HR Analytics Case Study

Background – HR Analytics Case Study

A large company named XYZ, employs, at any given point of time, around 4000 employees. However, every year, around 15% of its employees leave the company. Since the attrition level is too high, the management wants to use predictive modelling to bring it down.

Hence, the objectives of the analysis are to:

- Help company XYZ identify current employees that are very likely to leave
- Recommend ways for company XYZ to decrease its attrition level in the future

The analysis is divided into three parts:

- Data Understanding Source of data, patterns in the data
- Predictive modelling of attrition
- Recommending ways for company XYZ to decrease its level of attrition



Data Understanding – 4 broad sources of data

The data received for the analysis can be divided into 4 broad categories -

- General Data General data, acquired from HR
- Employee Survey Data Data collected from yearly employee survey
- Manager Survey Data Data collected from yearly manager survey
- Biometric Data Daily in and out times for each employee, collected using biometric attendance machines

General Data
Age
Attrition (Yes/No)
Department
Education Field

Manager Survey Data						
Job Involvement						
Performance Rating						

Employee Survey Data						
Environment Satisfaction						
Job Satisfaction						
Work Life Balance						

Biometric Data					
In Time					
Out Time					

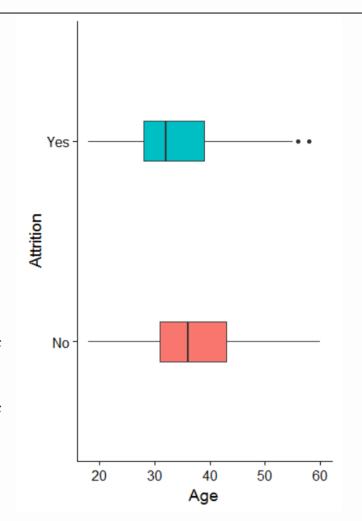


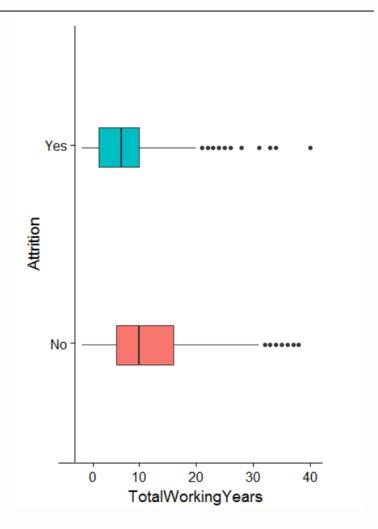
Age

- Employees aged 36 years and above are more likely to stay*
- Employees aged 32 years and below are more likely to leave*

Experience

- Employees that have worked for a total of 10 years or more are more likely to stay*
- Employees that have worked for a total of 7 years or less are more likely to leave*





^{*}Coefficients of the variables Age and TotalWorkingYears are significant. Among attritions, median age = 32 and median exp. = 7. Among non-attritions, median age = 36 and median exp. = 10

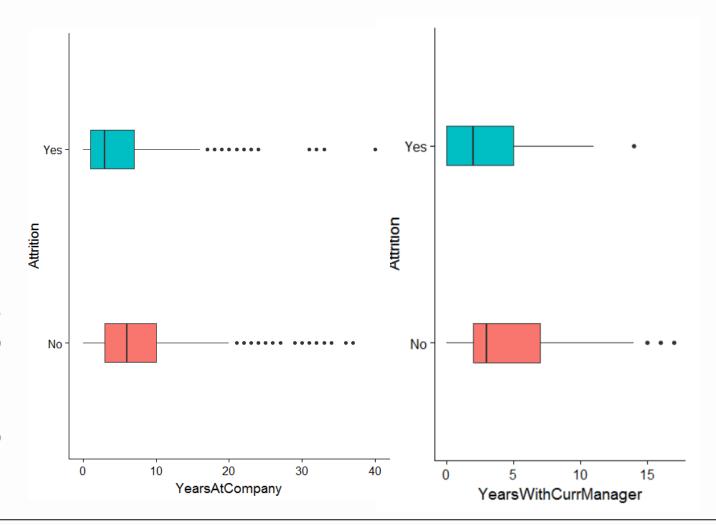


YearsAtCompany

- Employees those have spent >5 years are more likely to stay*
- Employees those are <5 years in this company are more likely to leave*

Years with Current Manager

- Employees that have spent 3 years or more under the same manager are more likely to stay*
- Employees that have spent 2 years or less under the same manager are more likely to leave*



^{*}Coefficients of the variables TrainingTimesLastYear and YearsWithCurrManager are significant. Rest of the data is based on means/medians etc.

