



2018-1004 IST 687 Applied Data Science

Group A Team Project:
Hyatt Hotels NPS Analysis

Team Members:
Charles Anyanwu, Will Holt, Matt Maloney

Submitted:
December 2018



Table of Contents

Background and Understanding	3
Executive summary	3
Business Questions	4
General Questions	4
NPS Contributors	4
Data load	5
Description of the data set	7
Data Cleansing	8
Data Quality Checks	9
Descriptive Statistics	10
Maps	25
Top and Bottom 10 Profiles	27
Models	30
Linear Model	30
Association Rules Mining	31
Support Vector Machine	34
Best Model	36
Business Question Insights	36
Recommendations	42

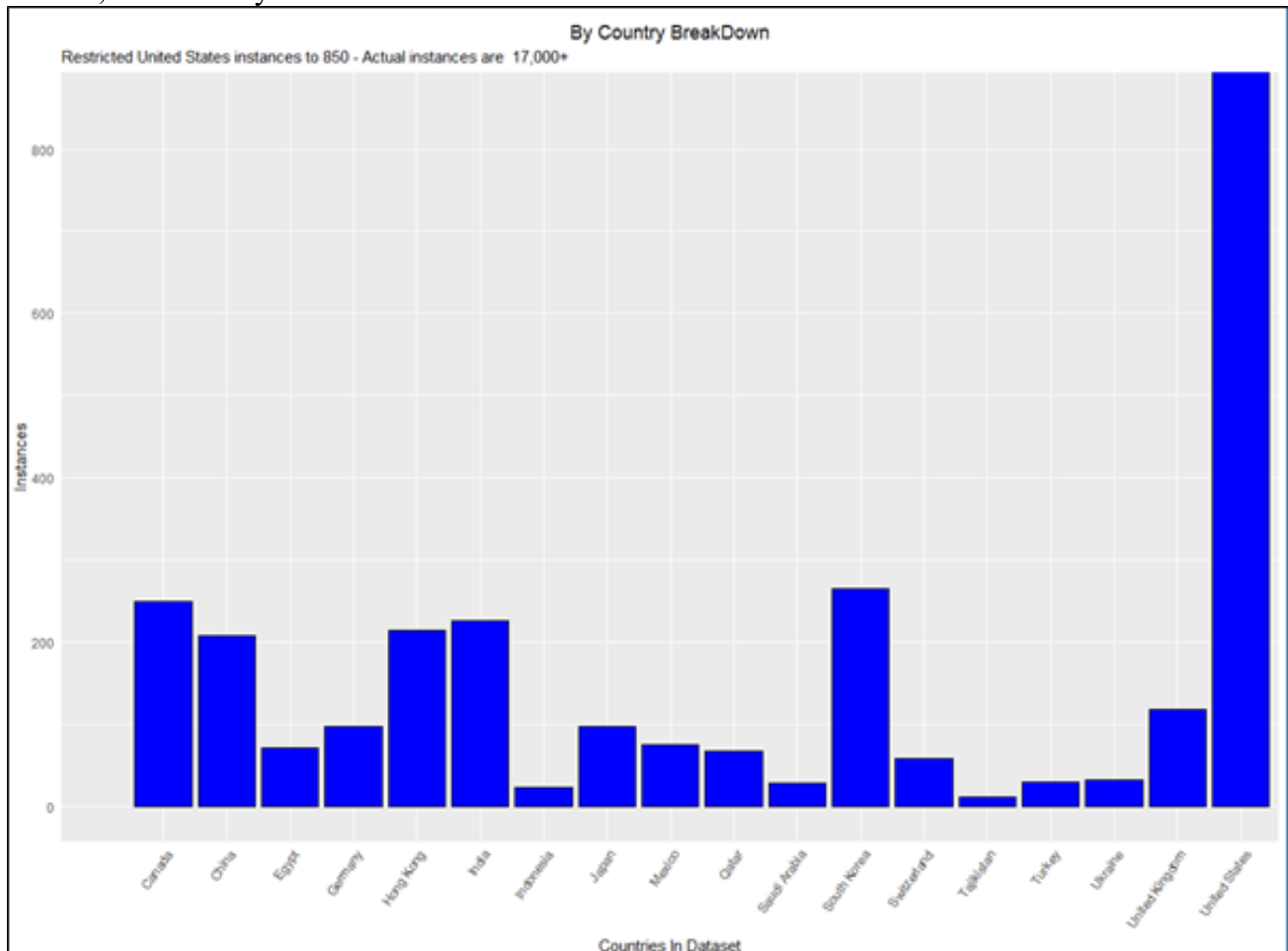
Background and Understanding

The Hyatt Hotels Corporation (Hyatt) is one of the largest multinational hospitality brands in the world. Headquartered in the United States, Hyatt operates more than 600 hotels, resorts, and vacation properties in more than fifty countries.

Hyatt requested Orange Consulting (Orange) to provide Hyatt with key takeaways after analyzing survey data from guests staying at the company's hotels during February 2014. Driven by a few important business questions, our analysis will deliver actionable insight into the main factors that contribute to Net Promoter Score.

Executive summary

Originally we analyzed data to see if any abnormalities existed. We quickly observed a great degree of skewness when the data was broken out into countries where Hyatt properties are located, as shown by the visual below.





The team consensus was to restrict the dataset to United States.

Based on analysis by using descriptive and different modelling technique such as Association Rules, SVM, etc. Recommendations below were made.

- Services must be packaged and tailored to specific Tiers to keep revenue streams.
- Improvements need to be made in attracting more customer feedback
- Hotel should focus on maintaining a high standard for hotel rooms and hotel condition
- A fair number of revenue and frequency are from business travelers in the age group from age group 36 to 65. This group has to be catered to for continuous patronage.

Business Questions

General Questions

Hyatt would like to explore and understand the following business questions. Orange addresses these questions through the document, but the direct answer is provided in the Business Questions Insights. Please see the Table of Contents for location.

- What is the overall NPS?
 - Provide an overview of the count and percentage for each category of Promoter, Passive and Detractor.
- What is the NPS across different demographics
 - By US Region
 - By State
 - 31 states + Washington D.C.
 - By Hotel (provides same insight as by city)
 - By Gender
 - By Purpose of Visit
- What days of the week are most popular amongst guests?
- What attributes are best at predicting if a guest will be a promoter?

