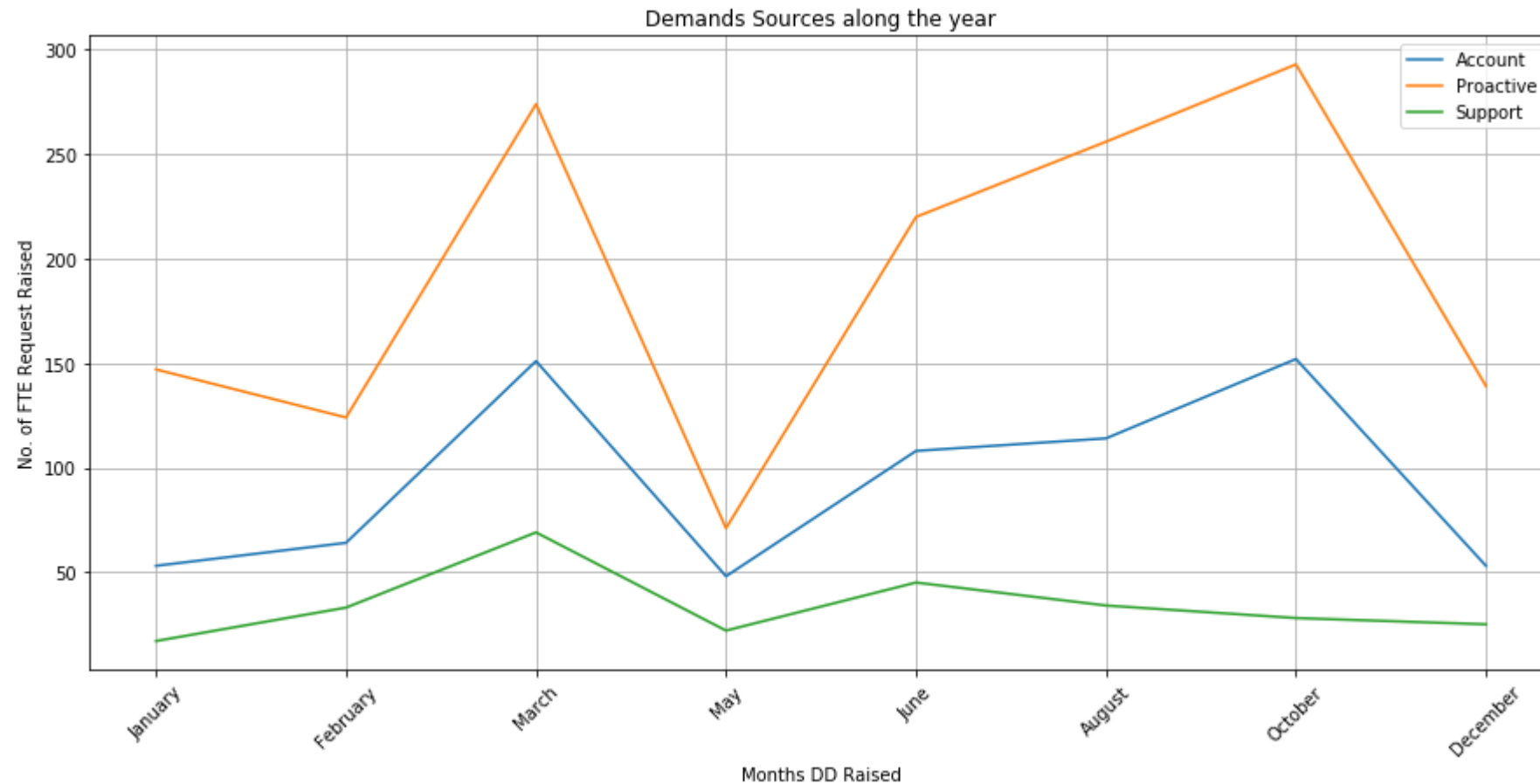
It is difficult to separate even the top skills...