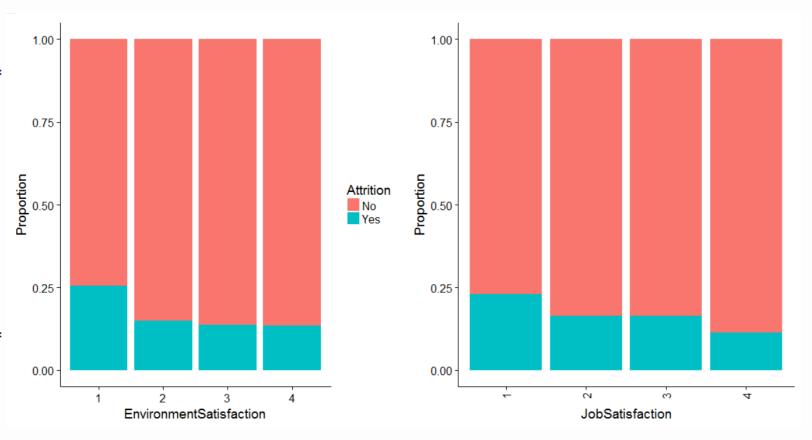


Job Satisfaction

- Employees that have medium, high or very high levels of job satisfaction, are more likely to stay*
- Employees that have low levels of job satisfaction, are more likely to leave*

Environment Satisfaction

- Employees that have medium, high or very high levels of environment satisfaction, are more likely to stay*
- Employees that have low levels of environment satisfaction, are more likely to leave*



^{*}Coefficients of the variables JobSatisfaction and EnvironmentSatisfaction are significant. Employees were asked to report their job satisfaction and work environment satisfaction levels in a survey.

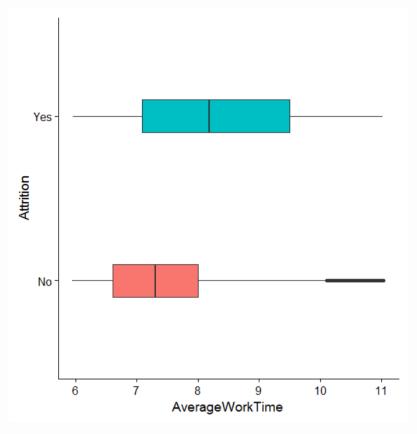


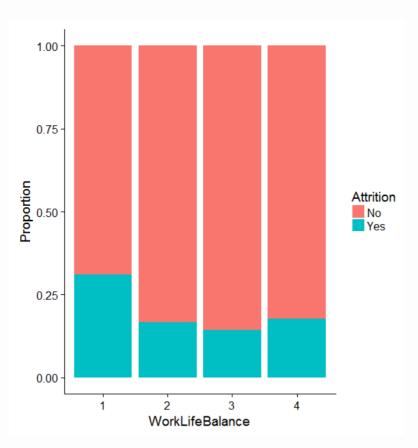
Average Work Hours

- Employees that, on average work for 7.3 hours or less, are more likely to stay*
- Employees that, on average work for 8.2 hours or more, are more likely to leave*

Work Life Balance

- Employees that rated their work life balance as good, better or best, are more likely to stay**
- Employees that rated their work life balance as bad, are more likely to leave**