NPS Contributors

- What are the key drivers for Promoters / Detractors / Passives

Data load

The dataset that was used by the team had 19,342 observations (nrows) and 58 variables (ncol). Using `str()` on ProjectSurvey, below is the output:

```
> str(ProjectSurvey)
'data.frame': 19342 obs. of 58 variables:
 $ MARKET_GROUP_C      : Factor w/ 7 levels "CONVENTION","DISCOUNT",...: 3 3 3 2 3 3 3 3 2 3 ...
 $ ROOM_TYPE_CODE_C     : Factor w/ 164 levels "", "1B2A", "1B2C",...: 100 100 20 130 20 65 20 129 100 129 ...
 $ WALK_IN_FLG_C        : Factor w/ 2 levels "N","Y": 1 1 1 1 1 1 1 1 1 1 ...
 $ CHECK_IN_DATE_C      : Factor w/ 37 levels "1/11/2014","1/26/2014",...: 5 32 24 6 19 20 4 10 16 16 ...
 $ CHECK_OUT_DATE_C     : Factor w/ 72 levels "1/16/2014","1/28/2014",...: 15 10 37 55 23 51 30 17 43 22 ...
 $ LENGTH_OF_STAY_C     : int 10 7 14 22 5 7 7 7 10 7 ...
 $ NUMBER_OF_ROOMS_C    : int 1 1 1 1 1 1 1 1 1 1 ...
 $ ADULT_NUM_C          : int 2 2 2 2 2 2 2 2 2 2 ...
 $ CHILDREN_NUM_C       : int NA NA NA NA NA NA NA NA NA ...
 $ POV_CODE_C           : Factor w/ 2 levels "BUSINESS","LEISURE": 1 1 1 2 1 1 1 1 1 1 ...
 $ QUOTED_RATE_C        : num 225 158 158 110 201 ...
 $ RESERVATION_DATE_R   : Factor w/ 295 levels "1/1/2014 0:00",...: 204 207 257 40 66 81 123 97 104 110 ...
 $ ENTRY_TIME_R         : Factor w/ 16580 levels "0:00:04","0:00:12",...: 13673 15568 11299 14066 15687 13616 195 190 13939 186 ...
 $ ENTRY_HOTEL_CODE_R   : Factor w/ 210 levels "ABE2B","ABQ2A",...: 195 195 195 110 195 195 195 195 110 195 ...
 $ LAST_CHANGE_DATE_R   : Factor w/ 258 levels "1/1/2014 0:00",...: 180 183 223 124 64 79 121 95 105 108 ...
 $ ROOM_TYPE_CODE_R     : Factor w/ 144 levels "1B2A","1B2C",...: 56 88 88 88 17 17 17 17 56 56 ...
 $ STATE_R              : Factor w/ 112 levels "", "AB","AC","AE",...: 1 1 1 1 1 25 25 1 12 1 ...
 $ GUEST_COUNTRY_R      : Factor w/ 92 levels "", "SCO","AC",...: 27 27 87 87 33 88 88 87 33 27 ...
 $ PACE_CATEGORY_R      : Factor w/ 9 levels "0-3 Days","15-21 Days",...: 8 8 8 8 8 8 4 6 4 ...
 $ PACE_R               : Factor w/ 284 levels "-1","0","1","10",...: 138 134 96 24 9 276 237 247 245 240 ...
 $ REVENUE_USD_R        : num 2250 1108 2217 2429 1406 ...
 $ Guest_Country_H      : Factor w/ 86 levels "", "American Samoa",...: 81 81 81 81 53 81 81 81 27 81 ...
 $ Gender_H             : Factor w/ 4 levels "", "Female","Male",...: 2 3 2 3 3 3 3 3 2 ...
 $ Age_Range_H          : Factor w/ 8 levels "", "18-25","26-35",...: 7 5 7 7 6 2 7 6 6 ...
 $ Language_H           : Factor w/ 13 levels "Arabic","Chinese Simplified",...: 4 4 4 4 4 4 4 4 6 4 ...
 $ Likelihood_Recommend_H : int 10 10 10 10 9 10 10 10 10 ...
 $ Overall_Sat_H        : int 9 9 10 10 9 9 8 10 10 10 ...
 $ Guest_Room_H         : int 10 10 10 10 10 9 9 10 8 10 ...
 $ Tranquility_H        : int 10 9 10 NA 10 9 NA NA 10 10 ...
 $ Condition_Hotel_H    : int 9 10 10 9 10 9 10 10 9 10 ...
 $ Customer_SVC_H       : int 10 10 10 10 10 10 10 9 10 ...
 $ Staff_Cared_H        : int 10 10 10 NA 10 10 NA NA 10 10 ...
 $ Internet_Sat_H       : int 10 7 NA NA NA 6 NA 9 9 ...
 $ Check_In_H           : int 10 9 10 NA 9 10 NA NA 8 10 ...
 $ F.B_FREQ_H           : int NA 1 NA NA NA 2 NA NA 3 3 ...
 $ F.B.Overall_Experience_H : num NA 9 NA NA NA ...
 $ Property_ID_PL       : int 3275 3275 3275 3275 3275 3275 3275 3275 3275 ...
 $ Hotel.Name.Long_PL   : Factor w/ 195 levels "Andaz Liverpool Street, London",...: 185 185 185 185 185 185 185 185 185 ...
 $ Hotel.Name.Short_PL  : Factor w/ 195 levels "AZ Liverpool Street",...: 175 175 175 175 175 175 175 175 175 ...
 $ Award.Category_PL    : int 2 2 2 2 2 2 2 2 2 ...
 $ City_PL              : Factor w/ 153 levels "Addison","Ahmedabad",...: 138 138 138 138 138 138 138 138 138 ...
 $ Country_PL           : Factor w/ 18 levels "Canada","China",...: 3 3 3 3 3 3 3 3 3 ...
 $ Ops.Region_PL        : Factor w/ 4 levels "Americas","ASPAC",...: 3 3 3 3 3 3 3 3 3 ...
 $ Property.Latitude_PL : num 28 28 28 28 28 ...
 $ Property.Longitude_PL : num 34.4 34.4 34.4 34.4 34.4 ...
 $ Currency_PL          : Factor w/ 18 levels "CAD","CHF","CNY",...: 4 4 4 4 4 4 4 4 4 ...
 $ Dom.Int.l_PL         : Factor w/ 2 levels "Domestic","International": 2 2 2 2 2 2 2 2 2 ...
 $ Brand_PL             : Factor w/ 7 levels "Andaz","Grand Hyatt",...: 6 6 6 6 6 6 6 6 6 ...
 $ Club.Type_PL         : Factor w/ 4 levels "", "Club","Grand Club",...: 4 4 4 4 4 4 4 4 4 ...
 $ Region_PL            : Factor w/ 4 levels "Americas","Asia Pacific",...: 4 4 4 4 4 4 4 4 4 ...
 $ Category_PL          : Factor w/ 2 levels "Full Service",...: 1 1 1 1 1 1 1 1 1 ...
 $ Type_PL              : Factor w/ 5 levels "Business","Convention",...: 4 4 4 4 4 4 4 4 4 ...
 $ Class_PL             : Factor w/ 3 levels "Luxury Class",...: 2 2 2 2 2 2 2 2 2 ...
 $ Location_PL          : Factor w/ 4 levels "Airport","Resort",...: 2 2 2 2 2 2 2 2 2 ...
 $ Relationship_PL      : Factor w/ 3 levels "Franchised","Managed",...: 2 2 2 2 2 2 2 2 2 ...
 $ GP_Tier              : Factor w/ 11 levels "", "CARD","Courtesy",...: 7 1 7 5 1 1 1 1 1 7 ...
 $ Booking_Channel      : Factor w/ 3 levels "Electronic Distribution",...: 3 3 3 2 3 3 3 3 2 3 ...
 $ NPS_Type             : Factor w/ 3 levels "Detractor","Passive",...: 3 3 3 3 3 3 3 3 3 3 ...
```

After restricting dataset to Hyatt operations in the United States, we reduced the data set 17,439 observations(nrows), but our team added additional variables. We also changed the names of the variables in order to more easily identify them throughout the analysis. While we did not explore all of the variables venues, we feel that our variable selection fairly represents the data set and offers important insight. Below is the updated and formatted data set used for analysis.


```
> str(ProjectSurveyUSA)
'data.frame': 17439 obs. of 66 variables:
 $ MarketGroup      : Factor w/ 7 levels "CONVENTION","DISCOUNT",...: 4 4 5 4 1 1 2 1 4 7 ...
 $ RTC_CheckOut     : Factor w/ 164 levels "", "1B2A", "1BCN",...: 161 162 160 162 161 160 162 162 160 161 ...
 $ WalkIn           : Factor w/ 2 levels "N","Y": 1 1 1 1 1 1 1 1 1 1 ...
 $ CheckInDate      : Date, format: "2014-02-28" "2014-02-14" "2014-02-03" "2014-02-15" ...
 $ CheckOut         : Date, format: "2014-03-04" "2014-02-15" "2014-02-06" "2014-02-17" ...
 $ LOS              : num 4 1 3 2 3 2 1 5 2 5 ...
 $ NumRooms         : num 1 1 1 1 1 1 1 1 1 1 ...
 $ Adult            : num 1 2 1 1 1 1 2 2 2 2 ...
 $ Children         : num 1.65 1.65 1.65 1.65 1.65 ...
 $ PurposeOfVisit   : Factor w/ 2 levels "BUSINESS","LEISURE": 1 2 1 2 1 1 1 1 1 1 ...
 $ QuotedRate       : num 229 340 183 209 209 ...
 $ ResvDate         : Date, format: "2014-01-28" "2014-01-28" "2014-01-28" "2014-01-29" ...
 $ ResvTime         : Factor w/ 16580 levels "0:00:04","0:00:12",...: 11302 11666 13044 742 1483 1837 4607 4694 350 14938 ...
 $ EntryHotelCode   : Factor w/ 210 levels "ABEZB","ABQZA",...: 165 165 165 164 165 165 150 165 165 202 ...
 $ ResvDateChange   : Date, format: "2014-01-28" "2014-01-31" "2014-01-28" "2014-01-30" ...
 $ RTC_Book         : Factor w/ 144 levels "1B2A","1BCN",...: 139 142 139 140 139 140 140 140 140 139 ...
 $ State            : Factor w/ 112 levels "", "AB", "AC", "AE",...: 52 25 95 78 76 76 104 71 25 25 ...
 $ GuestCountryOrigin: Factor w/ 92 levels "", "SCO", "AC",...: 88 88 88 14 88 88 88 88 88 88 ...
 $ FutureResvDaysCat : Factor w/ 9 levels "0-3 Days","15-21 Days",...: 4 2 5 2 2 2 7 2 1 3 ...
 $ FutureResvDays   : num 183 81 241 81 103 114 26 103 113 167 ...
 $ Revenue          : num 916 340 550 418 627 ...
 $ GuestCountryHome : Factor w/ 86 levels "", "American Samoa",...: 83 83 83 12 83 83 83 83 83 83 ...
 $ Gender           : Factor w/ 4 levels "", "Female", "Male",...: 2 3 3 3 3 2 2 3 3 3 ...
 $ AgeRange         : Factor w/ 8 levels "", "18-25", "26-35",...: 5 5 5 6 5 4 5 6 5 4 ...
 $ Language         : Factor w/ 13 levels "Arabic","Chinese Simplified",...: 4 4 4 4 4 4 4 4 4 4 ...
 $ LR_Score         : int 10 9 5 9 10 10 10 9 10 10 ...
 $ OS_Score         : num 10 8 6 9 10 10 10 9 9 9 ...
 $ GR_Score         : num 10 7 10 10 10 10 10 10 9 10 ...
 $ T_Score          : num 8.77 8.77 8.77 10 8.77 ...
 $ HC_Score         : num 10 9 8 10 9 10 10 9 9 10 ...
 $ CS_Score         : num 10 8 3 10 10 ...
 $ SC_Score         : num 9.04 9.04 9.04 9 9.04 ...
 $ IS_Score         : num 8.75 8.75 8.75 9 8.75 ...
 $ CI_Score         : num 9.28 9.28 9.28 10 9.28 ...
 $ FB_Frequency     : num 1.47 1.47 1.47 2 1.47 ...
 $ FB_Score         : num 8.63 8.63 8.63 9.5 8.63 ...
 $ PropertyID       : int 3328 3328 3328 3328 3328 3328 3328 3328 3328 3328 ...
 $ HotelName        : Factor w/ 195 levels "Andaz Liverpool Street, London",...: 176 176 176 176 176 176 176 176 176 ...
 $ HotelAbbrev      : Factor w/ 195 levels "AZ Liverpool Street",...: 167 167 167 167 167 167 167 167 167 ...
 $ AwardCat         : int 5 5 5 5 5 5 5 5 5 5 ...
 $ HotelCity        : Factor w/ 153 levels "Addison","Ahmedabad",...: 113 113 113 113 113 113 113 113 113 ...
 $ HotelCountry     : Factor w/ 18 levels "Canada","China",...: 18 18 18 18 18 18 18 18 18 ...
 $ OperatingRegion  : Factor w/ 4 levels "Americas","ASPAC",...: 1 1 1 1 1 1 1 1 1 1 ...
 $ Latitude         : num 28.4 28.4 28.4 28.4 28.4 ...
 $ Longitude        : num -81.5 -81.5 -81.5 -81.5 -81.5 ...
 $ Currency         : Factor w/ 18 levels "CAD","CHF","CNY",...: 18 18 18 18 18 18 18 18 18 18 ...
 $ ValidValues      : Factor w/ 2 levels "Domestic","International": 1 1 1 1 1 1 1 1 1 1 ...
 $ HotelBrand       : Factor w/ 7 levels "Andaz","Grand Hyatt",...: 6 6 6 6 6 6 6 6 6 6 ...
 $ Club             : Factor w/ 4 levels "", "Club", "Grand Club",...: 4 4 4 4 4 4 4 4 4 4 ...
 $ Region           : Factor w/ 4 levels "Americas","Asia Pacific",...: 1 1 1 1 1 1 1 1 1 1 ...
 $ Category         : Factor w/ 2 levels "Full Service",...: 1 1 1 1 1 1 1 1 1 1 ...
 $ Type             : Factor w/ 5 levels "Business","Convention",...: 4 4 4 4 4 4 4 4 4 4 ...
 $ Class            : Factor w/ 3 levels "Luxury Class",...: 2 2 2 2 2 2 2 2 2 2 ...
 $ Location         : Factor w/ 4 levels "Airport","Resort",...: 2 2 2 2 2 2 2 2 2 2 ...
 $ Relationship     : Factor w/ 3 levels "Franchised","Managed",...: 3 3 3 3 3 3 3 3 3 3 ...
 $ GPier            : Factor w/ 11 levels "", "CARD", "Courtesy",...: 7 1 11 1 7 7 11 7 6 7 ...
 $ Channel          : Factor w/ 3 levels "Electronic Distribution",...: 1 1 1 2 1 1 2 1 1 3 ...
 $ NPS              : Factor w/ 3 levels "Detractor","Passive",...: 3 3 1 3 3 3 3 3 3 3 ...
 $ CheckInDay       : chr "Friday" "Friday" "Monday" "Saturday" ...
 $ CheckOutDay      : chr "Tuesday" "Saturday" "Thursday" "Monday" ...
 $ Coords           : num 28.4 28.4 28.4 28.4 28.4 ...
 $ stateName        : chr "florida" "florida" "florida" "florida" ...
 $ USRegion         : chr "Southeast" "Southeast" "Southeast" "Southeast" ...
 $ Promoter         : num 1 1 0 1 1 1 1 1 1 1 ...
 $ Detractor        : num 0 0 1 0 0 0 0 0 0 0 ...
 $ Passive          : num 0 0 0 0 0 0 0 0 0 0 ...
```