# Demands during Months

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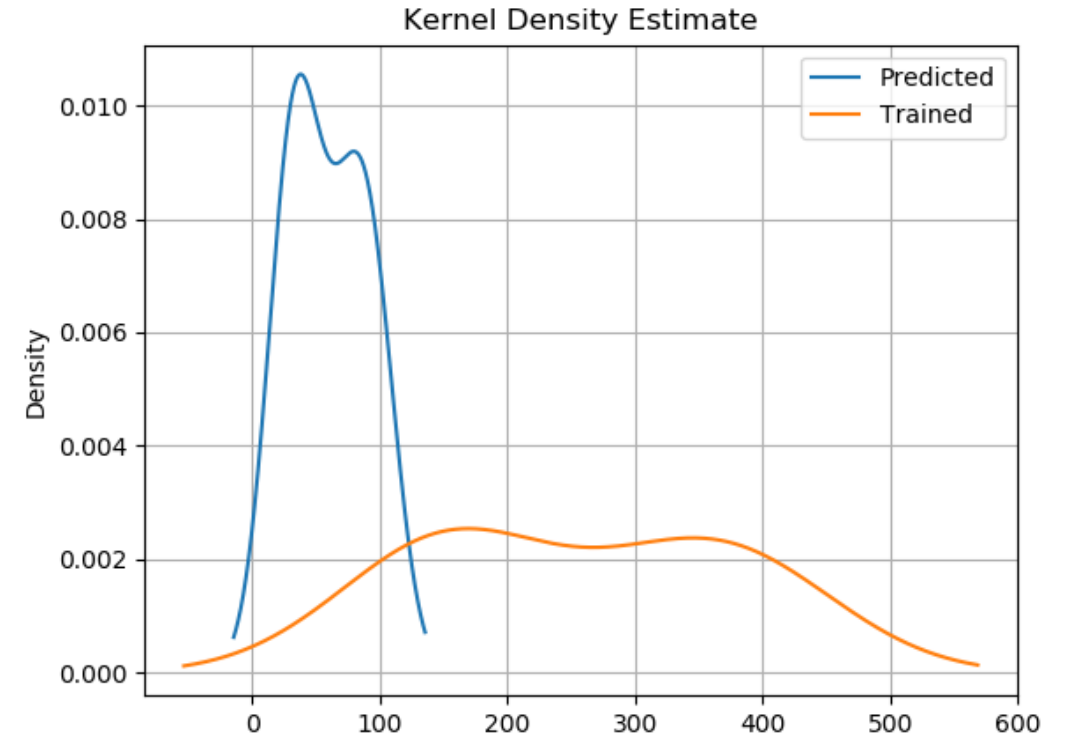
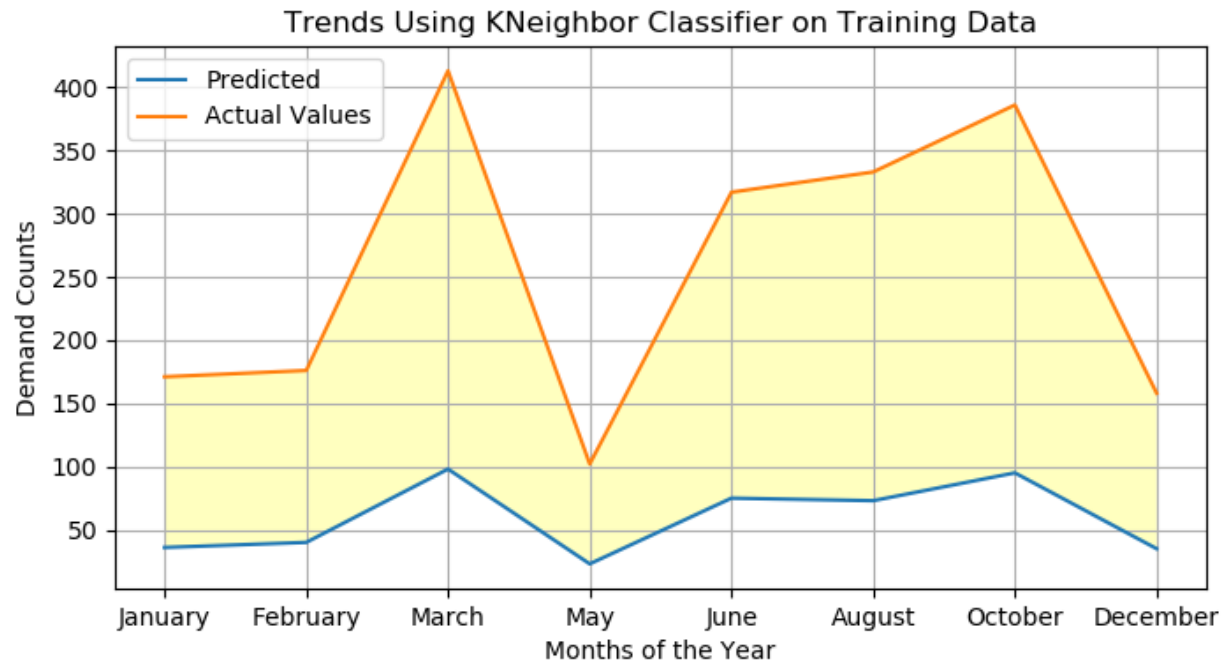