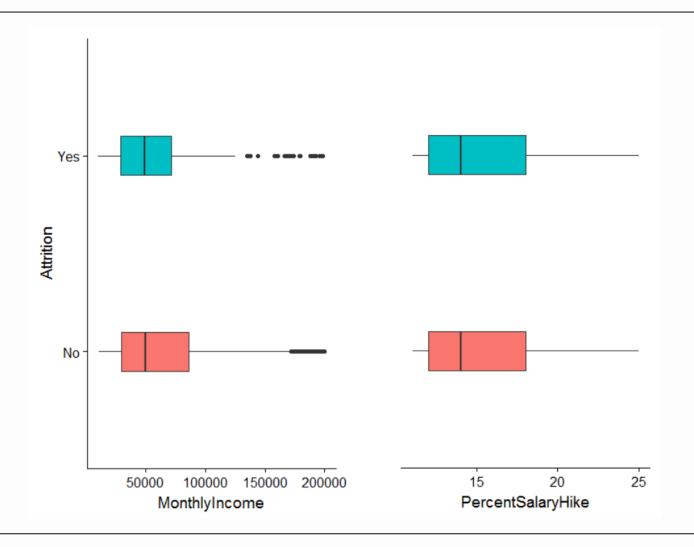
^{*}Coefficients of the variables AverageWorkTime and WorkLIfeBalance are significant. Average work hours data is based on means/medians etc. Employees were asked to report their level of work life balance in a survey.