Description of the data set

The data set contains survey data from guests staying at Hyatt hotels during February 2014. One of the most important data points across all observations is called Net Promoter Score (NPS), which measures customer experience and can help predict business growth. It provides a measurement of customer satisfaction.

Survey respondents are grouped as:

- Promoters (score 9-10) are loyal enthusiasts who will keep buying and refer others, fueling growth.
- Passives (score 7-8) are satisfied but unenthusiastic customers who are vulnerable to competitive offerings.
- Detractors (score 0-6) are unhappy customers who can damage the brand and impede growth through negative word-of-mouth.

§ Sources - <https://www.netpromoter.com/know>

While many of the variables will be used in our analysis, the following changes were made prior to beginning:

- Columns with a large number of NAs were given various treatments, including converting to the mean
- “Children_Num_C” were discarded from the beginning based on the numbers of NAs (15,874) within its column of data or Internet_Sat_H which had 9,649 NAs
- “Check_in_H” – 6,423
- F.B_FREQ_H - 8,605
- F.B_Overall_Exprience – 8,605
- Tranquility – 6,497
- Staff_cared - 6,422

The additional columns added helped in the data manipulation and explanation of the NPS score

Data Cleansing

In order to run some descriptive analysis, the team had to address the NAs by converting factors to numeric values, factors to data, and factors to characters where needed. Some of these methods were achieved by creating functions and data manipulations.