# Demand & Demand Source Along the Year

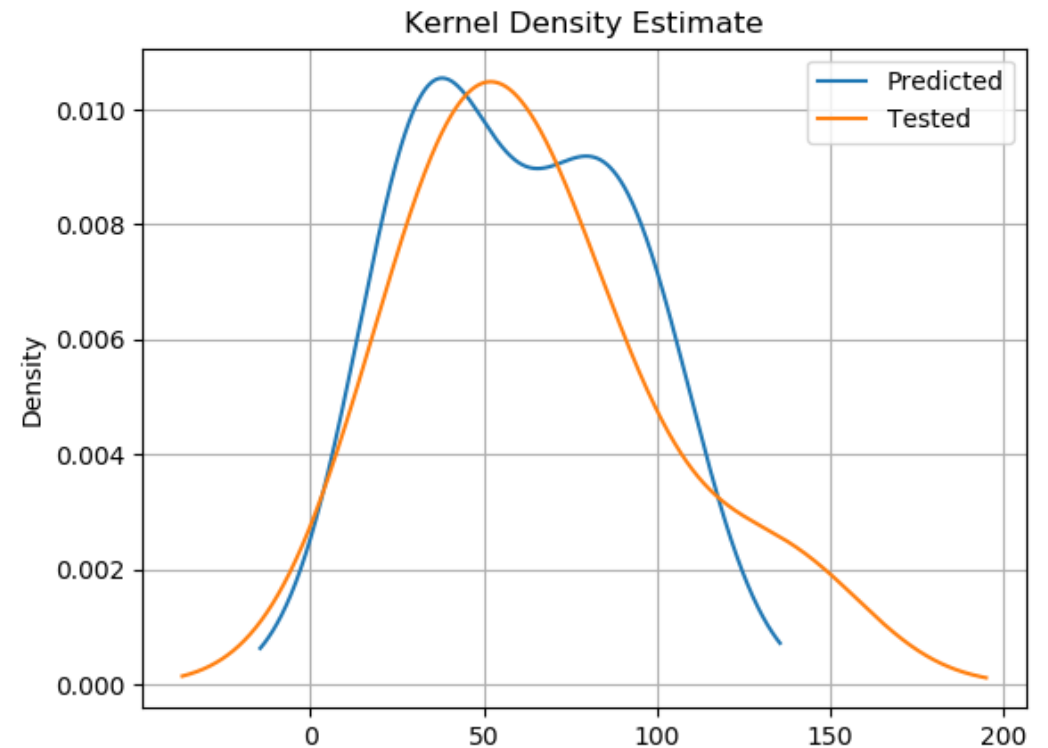
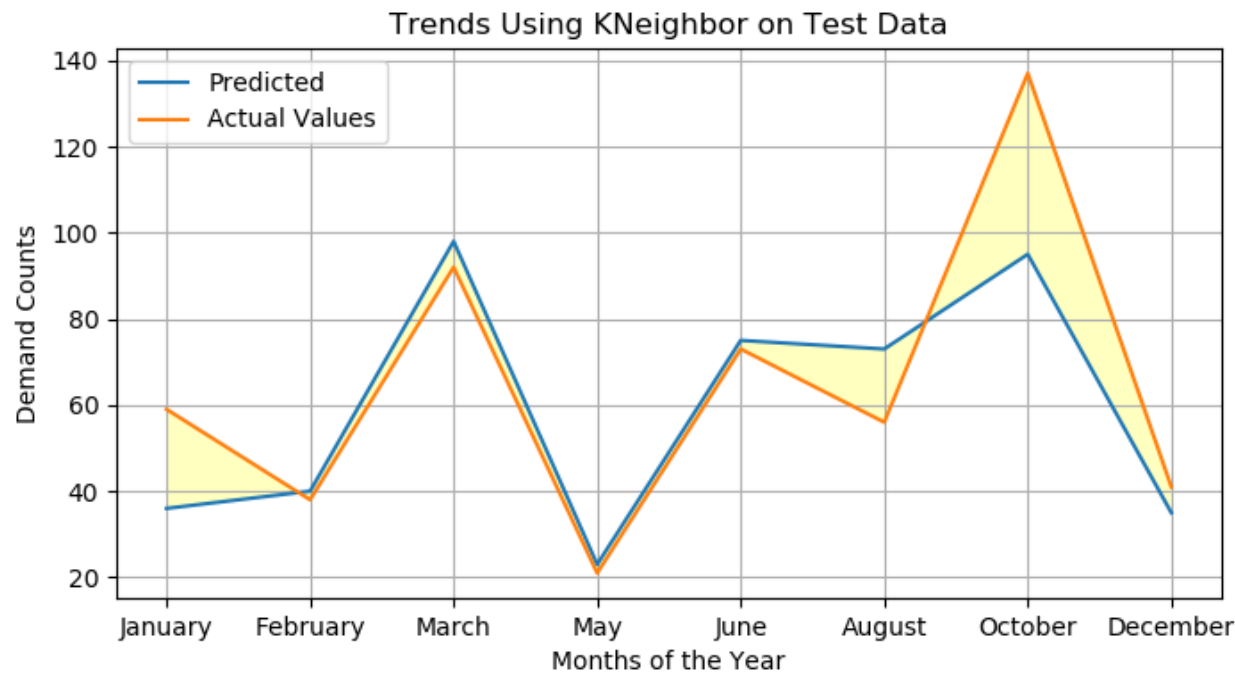