Page -7-



Recommendations – Factors that surprisingly don't affect attrition

Monthly Income and Percent Salary
 Hike do not affect attrition*



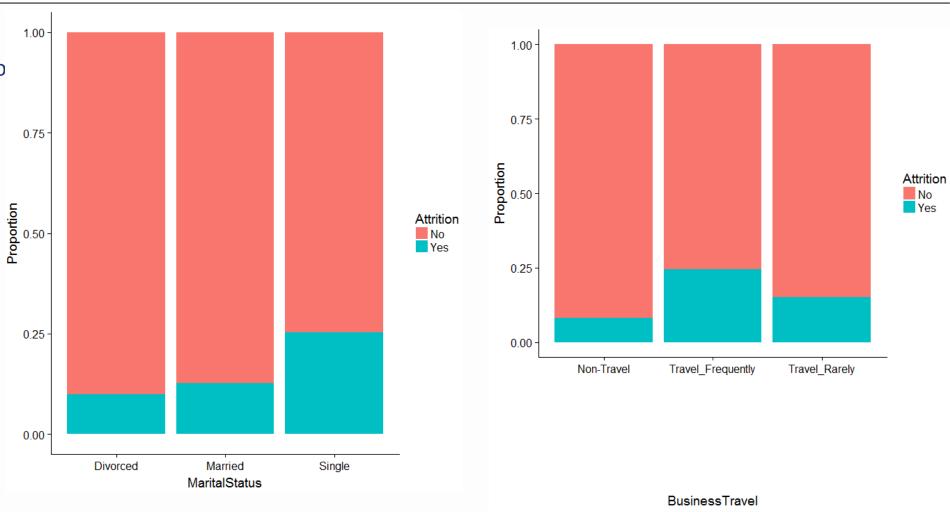


^{*}Coefficients of these variables are not significant

Recommendations – Factors that affects attrition

Employees taking frequent Business travels are likely to leave*

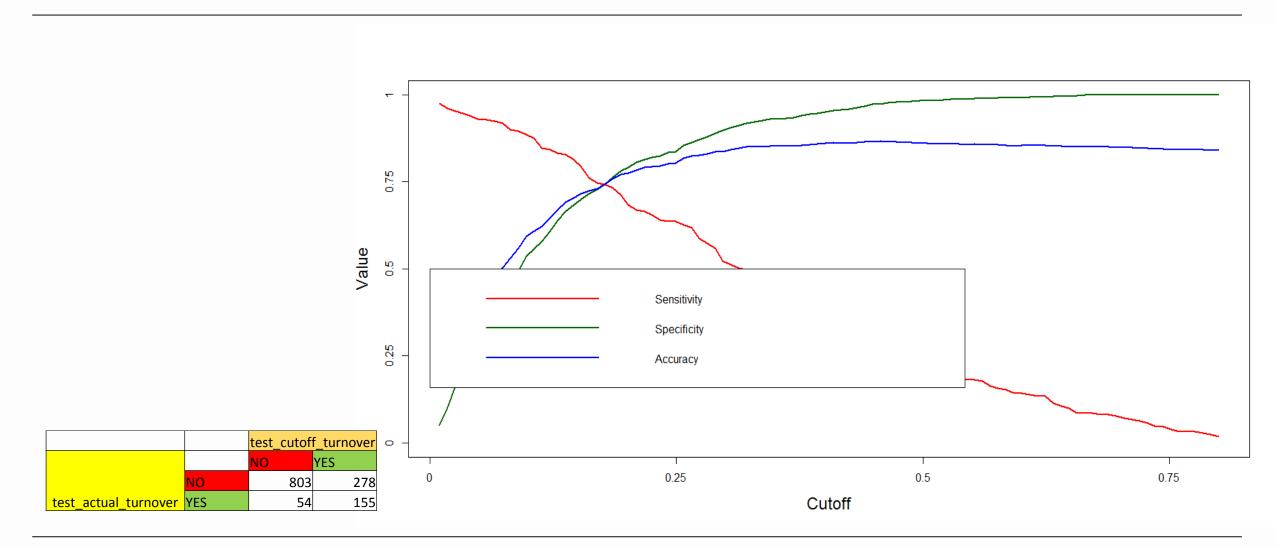
 Singles/Bachelors are likely to leave*



^{*}Coefficients of these variables are not significant



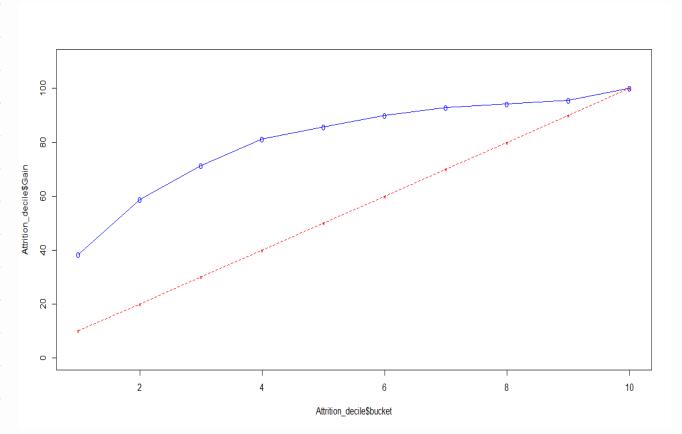
Model testing plots





Lift and Gain Chart

bucket	tot	tal	totalresp	Cumre	sp Gain	Cumli	Ft
1	1	129	80.0	80.	38.3	3.83	
2	2	129	43.0	123	58.9	2.94	
3	3	129	26.0	149	71.3	2.38	
4	4	129	21.0	170	81.3	2.03	
5	5	129	9.00	179	85.6	1.71	
6	6	129	9.00	188	90.0	1.50	
7	7	129	6.00	194	92.8	1.33	
8	8	129	3.00	197	94.3	1.18	
9	9	129	3.00	200	95.7	1.06	
10 1	0	129	9.00	209	100	1.00	



• By addressing top 40 % of employees predicted by this model, we can target 81% turnover candidates.



Recommendations

Current employees:

- Work life balance should be improved
- Work environment should be improved. HR s can arrange for fun events involving family visits. Organise games, competitions, trips.
- The manager of an employee should not be changed very often
- Employees should be provided relevant training regularly, especially for its younger/singles employees and should be engaged in R&D works for their relevant departments and must be awarded.
- Work load should be reduced by properly managing the work among the employees. Few employees seems to stretch their work hours over the entire year.
- Give timely promotion or introduce Bonus.

Future employees (changes in hiring process):

- The company should follow either one of the strategies given below
 - Hire older people with decent work experience.
 - Avoid Hiring candidates who seem to switch too frequently.
- It could also opt for a combination of the two