Before →

Summary(ProjectSurvey)												
MARKET_GROUP_C	ROOM_TYPE_CODE_C	WALK_IN_FLG_C	CHECK_IN_DATE_C	CHECK_OUT_DATE_C	LENGTH_OF_STAY_C	NUMBER_OF_ROOMS_C	ADULT_NUM_C	CHILDREN_NUM_C	POV_CODE_C	QUOTED_RATE_C	RESERVATION_DATE_R	
CONVENTION:3847	KING	:8105	N:19329	2/14/2014:1355	2/23/2014:1481	Min.:1.000	Min.:1	Min.:1.000	Min.:1.000	BUSINESS:15678	Min.:0	2/3/2014 0:00:617
DISCOUNT:4519	DOBL	:4011	Y:13	2/21/2014:1133	2/16/2014:1401	1st Qu.:1.000	1st Qu.:1	1st Qu.:1.000	1st Qu.:1.000	LEISURE:3664	1st Qu.:92	2/4/2014 0:00:587
OTHER:936	18KN	:986		2/28/2014:1083	2/9/2014:1086	Median:2.000	Median:1	Median:1.000	Median:2.000		Median:119	2/10/2014 0:00:582
RACK:5134	QKN	:856		2/15/2014:1005	2/21/2014:831	Mean:2.408	Mean:1	Mean:1.432	Mean:1.622		Mean:5340	2/6/2014 0:00:580
SPECIAL:1402	QUEN	:314		2/22/2014:919	2/28/2014:827	3rd Qu.:3.000	3rd Qu.:1	3rd Qu.:2.000	3rd Qu.:2.000		3rd Qu.:169	2/11/2014 0:00:546
VOLUME:2903	DLXX	:312		2/7/2014:895	2/15/2014:771	Max.:97.000	Max.:1	Max.:8.000	Max.:4.000		Max.:2213984	2/5/2014 0:00:522
WHOLESALE:601	(Other):4758			(Other):12952	(Other):12945	NA's:97	NA's:51	NA's:7	NA's:17654		(Other):15908	
ENTRY_TIME_R	ENTRY_HOTEL_CODE_R	LAST_CHANGE_DATE_R	ROOM_TYPE_CODE_R	STATE_R	GUEST_COUNTRY_R	PACE_CATEGORY_R	PACE_R	REVENUE_USD_R	Guest_Country_H			
11:18:16:5	OMGDS:11569	2/3/2014 0:00:634	KING:9145	:3342	UNITED STATES:15448	0-3 Days:5884	0	:1886	Min.:0.0	USA	:17221	
11:21:42:5	OMARO:3418	2/18/2014 0:00:626	DOBL:3895	TX:1837	:1964	8-14 Days:2758	1	:1381	1st Qu.:115.0	Canada	:359	
14:47:25:5	MARRO:1079	2/6/2014 0:00:622	18KN:1053	CA:1203	CANADA:484	31-60 Days:2726	2	:941	Median:210.0	Korea	:246	
10:18:58:4	BUSPH:175	2/5/2014 0:00:605	QKN:854	FL:1058	SOUTH KOREA:233	4-7 Days:2672	3	:876	Mean:339.7	United Kingdom:178		
10:19:06:4	EURRC:135	2/4/2014 0:00:587	QUEN:486	IL:946	JAPAN:146	15-21 Days:2141	4	:808	3rd Qu.:398.0	Germany:150		
10:25:50:4	INDRO:93	2/12/2014 0:00:585	DLXN:357	NY:679	UNITED KINGDOM:141	22-30 Days:2063	5	:683	Max.:18415.0	Japan:148		
(Other):19315	(Other):2873	(Other):15683	(Other):3552	(Other):10277	(Other):1006	(Other):1898	(Other):12767	(Other):1040				
Gender_M	Age_Range_H	Language_H	Likelihood_Recommend_M	Overall_Sat_H	Guest_Room_H	Tranquility_M	Condition_Hotel_H	Customer_SVC_H	Staff_Cared_H	Internet_Sat_H		
Female:	213 46-55:15783	English:18382	Min.:1.0	Min.:1.000	Min.:1.000	Min.:1.000	Min.:1.00	Min.:1.000	Min.:1.000	Min.:1.000		
Male:	18271 56-65:14027	German:208	1st Qu.:8.0	1st Qu.:8.000	1st Qu.:8.000	1st Qu.:8.000	1st Qu.:8.00	1st Qu.:9.000	1st Qu.:9.000	1st Qu.:8.000		
Prefer not to answer:	347 26-35:12714	Japanese ja_JP:152	Mean:8.7	Mean:8.666	Mean:8.792	Mean:8.777	Mean:8.94	Mean:9.098	Mean:9.025	Mean:8.681		
	66-75:11262	Spanish (Americas):124	3rd Qu.:10.0	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:10.00	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:10.000		
	18-25:560	Chinese Simplified:95	Max.:10.0	Max.:10.000	Max.:10.000	Max.:10.000	Max.:10.00	Max.:10.000	Max.:10.000	Max.:10.000		
(Other):	591 (Other):223	NA's:21	NA's:188	NA's:17182	NA's:212	NA's:276	NA's:17108	NA's:10633				
Check_In_H	F.B.FREQ_H	F.B.Overall_Experience_H	Property_ID_PL	Hotel.Name.Long_PL		Hotel.Name.Short_PL		Award.Category_PL	City_PL	Country_PL		
Min.:1.000	Min.:1.000	Min.:0.000	Min.:3010	Hyatt Regency Jacksonville Riverfront:	579	HR Jacksonville:	579	Min.:1.000	Washington:1888	United States:17467		
1st Qu.:9.000	1st Qu.:1.000	1st Qu.:8.000	1st Qu.:3496	Hyatt Regency O'Hare:	579	HR O'Hare:	579	1st Qu.:1.000	Jacksonville:667	South Korea:265		
Median:10.000	Median:1.000	Median:9.000	Median:3583	Grand Hyatt Washington:	484	GR Washington:	484	Median:2.000	Orlando:667	Canada:250		
Mean:9.252	Mean:1.498	Mean:8.637	Mean:5067	Hyatt Regency Washington on Capitol Hill:	425	HR Washington:	425	Mean:2.227	Rosemont:579	India:226		
3rd Qu.:10.000	3rd Qu.:2.000	3rd Qu.:10.000	3rd Qu.:3694	Hyatt Regency Grand Cypress:	414	HR Grand Cypress:	414	3rd Qu.:3.000	Arlington:473	Hong Kong:215		
Max.:10.000	Max.:4.000	Max.:10.000	Max.:39001	Hyatt Regency Santa Clara:	351	HR Santa Clara:	351	Max.:9.000	Atlanta:440	China:200		
NA's:7112	NA's:9640	NA's:9640	(Other):711	(Other):16510	(Other):16510	(Other):15428	(Other):711					
Ops.Region_PL	Property.Latitude_PL	Property.Longitude_PL	Currency_PL	Dom.Int.1_PL	Brand_PL	Club.Type_PL	Region_PL	Category_PL	Type_PL			
Americas:17859	Min.: -6.909	Min.: -123.12	USD:17467	Domestic:17467	Andaz:118	:14818	Americas:17792	Full Service:6490	Business:2207			
ASPAC:1035	1st Qu.:32.751	1st Qu.: -96.77	KRW:265	International:1875	Grand Hyatt:929	Club:94	Asia Pacific:1035	Select Service:12852	Convention:2512			
EMEA:330	Median:35.829	Median: -84.36	CAD:250		Hyatt:743	Grand Club:681	Europe:348		Franchise:753			
Europe:118	Mean:35.781	Mean: -75.49	INR:226		Hyatt House:2206	Regency Club:3749	Middle East & Africa:167		Resort:1018			
	3rd Qu.:39.940	3rd Qu.: -77.56	HKD:215		Hyatt Place:10646				Select Service:12852			
	Max.:51.517	Max.:139.69	CNY:208		Hyatt Regency:4099							
			(Other):711		Park Hyatt:601							
Class_PL	Location_PL	Relationship_PL	GP_Tier	Booking_Channel	NPS_Type							
Luxury Class:1648	Airport:3963	Franchised:8514	Gold:7157	Electronic Distribution:11166	Detractor:2169							
Upper Upscale Class:4842	Resort:1115	Managed:6555	:5955	Global Contact Center:5457	Passive:3603							
Upscale Class:12852	Suburban:10537	Owned:4273	GOLD:3061	Hotel:2719	Promoter:13570							
	Urban:3727		Platinum:2084									
			Diamond:759									
			PLAT:244									
			(Other):82									

After →

Summary(ProjectSurveyUSA)														
MarketGroup	RTC	CheckOut	WalkIn	CheckInDate	CheckOut	LOS	NumRooms	Adult	Children	PurposeOfVisit	QuotedRate	ResvDate	ResyTime	
CONVENTION:3720	KING	17540	N:17426	Min. :2014-02-01	Min. :2014-02-01	Min. :1.000	Min. :1	Min. :1.000	Min. :1.000	BUSINESS:14006	Min. :0.0	Min. :2012-08-21	11:21:42:	5
DISCOUNT :4353	DOOL	13932	Y: 13	1st Qu.:2014-02-09	1st Qu.:2014-02-11	1st Qu.:1.000	1st Qu.:1	1st Qu.:1.000	1st Qu.:1.649	LEISURE :3353	1st Qu.:89.0	1st Qu.:2014-01-15	14:47:25:	5
OTHER :901	180N	1084		Median :2014-02-15	Median :2014-02-17	Median :2.000	Median :1	Median :1.000	Median :1.649		Median :113.0	Median :2014-02-01	10:18:50:	4
RACK :4514	QWQ	1855		Mean :2014-02-15	Mean :2014-02-17	Mean :2.366	Mean :1	Mean :1.423	Mean :1.650		Mean :125.2	Mean :2014-01-22	10:19:00:	4
SPECIAL :1322	DLXN	1297		3rd Qu.:2014-02-22	3rd Qu.:2014-02-24	3rd Qu.:3.000	3rd Qu.:1	3rd Qu.:2.000	3rd Qu.:1.649		3rd Qu.:152.2	3rd Qu.:2014-02-11	10:25:50:	4
VOLUME :2389	QWEN	1267		Max. :2014-02-28	Max. :2014-05-14	Max. :197.000	Max. :1	Max. :8.000	Max. :4.000		Max. :1909.0	Max. :2014-02-28	11:04:28:	4
WHOLESALE :240	(Other):3564													(Other):17413
EntryHotelCode	ResvDateChange	RTC_Book	State	GuestCountryOrigin	FutureResvDaysCat	FutureResvDays	Revenue	GuestCountryHome	Gender	AgeRange				
QNGOS :10743	Min. :2013-03-05	KING	18215	UNITED STATES :15099	0-3 Days :4660	Min. :1.0	Min. :0.0	USA :16099		196	46-55	15295		
OWARO :3346	1st Qu.:2014-01-21	DOOL	18024	TX :1815	31-60 Days:2477	1st Qu.:26.0	1st Qu.:109.7	Canada :198	Female	17950	36-45	13826		
HARRO :1053	Median :2014-02-04	180N	10508	CA :1157	8-14 Days :2430	Median :143.0	Median :190.0		79	Male	18968	56-65	13763	
OKTRO :52	Mean :2014-01-27	QWQ	1853	FL :1033	4-7 Days :2367	Mean :132.6	Mean :311.7	United Kingdom: 34	Prefer not to answer:	325	26-35	12304		
ORLZA :49	3rd Qu.:2014-02-13	QWEN	1427	IL :911	15-21 Days:1917	3rd Qu.:1221.0	3rd Qu.:368.0	Germany :28			66-75	11190		
FLIZA :42	Max. :2014-03-19	DLXN	1357	MI :657	22-30 Days:1861	Max. :284.0	Max. :18415.0	Mexico :25			18-25	501		
(Other):2154				(Other):2713	(Other):9811	(Other):241	(Other):1727	(Other):176			(Other):560			
Language	LR_Score	OS_Score	GR_Score	T_Score	HC_Score	CS_Score	SC_Score	IS_Score	CI_Score	FB_Frequency	FB_Score			
English :17249	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :0.00			
Spanish (Americas): 53	1st Qu.:8.000	1st Qu.:8.00	1st Qu.:8.00	1st Qu.:8.766	1st Qu.:8.000	1st Qu.:9.000	1st Qu.:9.000	1st Qu.:8.747	1st Qu.:9.281	1st Qu.:1.000	1st Qu.:8.63			
German :47	Median :9.000	Median :9.00	Median :9.00	Median :8.766	Median :9.000	Median :10.000	Median :9.035	Median :8.747	Median :9.281	Median :1.475	Median :8.63			
French :15	Mean :8.688	Mean :8.66	Mean :8.79	Mean :8.766	Mean :8.936	Mean :9.112	Mean :9.035	Mean :8.747	Mean :9.281	Mean :1.475	Mean :8.63			
Japanese ja_3P :14	3rd Qu.:10.000	3rd Qu.:10.00	3rd Qu.:10.00	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:10.000	3rd Qu.:9.000	3rd Qu.:10.000	3rd Qu.:1.475	3rd Qu.:9.00			
Korean :12	Max. :10.000	Max. :10.00	Max. :10.00	Max. :10.000	Max. :10.000	Max. :10.000	Max. :10.000	Max. :10.000	Max. :10.000	Max. :4.000	Max. :10.00			
(Other):49														
PropertyID	HotelName	HotelAbbrev	AwardCat	HotelCity	HotelCountry	OperatingRegion	Latitude	Longitude	Currency					
Min. :3010	Hyatt Regency Jacksonville Riverfront	:579	HR Jacksonville	:579	Min. :1.000	Washington :1006	United States:17439	Americas:17439	Min. :25.78	Min. : -122.27	USD :17439			
1st Qu.:3498	Hyatt Regency O'Hare	:579	HR O'Hare	:579	1st Qu.:1.000	Jacksonville: 667	Canada :0	ASPAC :0	1st Qu.:32.88	1st Qu.: -96.82	CAD :0			
Median :3577	Grand Hyatt Washington	:483	GH Washington	:483	Median :2.000	Orlando :667	China :0	EAME :0	Median :36.05	Median : -84.45	CHF :0			
Mean :4604	Hyatt Regency Washington on Capitol Hill	:424	HR Washington	:424	Mean :2.038	Rosemont :579	Egypt :0	Europe :0	Mean :35.97	Mean : -88.67	CNY :0			
3rd Qu.:3674	Hyatt Regency Grand Cypress	:414	HR Grand Cypress	:414	3rd Qu.:2.000	Arlington :473	Germany :0		3rd Qu.:35.92	3rd Qu.: -80.16	EGP :0			
Max. :134001	Hyatt Regency Santa Clara	:351	HR Santa Clara	:351	Max. :18.000	Atlanta :440	Hong Kong :0		Max. :44.86	Max. : -71.11	EUR :0			
(Other):						(Other):13527	(Other):0				(Other):0			
ValidValues	HotelBrand	Club	Region	Category	Type	Class	Location	Relationship						
Domestic :17439	Andaz :0	:13989	Americas :17439	Full Service :4611	Business :1005	Luxury Class :910	Airport :3958	Franchised:8494						
International: 0	Grand Hyatt :731	Club :0	Asia Pacific :0	Select Service:12828	Convention :2162	Upper Upscale Class: 3701	Resort :976	Managed :5101						
	Hyatt :507	Grand Club :483	Europe :0		Franchise :751	Upscale Class :12828	Suburban:9940	Owned :3844						
	Hyatt House :2198	Agency Club: 2967	Middle East & Africa: 0		Resort :693		Urban :2565							
	Hyatt Place :10630				Select Service:12828									
	Hyatt Regency :3194													
	Park Hyatt :179													
GPTier	Channel	NPS	CheckInDay	CheckOutDay	Coords	stateName	USRegion	Promoter	Detractor	Passive				
Gold :6741	Electronic Distribution:10211	Detractor: 2006	Length:17439	Length:17439	Min. :25.78	Length:17439		Min. :0.000	Min. :0.000	Min. :0.000				
:5441	Global Contact Center :5073	Passive :3208	Class :character	Class :character	1st Qu.:32.88	Class :character	Class :character	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.000				
GOLD :2517	Hotel :2155	Promoter :12225	Mode :character	Mode :character	Median :36.05	Mode :character	Mode :character	Median :1.000	Median :0.000	Median :0.000				
Platinum:1888					Mean :35.97			Mean :0.701	Mean :0.115	Mean :0.184				
Diamond :603					3rd Qu.:39.92			3rd Qu.:1.000	3rd Qu.:0.000	3rd Qu.:0.000				
PLAT :196					Max. :44.86			Max. :1.000	Max. :1.000	Max. :1.000				
(Other):53														