# MODELS AND PREDICTIONS

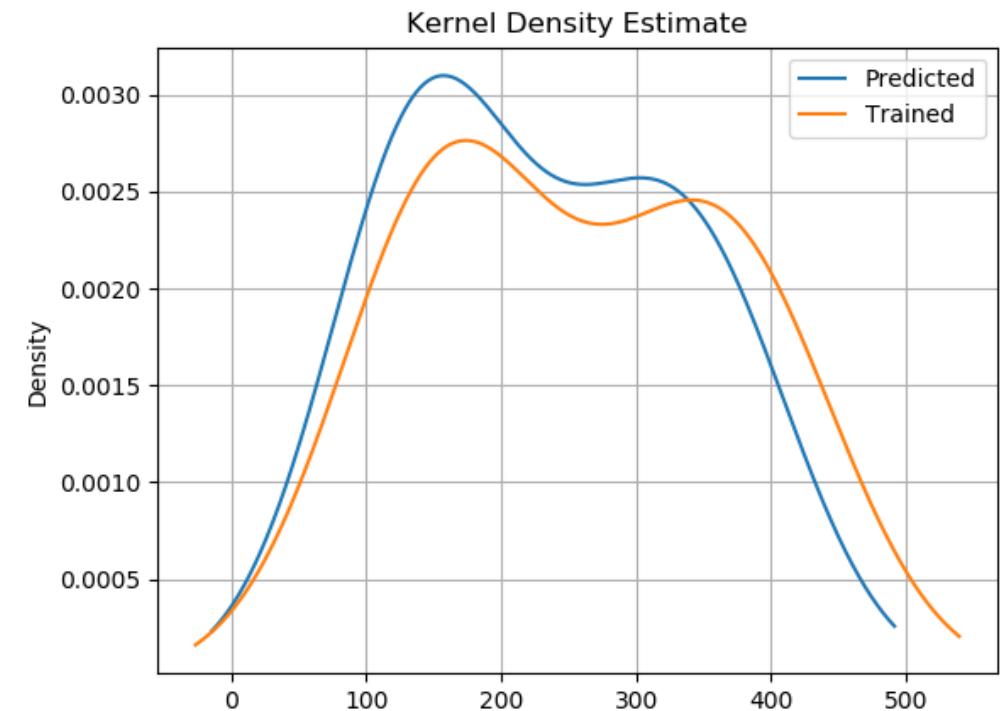
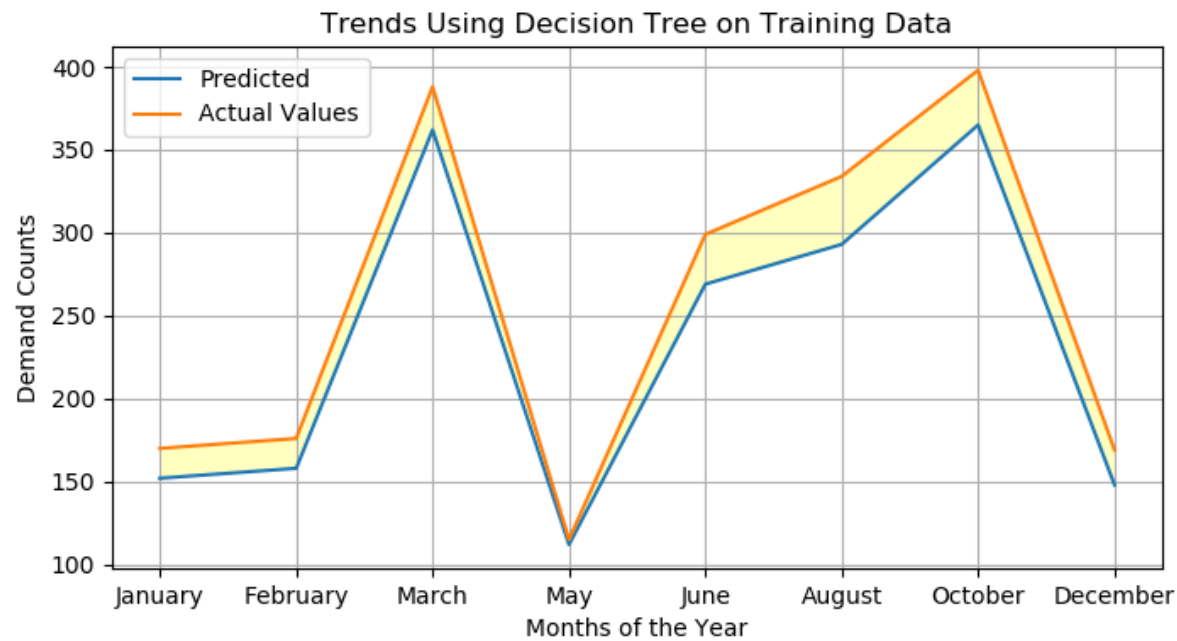
# Prediction Using KNeighbor Classifier on Training Data



