



Prediction Challenge 1

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Overall Summary/Breakdown of DataSet

```
> table(HireTrainApr10[HireTrainApr10$Coding == "Excellent" & HireTrainApr10$Hired=="Yes",])  
, , Major = CS, College = BestCollege, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = DataScience, College = BestCollege, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = IT, College = BestCollege, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = Stats, College = BestCollege, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = CS, College = BYU, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = DataScience, College = BYU, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = IT, College = BYU, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = Stats, College = BYU, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

```
, , Major = CS, College = Peters, Hired = No
```

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

Overall Summary/Breakdown of DataSet Conti.

, , Major = DataScience, College = Peters, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = IT, College = Peters, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = Stats, College = Peters, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = CS, College = PJIT, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = DataScience, College = PJIT, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = IT, College = PJIT, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = Stats, College = PJIT, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = CS, College = Redbrick, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = DataScience, College = Redbrick, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = IT, College = Redbrick, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

Overall Summary/Breakdown of DataSet Conti.

, , Major = Stats, College = Redbrick, Hired = No

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	0	0	0	0
OK	0	0	0	0
Weak	0	0	0	0

, , Major = CS, College = BestCollege, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	10	5	8	3
OK	0	0	0	0
Weak	0	0	0	0

, , Major = DataScience, College = BestCollege, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	11	8	12	9
OK	0	0	0	0
Weak	0	0	0	0

, , Major = IT, College = BestCollege, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	8	0	11	1
OK	0	0	0	0
Weak	0	0	0	0

, , Major = Stats, College = BestCollege, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	8	10	7	5
OK	0	0	0	0
Weak	0	0	0	0

, , Major = CS, College = BYU, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	12	13	9	10
OK	0	0	0	0
Weak	0	0	0	0

, , Major = DataScience, College = BYU, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	5	8	7	4
OK	0	0	0	0
Weak	0	0	0	0

, , Major = IT, College = BYU, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	7	9	7	7
OK	0	0	0	0
Weak	0	0	0	0

, , Major = Stats, College = BYU, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	11	5	7	5
OK	0	0	0	0
Weak	0	0	0	0

, , Major = CS, College = Peters, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	9	5	11	8
OK	0	0	0	0
Weak	0	0	0	0

, , Major = DataScience, College = Peters, Hired = Yes

Coding	Impression			
	Confident	Nerdy	Outgoing	Shy
Excellent	8	12	14	5
OK	0	0	0	0
Weak	0	0	0	0

Overall Summary/Breakdown of DataSet Conti.

, , Major = IT, College = Peters, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	5	12	9	11
OK	0	0	0	0
weak	0	0	0	0

, , Major = Stats, College = Peters, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	13	9	6	7
OK	0	0	0	0
weak	0	0	0	0

, , Major = CS, College = PJIT, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	5	9	6	10
OK	0	0	0	0
weak	0	0	0	0

, , Major = DataScience, College = PJIT, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	11	1	7	8
OK	0	0	0	0
weak	0	0	0	0

, , Major = IT, College = PJIT, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	8	9	4	6
OK	0	0	0	0
weak	0	0	0	0

, , Major = Stats, College = PJIT, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	10	8	9	7
OK	0	0	0	0
weak	0	0	0	0

, , Major = CS, College = Redbrick, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	7	6	7	8
OK	0	0	0	0
weak	0	0	0	0

, , Major = DataScience, College = Redbrick, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	4	9	11	13
OK	0	0	0	0
weak	0	0	0	0

, , Major = IT, College = Redbrick, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	6	3	9	7
OK	0	0	0	0
weak	0	0	0	0

, , Major = Stats, College = Redbrick, Hired = Yes

Coding	Impression			
	Confident	Nerdy	outgoing	shy
Excellent	12	6	13	6
OK	0	0	0	0
weak	0	0	0	0



Analysis

- The previous four slides contain 40 tables that give us extremely detailed breakdown of the dataset
- Broken down by majors and universities as well as impression and coding, the tables are very helpful in understanding what factors play an important role in getting a person hired.
- They are very precise in giving us exact numbers that help us understand what factors play an important role in getting a person hired.
- We only need the information on what gets a person hired, hence the code I wrote was so that we only get the tables that give us the information on what factors play a role in getting people hired. The tables do not give us any statistical information about the people who were not hired. We do not know the coding skills, impression, major or university for the people that were not hired as the tables for those who are not hired is just all zeros. However, not having this information doesn't impact us as much because we want to figure out what GETS people HIRED.

Coding Skills vs Hired