Data Quality Checks

To double check data integrity for NAs, we used the below command to find which columns had NA's: names(which(sapply(ProjectSurveyUSA, anyNA)))

We then created a text editor to validate the definition of variable while working to clean and check the quality of the data with :

- #Creating a Data Frame Using the R Data Editor
- Glossary<- data.frame(Column = character(), Defintion = character(), NewName = character())
- Glossary<- edit(Glossary)
- View(Glossary)

Column	Definition	NewName
1 MARKET_GROUP_C	Sub-group of major markets	MarketGroup
2 ROOM_TYPE_CODE_C	Hyatt standard room type code of the guest's room upo...	RTC_CheckOut
3 WALK_IN_FLG_C	Flag indicating a walk-in	Walkin
4 CHECK_IN_DATE_C	Check in date; for WALK status adjusted to the first in-h...	CheckInDate
5 CHECK_OUT_DATE_C	Check out date	CheckOut
6 LENGTH_OF_STAY_C	Length of stay	LOS
7 NUMBER_OF_ROOMS_C	Number of rooms occupied. Valid values are 0 or 1; for c...	NumRooms
8 ADULT_NUM_C	Number of adults on the last day of the stay	Adult
9 CHILDREN_NUM_C	Check in length	Children
10 POV_CODE_C	Purpose of visit	POVCode
11 QUOTED_RATE_C	Quoted rate derived from Reservation table	QuotedRate
12 RESERVATION_DATE_R	Date when the booking was made	ResvDate
13 ENTRY_TIME_R	Time the reservation was made	ResvTime
14 ENTRY_HOTEL_CODE_R	Entry hotel code	EntryHotelCode
15 LAST_CHANGE_DATE_R	Date of last change to the reservation	ResvDateChange
16 ROOM_TYPE_CODE_R	Hyatt standard room type code of the guest's room as p...	RTC_Book
17 STATE_R	State of the party making the reservation	State
18 GUEST_COUNTRY_R	The country to which the actual guest belongs to. Differ...	GuestCountryOrigin
19 PACE_CATEGORY_R	Categorizes the number of nights in the future the reser...	FutureResvDaysCat
20 PACE_R	Pace identifies the number of nights in the future the re...	FutureResvDays
21 REVENUE_USD_R	Total revenue from the reservation in USD	Revenue

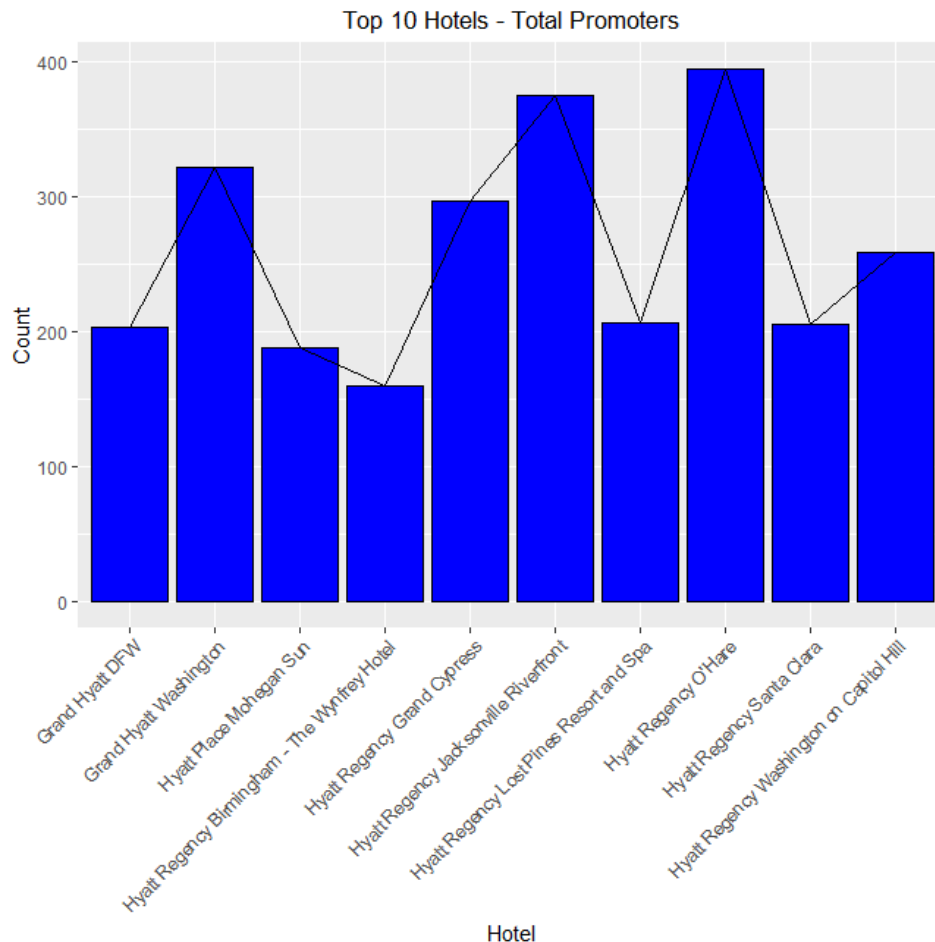
Descriptive Statistics

After the data had been cleansed and checked for quality, we were able to gather some initial descriptive statistics on the data set. Before diving into specifics around NPS, we wanted to look at the ten hotels with the:

- most promoters
- fewest promoters
- most detractors/passives
- highest percentage of promoters among survey takers
- lowest percentage of promoters among survey takers

Ten hotels with the most promoters:

	HotelName	Promoter	n
1	Hyatt Regency O'Hare	1	394
2	Hyatt Regency Jacksonville Riverfront	1	375
3	Grand Hyatt Washington	1	321
4	Hyatt Regency Grand Cypress	1	296
5	Hyatt Regency Washington on Capitol Hill	1	258
6	Hyatt Regency Lost Pines Resort and Spa	1	206
7	Hyatt Regency Santa Clara	1	205
8	Grand Hyatt DFW	1	203
9	Hyatt Place Mohegan Sun	1	188
10	Hyatt Regency Birmingham - The Wynfrey Hotel	1	160

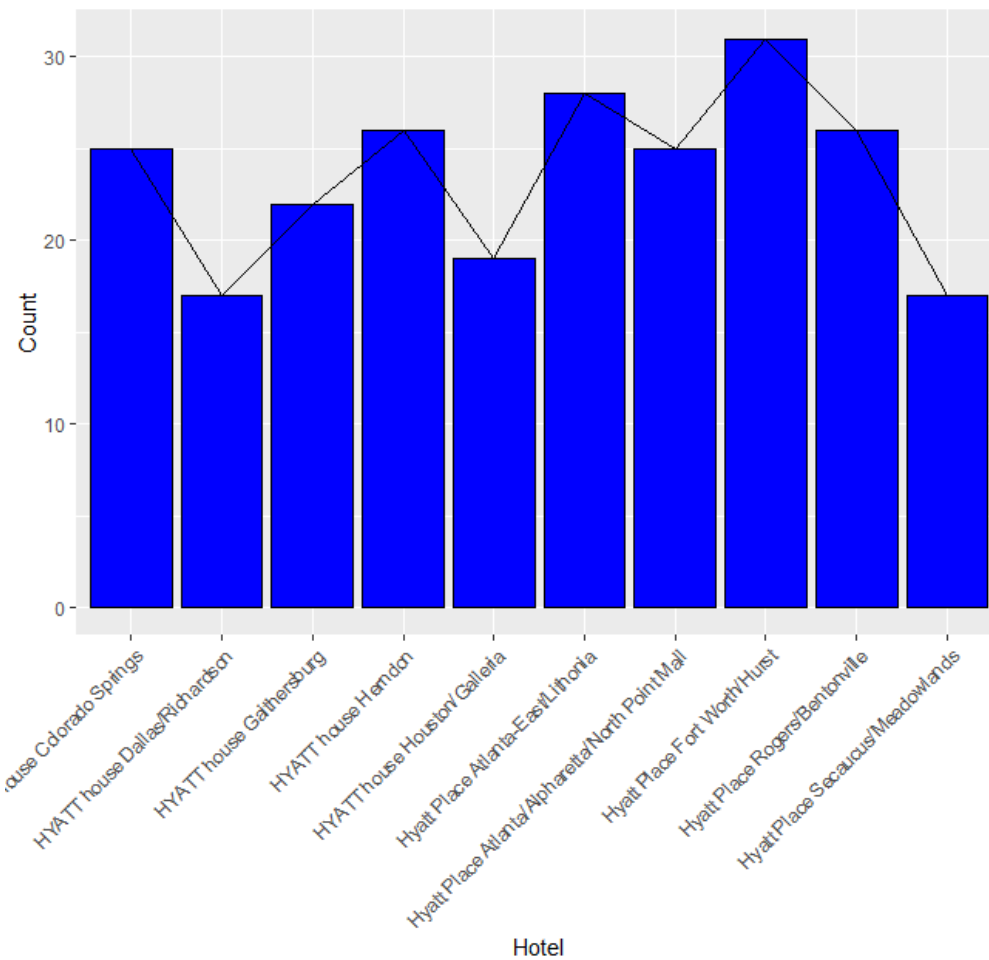


At this point, it is worthwhile to remember that we are looking at a group of 17,000+ surveys across 170 hotels. With this in mind, we see that each of the top ten hotels have a significant sample size (all above 160). The Hyatt Regency O'Hare led the way with 394 surveys resulting in promoters.

Ten hotels with the fewest promoters:

	HotelName	Promoter	n
161	Hyatt Place Fort Worth/Hurst	1	31
162	Hyatt Place Atlanta-East/Lithonia	1	28
163	HYATT house Herndon	1	26
164	Hyatt Place Rogers/Bentonville	1	26
165	HYATT house Colorado Springs	1	25
166	Hyatt Place Atlanta/Alpharetta/North Point Mall	1	25
167	HYATT house Gaithersburg	1	22
168	HYATT house Houston/Galleria	1	19
169	HYATT house Dallas/Richardson	1	17
170	Hyatt Place Secaucus/Meadowlands	1	17

Bottom 10 Hotels - Total Promoters

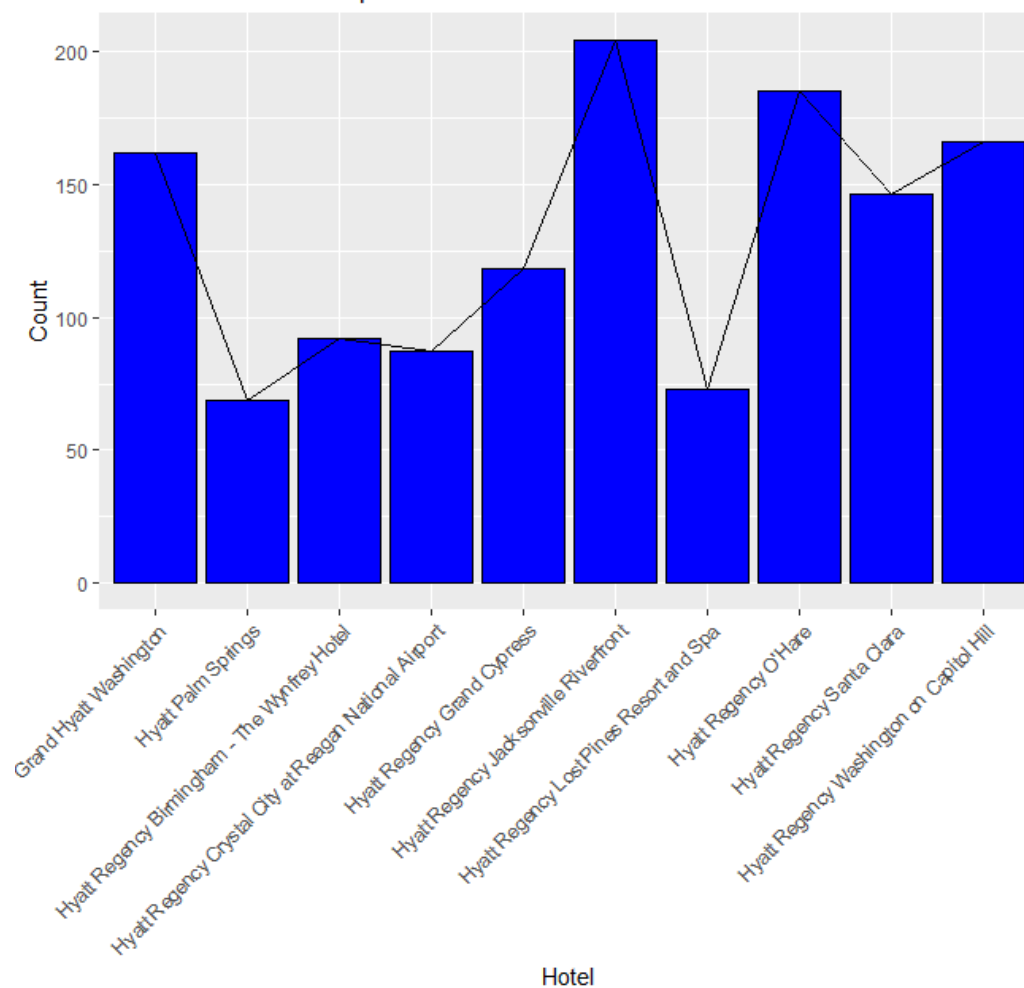


When we compare the surveys resulting in promoters from the hotels in spots 161-170 on this same list, we see that the sample sizes were significantly lacking. More than anything, this tells us that the hotels could do a better job of promoting and/or incentivizing survey completion among guests.