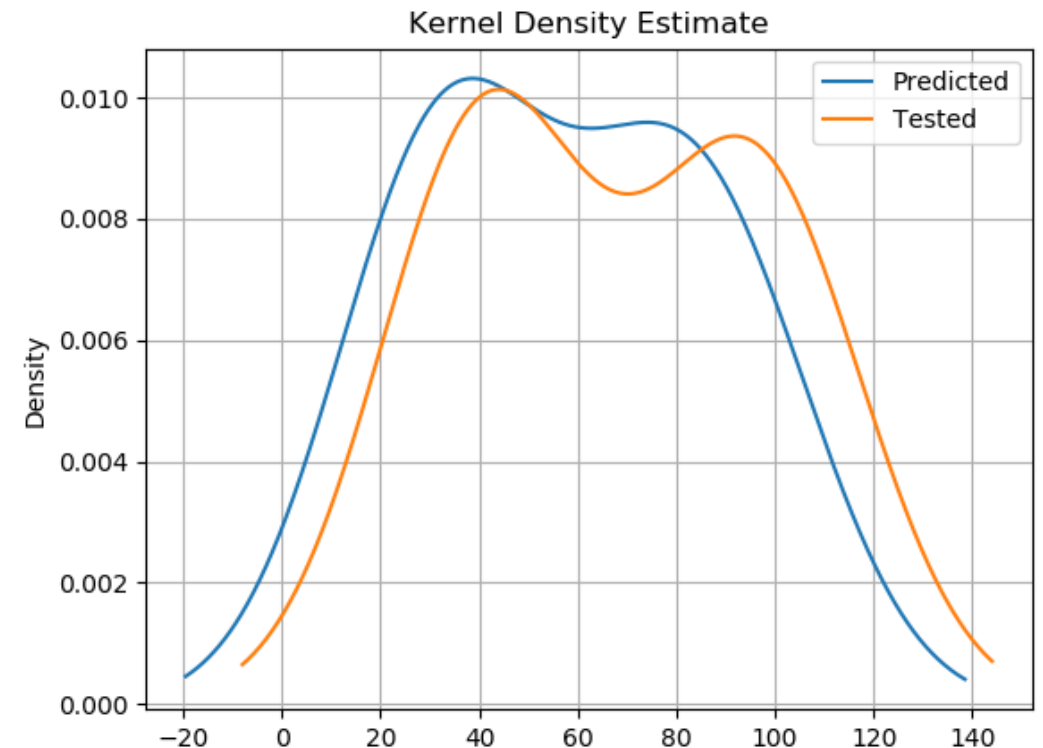
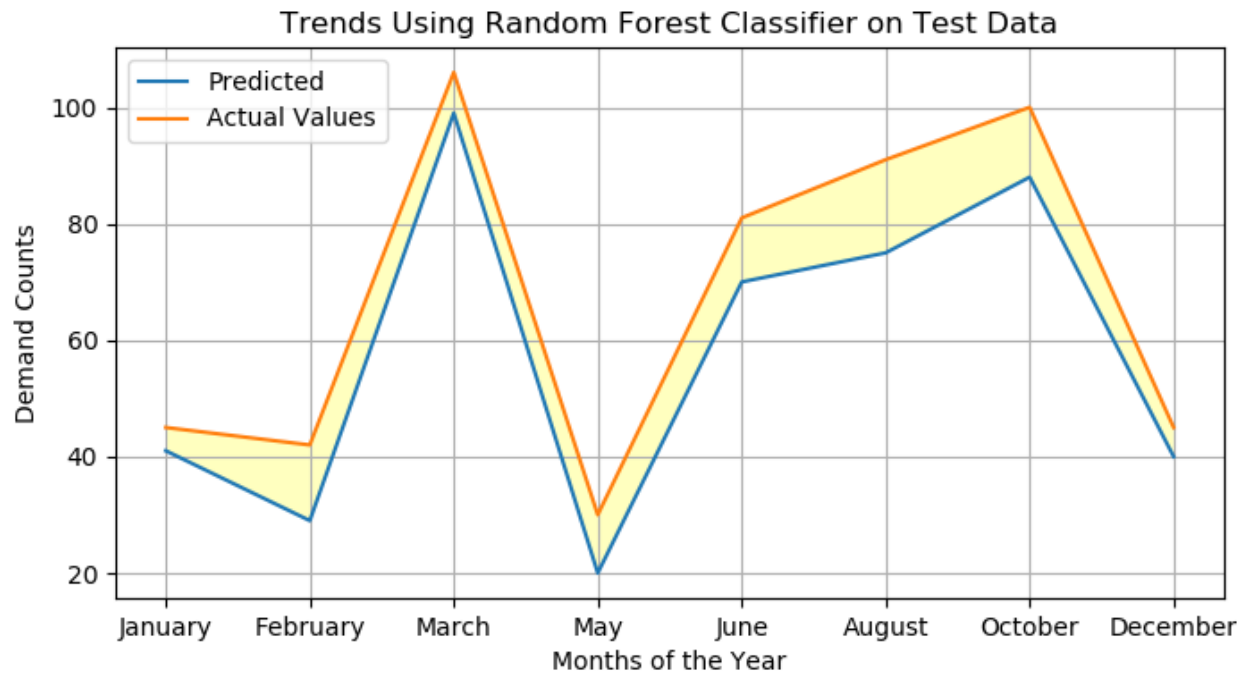
# Prediction Using KNeighbor Classifier on Test Data