```
> table(HireTrainApr10$Coding == "Excellent" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1369	631

```
> table(HireTrainApr10$Coding == "weak" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1931	69

```
> table(HireTrainApr10$Coding == "OK" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1445	555

```
> |
```

- The tables to the left show us that although we may have thought that coding skills play a huge role in deciding who gets hired, coding skills may not be the **ONLY** huge factor in decision making
- The tables do show, however, that not a lot of people who have “weak” coding skills get hired. Candidates with “excellent” or “OK” coding skills are given preference to be hired

Impression vs Hired



```
> table(HireTrainApr10$Impression == "Confident" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1679	321

```
> table(HireTrainApr10$Impression == "Nerdy" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1677	323

```
> table(HireTrainApr10$Impression == "Outgoing" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1675	325

```
> table(HireTrainApr10$Impression == "shy" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1714	286

```
> |
```

- The tables to the left show us that although we may have thought that a first impression matters to the recruiting/hiring manager, it doesn't necessarily mean that you will or will not be hired.
- The tables show that regardless of the initial impression the candidate made, the hiring decision was not based solely on whether the applicant was confident, nerdy, outgoing or shy
- However, we can notice that candidates who were "Confident", "Nerdy" and "Outgoing" were more likely to be hired than candidates who were "shy"

College vs Hired

```
2000
> table(HireTrainApr10$College == "BestCollege" & HireTrainApr10$Hired == "Yes")

FALSE  TRUE
 1816   184

> table(HireTrainApr10$College == "BYU" & HireTrainApr10$Hired == "Yes")

FALSE  TRUE
 1737   263

> table(HireTrainApr10$College == "Peters" & HireTrainApr10$Hired == "Yes")

FALSE  TRUE
 1732   268

> table(HireTrainApr10$College == "PJIT" & HireTrainApr10$Hired == "Yes")

FALSE  TRUE
 1767   233

> table(HireTrainApr10$College == "Redbrick" & HireTrainApr10$Hired == "Yes")

FALSE  TRUE
 1693   307

> |
```

- We can see, similarly to the other attributes, that college is not the sole thing that gets a candidate hired.
- However, it can be seen that there is a lower number of candidates that are hired from BestCollege versus having a slightly higher number of candidates that get hired from the remaining colleges (BYU, Peters, PJIT, and Redbrick)
- However, the numbers are not significantly too apart hence there is a chance that college does not impact the hiring decision

Major vs Hired



```
> table(HireTrainApr10$Major == "CS" & HireTrainApr10$Hired == "Yes")  
  
FALSE  TRUE  
 1668   332  
> table(HireTrainApr10$Major == "DataScience" & HireTrainApr10$Hired == "Yes")  
  
FALSE  TRUE  
 1687   313  
> table(HireTrainApr10$Major == "IT" & HireTrainApr10$Hired == "Yes")  
  
FALSE  TRUE  
 1702   298  
> table(HireTrainApr10$Major == "Stats" & HireTrainApr10$Hired == "Yes")  
  
FALSE  TRUE  
 1688   312
```

- Major usually plays a big role in the hiring decision however, the tables to the left show us that the number of candidates that got hired from all four majors is relatively the same across all four majors (CS, DataScience, IT, and Stats)
- This shows us that major doesn't necessarily have a very big impact on the hiring decision.



Analysis

- As seen previously, none of the individual factors (coding, impression, major and college) seem to have a huge impact individually on the hiring decision
- This leads me to believe that a combination of these factors is what will play a role in deciding who gets hired versus who doesn't
- Using this knowledge, I made more tables regarding a combination of these factors, which allowed me to understand what combination of factors actually make an impact in who gets hired.



BestCollege vs Major vs Hired

```
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'CS'],)$Hired)
  No Yes
  51  44
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'IT'],)$Hired)
  No Yes
  66  35
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'DataScience'],)$Hired)
  No Yes
  51  54
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'Stats'],)$Hired)
  No Yes
  50  51
> |
```



BYU vs Major vs Hired

```
-- --  
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'CS'],]$Hired)
```

```
No Yes  
33  84
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'Stats'],]$Hired)
```

```
No Yes  
35  67
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'IT'],]$Hired)
```

```
No Yes  
41  58
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'DataScience'],]$Hired)
```

```
No Yes  
42  54
```

```
> |
```



Peters vs Major vs Hired

```
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'CS',]$Hired)

No Yes
32  70
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'Stats',]$Hired)

No Yes
39  62
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'IT',]$Hired)

No Yes
30  65
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'DataScience',]$Hired)

No Yes
28  71
> |
```