Ten hotels with the most detractors/passives:

	HotelName	Promoter	n
1	Hyatt Regency Jacksonville Riverfront	0	204
2	Hyatt Regency O'Hare	0	185
3	Hyatt Regency Washington on Capitol Hill	0	166
4	Grand Hyatt Washington	0	162
5	Hyatt Regency Santa Clara	0	146
6	Hyatt Regency Grand Cypress	0	118
7	Hyatt Regency Birmingham - The Wynfrey Hotel	0	92
8	Hyatt Regency Crystal City at Reagan National Airport	0	87
9	Hyatt Regency Lost Pines Resort and Spa	0	73
10	Hyatt Palm Springs	0	69

Top 10 Hotels - Total Detractors/Passives

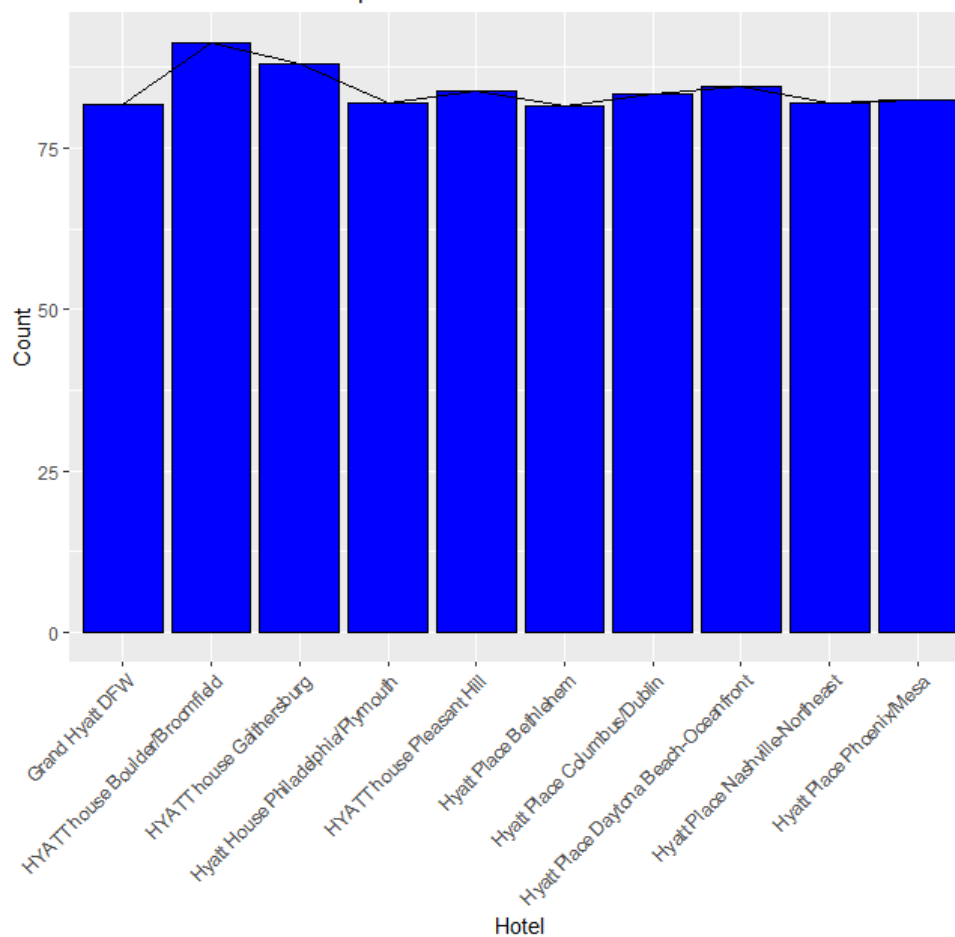


As stated above, looking at the total number of promoters (or non-promoters in this case) is not the most useful metric, simply because the sample sizes vary so greatly between hotels. The hotels above have the most non-promoters, but this may be just a function of their surveys being completed by a wider audience.

Ten hotels with the highest percentage of promoters among survey takers:

	HotelName	Promoter	n	p
93	HYATT house Boulder/Broomfield	1	53	91.37931
167	HYATT house Gaithersburg	1	22	88.00000
17	Hyatt Place Daytona Beach-Oceanfront	1	109	84.49612
123	HYATT house Pleasant Hill	1	47	83.92857
53	Hyatt Place Columbus/Dublin	1	70	83.33333
31	Hyatt Place Phoenix/Mesa	1	90	82.56881
86	Hyatt House Philadelphia/Plymouth	1	55	82.08955
88	Hyatt Place Nashville-Northeast	1	55	82.08955
8	Grand Hyatt DFW	1	203	81.85484
94	Hyatt Place Bethlehem	1	53	81.53846

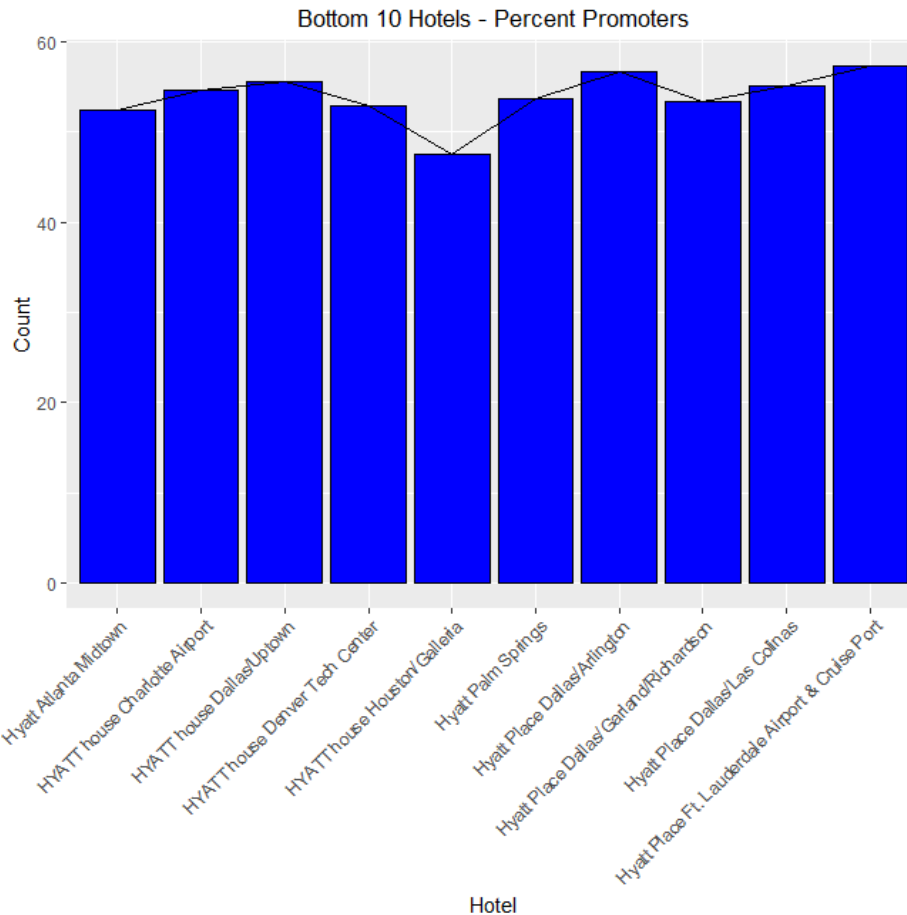
Top 10 Hotels - Percent Promoters



Here we find a more useful view of the top ten hotels: viewing them by “promoter percentage” - the percentage of surveys that resulted in promoters. While the data does include some hotels with a very small sample size (Hyatt House Gaithersburg is the most extreme example with 22), it gives us a better idea of the overall service that a hotel provides.

Ten hotels with the lowest percentage of promoters among survey takers:

	HotelName	Promoter	n	p
87	Hyatt Place Ft. Lauderdale Airport & Cruise Port	1	55	57.29167
146	Hyatt Place Dallas/Arlington	1	38	56.71642
112	HYATT house Dallas/Uptown	1	50	55.55556
147	Hyatt Place Dallas/Las Colinas	1	38	55.07246
136	HYATT house Charlotte Airport	1	41	54.66667
41	Hyatt Palm Springs	1	80	53.69128
139	Hyatt Place Dallas/Garland/Richardson	1	40	53.33333
148	HYATT house Denver Tech Center	1	36	52.94118
131	Hyatt Atlanta Midtown	1	43	52.43902
168	HYATT house Houston/Galleria	1	19	47.50000

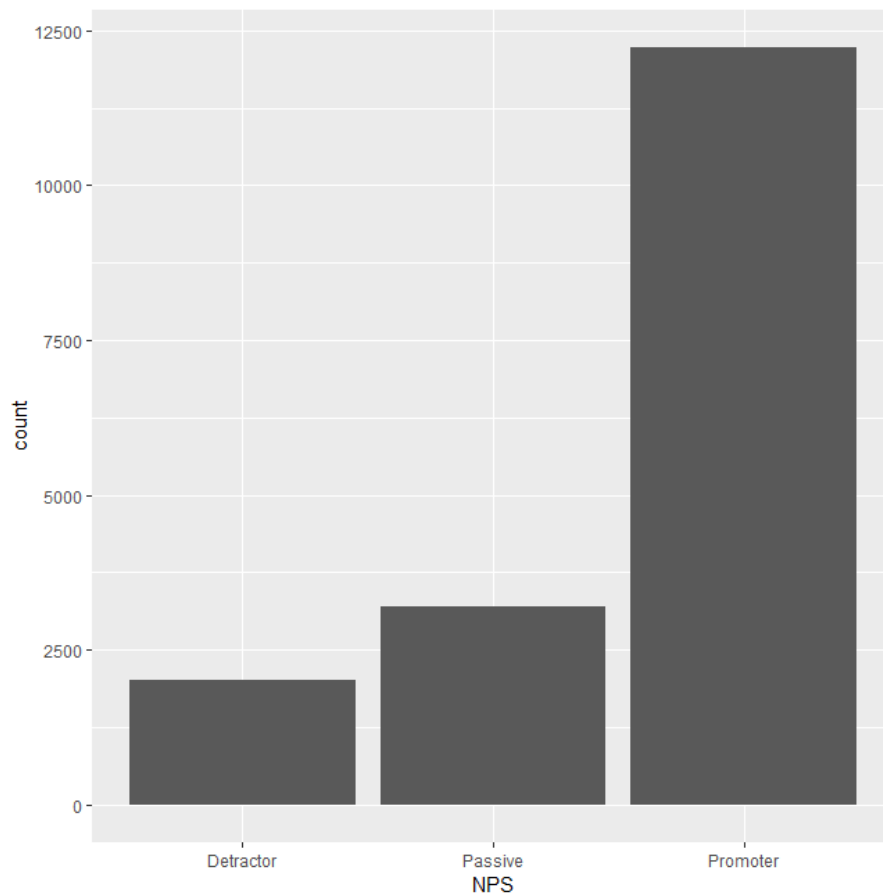


Lastly, we have a snapshot that looks at the lowest promoter rates among the 170 hotel data set. Again, sample size issues affect this group (in fact, the number of surveys here is significantly lower than those at the top). That said, we can still glean that there may be pervasive issues at some of these hotels that leave customers dissatisfied with their stays.