# Prediction Using Decision Tree Classifier on Training Data

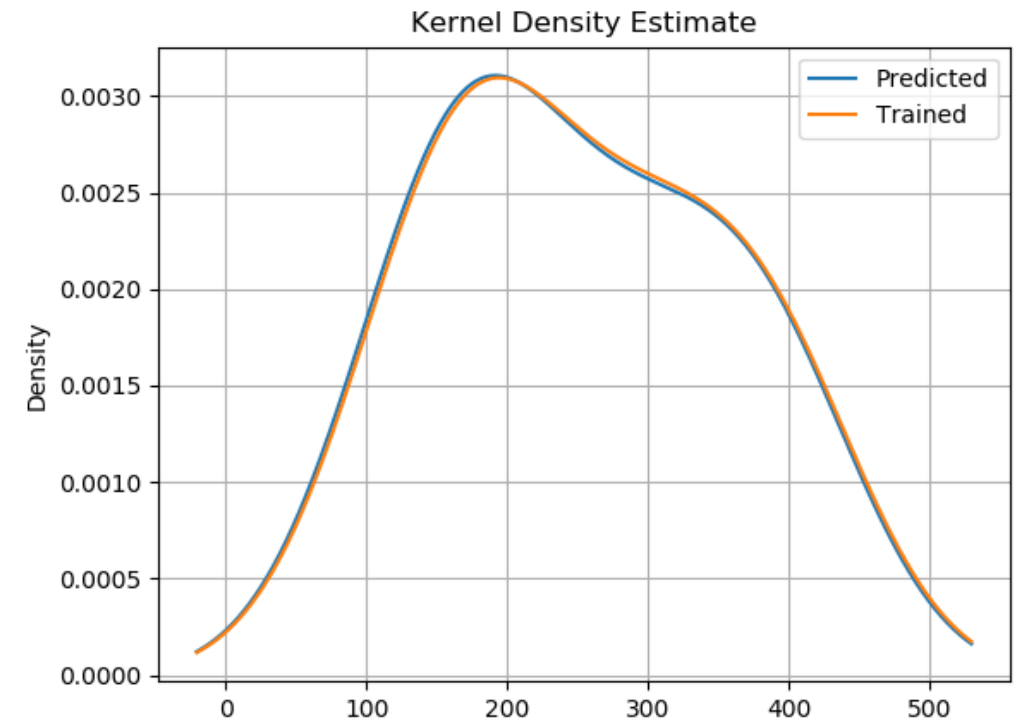
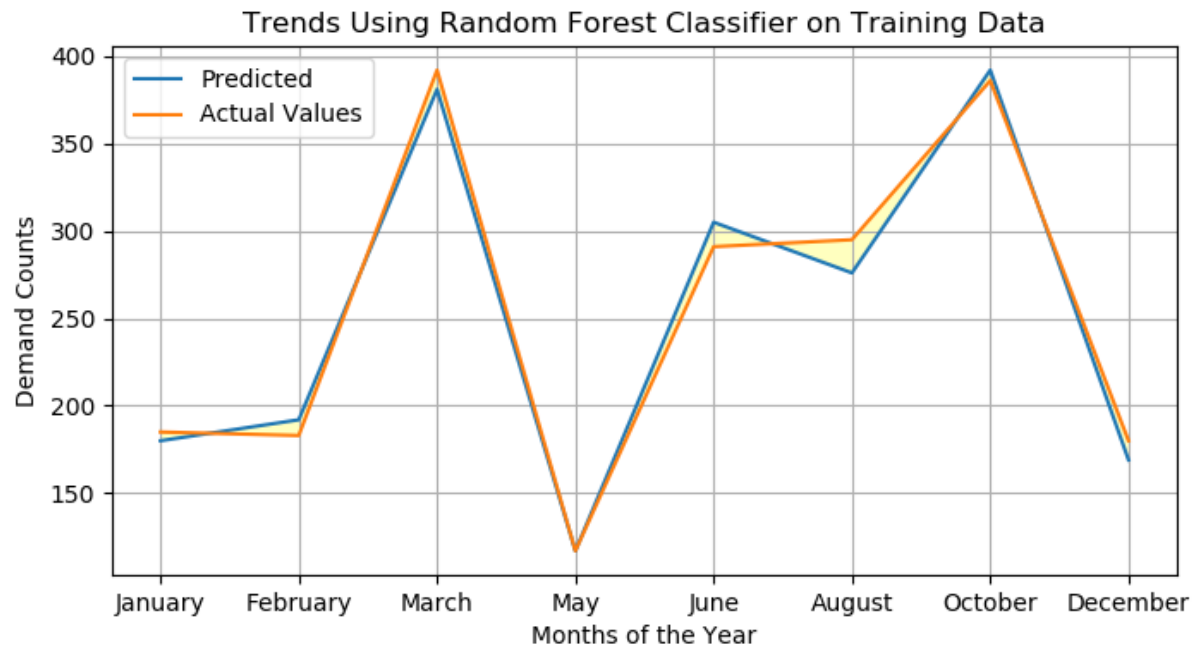