PJIT vs Major vs Hired

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'DataScience'],)$Hired)
```

```
No Yes  
32 61
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'Stats'],)$Hired)
```

```
No Yes  
34 55
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'IT'],)$Hired)
```

```
No Yes  
28 61
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'CS'],)$Hired)
```

```
No Yes  
32 56
```

```
> |
```




Redbrick vs Major vs Hired

```
> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'CS'],]$Hired)

No Yes
33  78

> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'IT'],]$Hired)

No Yes
27  79

> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'Stats'],]$Hired)

No Yes
31  77

> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'DataScience'],]$Hired)

No Yes
30  73

> |
```



Analysis

- As seen in the previous four slides, I made tables of each major in each school to see how many people got hired for each section.
- We can see that the number of people that got hired from BestCollege is lower as compared to the other four colleges, which leads me to believe that there are not a lot of candidates that get hired from BestCollege.
- For the remaining colleges, the number of people hired remain relatively close, showing me that out of the remaining four colleges, where the candidate went does not make that big of an impact.
- However, one key thing that is crucial to be noticed is the fact that for each college, all the majors hired seem to have relatively close numbers, showing me that the major of the candidate didn't really make a difference in whether the candidate was hired or not
- After focusing on these two aspects, I could notice that these didn't seem to make too much of an impact on the hiring decision, which lead me to believe that the coding skills and the impression is what makes the most impact.
- As seen from the forty tables in the beginning, I noticed the trends between coding skills and impressions, as that made the most impact. I saw that weak coding skills didn't usually get people hired while "Excellent" and "OK" skills did. I also noticed that "Shy" people tended to not get hired, while "Outgoing", "Nerdy" and "Confident" got hired. Using this information (from the forty tables in the beginning) and this information (major and college doesn't have a great impact on hiring decision), I created the decision vector that can be seen on the next slide.

Cross Validation

```
> myprediction <- HireTrainApr10
> 
> decision <- rep('No', nrow(myprediction))
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "Excellent"] <- 'Yes'
> 
> decision[myprediction$Impression == "outgoing" & myprediction$Coding == "Excellent"] <- 'Yes'
> 
> 
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "Excellent"] <- 'Yes'
> 
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "Excellent"] <- 'Yes'
> 
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "OK"] <- 'Yes'
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "OK"] <- 'Yes'
> 
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "OK"] <- 'Yes'
> 
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "OK"] <- 'Yes'
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "weak"] <- 'No'
> 
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "weak"] <- 'No'
> 
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "weak"] <- 'No'
> 
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "weak"] <- 'No'
> 
> myprediction$Hired <- decision
> 
> error <- mean(HireTrainApr10$Hired != myprediction$Hired)
> 
> error
[1] 0.103
> |
```

- Using the idea that impression and coding make the most impact on getting a candidate hired, I came up with the decision vector seen on the left.
- Testing these conditions/checks on the testing dataset provided to us gave me an error value of 10.3% which is not the best, however, it gives us a general idea of what gets a person hired which is what we were looking for in the first place

Kaggle Submission Code

```
> test_challenge1 <- read.csv("C:/Users/msraa/Downloads/test_challenge1.csv", stringsAsFactors=TRUE)
> view(test_challenge1)
> sample_submission_challenge1 <- read.csv("C:/Users/msraa/Downloads/sample_submission_challenge1.csv", stringsAsFactors=TRUE)
> view(sample_submission_challenge1)
> test <- test_challenge1
> submission <- sample_submission_challenge1
> myprediction <- test
> decision <- rep('No', nrow(myprediction))
>
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "Excellent"] <- 'Yes'
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "weak"] <- 'No'
>
```

Kaggle Submission Continued

```
> submission$Prediction <- decision  
> submission
```

	Id	Prediction
1	1	Yes
2	2	No
3	3	No
4	4	Yes
5	5	No
6	6	Yes
7	7	Yes
8	8	Yes
9	9	No
10	10	No
11	11	No
12	12	Yes
13	13	No
14	14	Yes
15	15	Yes
16	16	No
17	17	Yes
18	18	Yes
19	19	No
20	20	Yes
21	21	No
22	22	No
23	23	No
24	24	Yes
25	25	Yes
26	26	No

To create the submission file, I used the following command:

```
> write.csv(submission, "submission.csv", row.names = FALSE)
```

Lines continue in a similar fashion

(listing all the predictions for 500 rows)



Analysis: All Code

```
> table(HireTrainApr10[HireTrainApr10$Coding == "Excellent" & HireTrainApr10$Hired=='Yes',])
```