# Prediction Using Decision Tree Classifier on Test Data

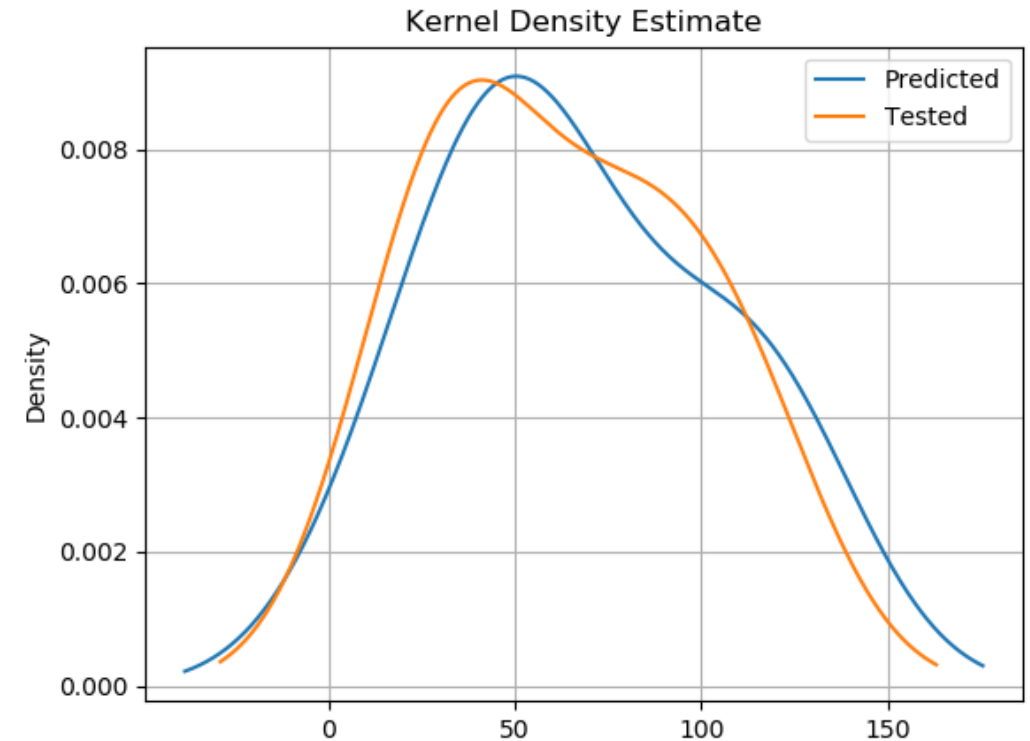
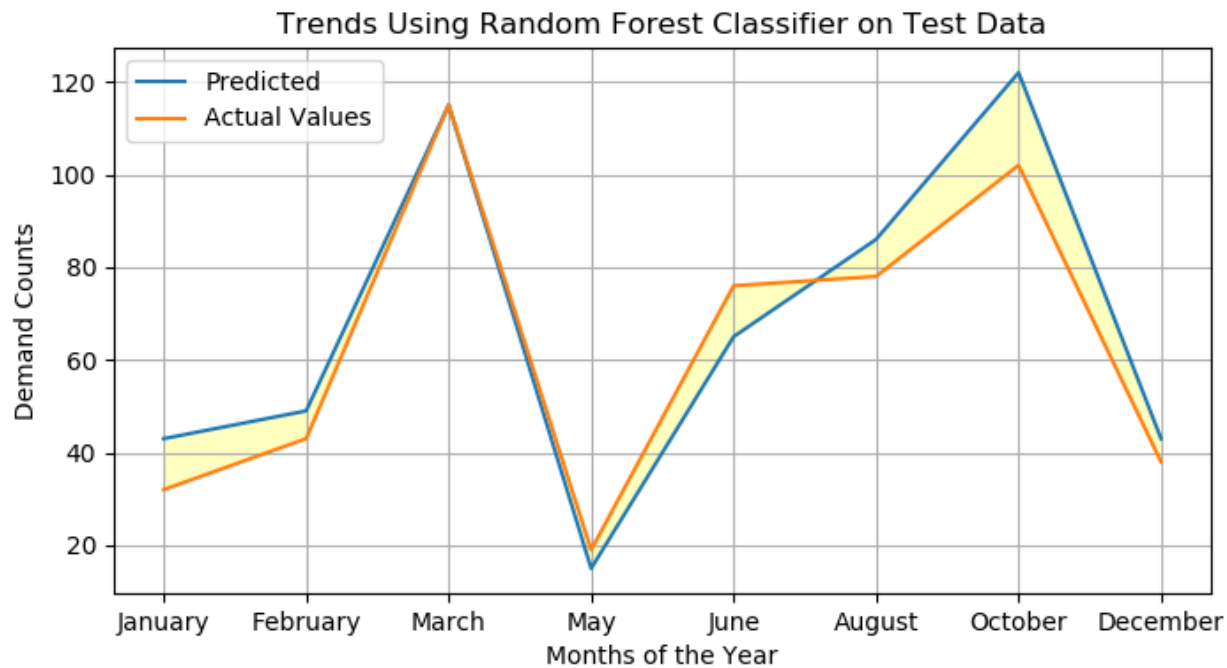




# Prediction Using Random Forest Classifier on Training Data



# Prediction Using Random Forest Classifier on Test Data



# Notes

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- I have tried to use multi-output multi-label classifier for this problem .
- The maximum score attained for all the classifiers are 88% on the Training Set and 66% on the Test Set.
- I have tried classifications methods to discover the possibilities of these classifiers in a problem where it is difficult to classify the labels.
- The output of these classifiers are in the form of months and the count of each unique skill required, each skill has its individual id.
- The intention was to then map the required skills with the people in the bench in the Headcount file and then calculating the Profit/Loss.

Thank You