```
> table(HireTrainApr10$Coding == "Excellent" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1369	631

```
> table(HireTrainApr10$Coding == "weak" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1931	69

```
> table(HireTrainApr10$Coding == "OK" & HireTrainApr10$Hired == "Yes")
```

FALSE	TRUE
1445	555

```
> |
```

Analysis: All Code Continued

```
> table(HireTrainApr10$Impression == "Confident" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1679 321
```

```
> table(HireTrainApr10$Impression == "Nerdy" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1677 323
```

```
> table(HireTrainApr10$Impression == "Outgoing" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1675 325
```

```
> table(HireTrainApr10$Impression == "Shy" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1714 286
```

```
> |
```

```
2000
```

```
> table(HireTrainApr10$College == "BestCollege" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1816 184
```

```
> table(HireTrainApr10$College == "BYU" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1737 263
```

```
> table(HireTrainApr10$College == "Peters" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1732 268
```

```
> table(HireTrainApr10$College == "PJIT" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1767 233
```

```
> table(HireTrainApr10$College == "Redbrick" & HireTrainApr10$Hired == "Yes")
```

```
FALSE TRUE  
1693 307
```

```
> |
```


Analysis: All Code Continued

```
> table(HireTrainApr10$Major == "CS" & HireTrainApr10$Hired == "Yes")
```

```
FALSE  TRUE  
1668   332
```

```
> table(HireTrainApr10$Major == "DataScience" & HireTrainApr10$Hired == "Yes")
```

```
FALSE  TRUE  
1687   313
```

```
> table(HireTrainApr10$Major == "IT" & HireTrainApr10$Hired == "Yes")
```

```
FALSE  TRUE  
1702   298
```

```
> table(HireTrainApr10$Major == "Stats" & HireTrainApr10$Hired == "Yes")
```

```
FALSE  TRUE  
1688   312
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'CS',]$Hired)
```

```
No Yes  
51  44
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'IT',]$Hired)
```

```
No Yes  
66  35
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'DataScience',]$Hired)
```

```
No Yes  
51  54
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BestCollege" & HireTrainApr10$Major == 'Stats',]$Hired)
```

```
No Yes  
50  51
```

```
> |
```

Analysis: All Code Continued

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'CS'],]$Hired)
```

```
No Yes  
33 84
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'Stats'],]$Hired)
```

```
No Yes  
35 67
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'IT'],]$Hired)
```

```
No Yes  
41 58
```

```
> table(HireTrainApr10[HireTrainApr10$College == "BYU" & HireTrainApr10$Major == 'DataScience'],]$Hired)
```

```
No Yes  
42 54
```

```
> |
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'CS'],]$Hired)
```

```
No Yes  
32 70
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'Stats'],]$Hired)
```

```
No Yes  
39 62
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'IT'],]$Hired)
```

```
No Yes  
30 65
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Peters" & HireTrainApr10$Major == 'DataScience'],]$Hired)
```

```
No Yes  
28 71
```

```
> |
```

Analysis: All Code Continued

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'DataScience'],)$Hired)
```

```
No Yes  
32 61
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'Stats'],)$Hired)
```

```
No Yes  
34 55
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'IT'],)$Hired)
```

```
No Yes  
28 61
```

```
> table(HireTrainApr10[HireTrainApr10$College == "PJIT" & HireTrainApr10$Major == 'CS'],)$Hired)
```

```
No Yes  
32 56
```

```
> |
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'CS'],)$Hired)
```

```
No Yes  
33 78
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'IT'],)$Hired)
```

```
No Yes  
27 79
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'Stats'],)$Hired)
```

```
No Yes  
31 77
```

```
> table(HireTrainApr10[HireTrainApr10$College == "Redbrick" & HireTrainApr10$Major == 'DataScience'],)$Hired)
```

```
No Yes  
30 73
```

```
> |
```

Analysis: All Code Continued



```
> myprediction <- HireTrainApr10
>
> decision <- rep('No', nrow(myprediction))
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "Excellent"] <- 'Yes'
>
>
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "Excellent"] <- 'Yes'
>
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "OK"] <- 'Yes'
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "OK"] <- 'Yes'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "OK"] <- 'Yes'
> decision[myprediction$Impression == "Confident" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Shy" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Outgoing" & myprediction$Coding == "weak"] <- 'No'
>
> decision[myprediction$Impression == "Nerdy" & myprediction$Coding == "weak"] <- 'No'
>
> myprediction$Hired <- decision
>
> error <- mean(HireTrainApr10$Hired != myprediction$Hired)
>
> error
[1] 0.103
> |
```