NPS type analysis

Since our aim is to actually discover which factors drive NPS, we thought it best to next dive into some analysis of the NPS types within this data set.

Looking at our 17,439 observations, we are able to see that far and away, Hyatt's guests are promoters. Once that had been established, we needed to drill down further to see if NPS type changed widely across our other variables.

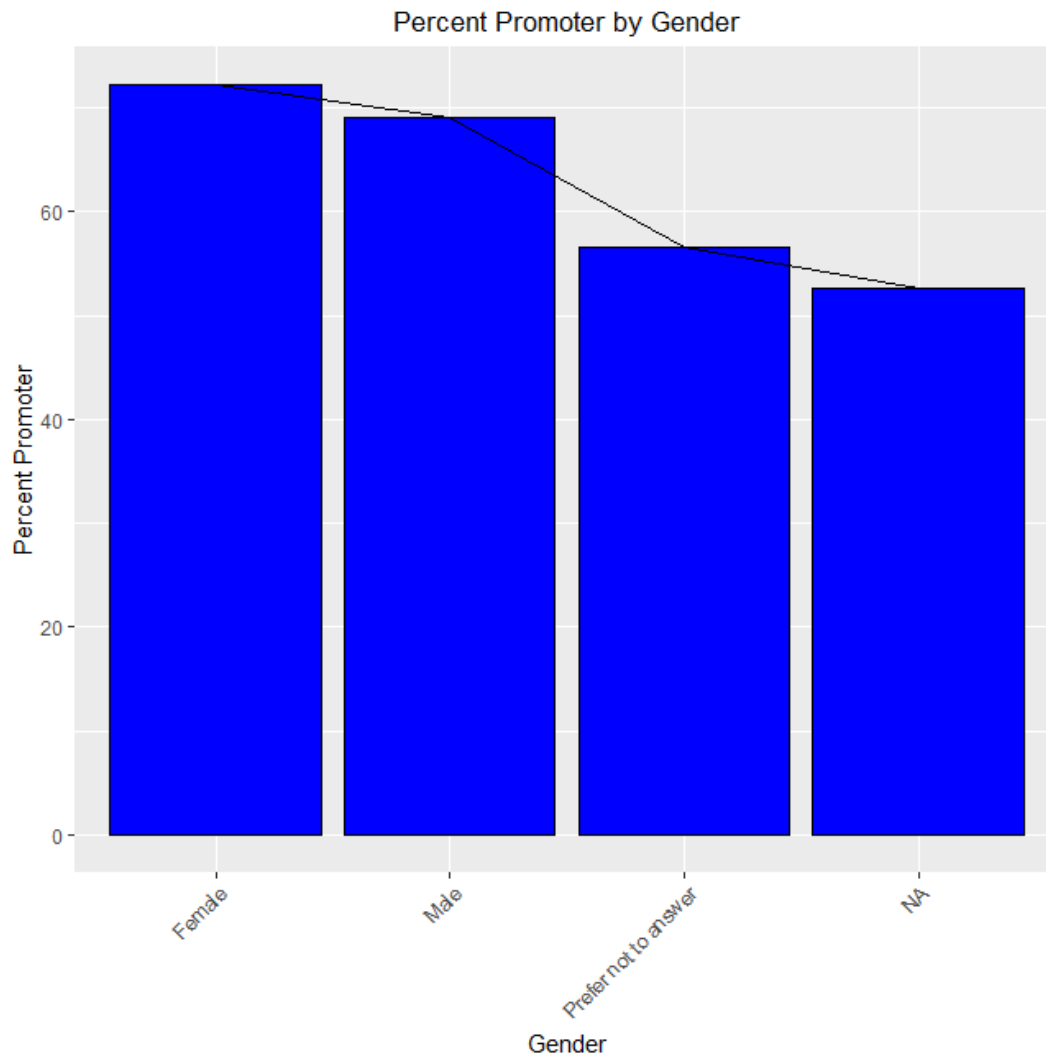


In addition to this high-level analysis of NPS types, we also analyzed how the following criteria affected NPS type:

- a. **Gender**
- b. **Age**
- c. **Tier status**
- d. **Purpose of visit**
- e. **Hotel state**
- f. **Hotel brand**
- g. **Hotel ownership type**

Gender breakdown of NPS promoters:

	Gender	Promoter	n	tn	p
2	Female		1 5740	7950	72.20126
1	Male		1 6198	8968	69.11240
3	Prefer not to answer		1 184	325	56.61538
4	<NA>		1 103	196	52.55102

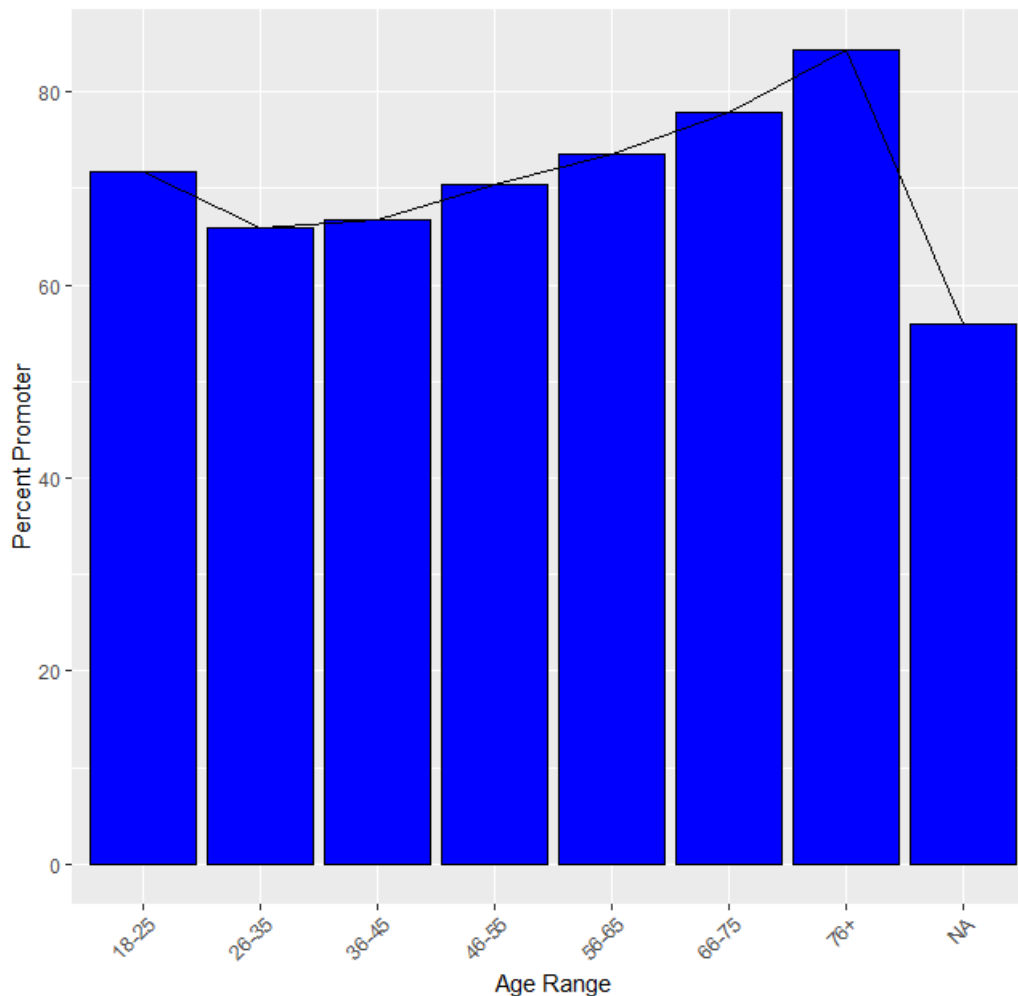


Here we see that the majority of men (6198 of 8968, or 69%), women (5740 of 7950, or 72%), and those that preferred not to answer (184 of 325, or 57%) were classified as promoters. One of the largest takeaways here (one that will be echoed throughout our analysis) is that Hyatt's data collection could be more effective. By making this question mandatory, Hyatt could have prevented the 196 <NA> values that appear in this chart and graph.

Age breakdown of NPS promoters:

	AgeRange	Promoter	n	tn	p
8	76+	1	166	197	84.26396
5	66-75	1	927	1190	77.89916
2	56-65	1	2769	3763	73.58491
6	18-25	1	359	501	71.65669
1	46-55	1	3731	5295	70.46270
3	36-45	1	2552	3826	66.70152
4	26-35	1	1518	2304	65.88542
7	<NA>	1	203	363	55.92287

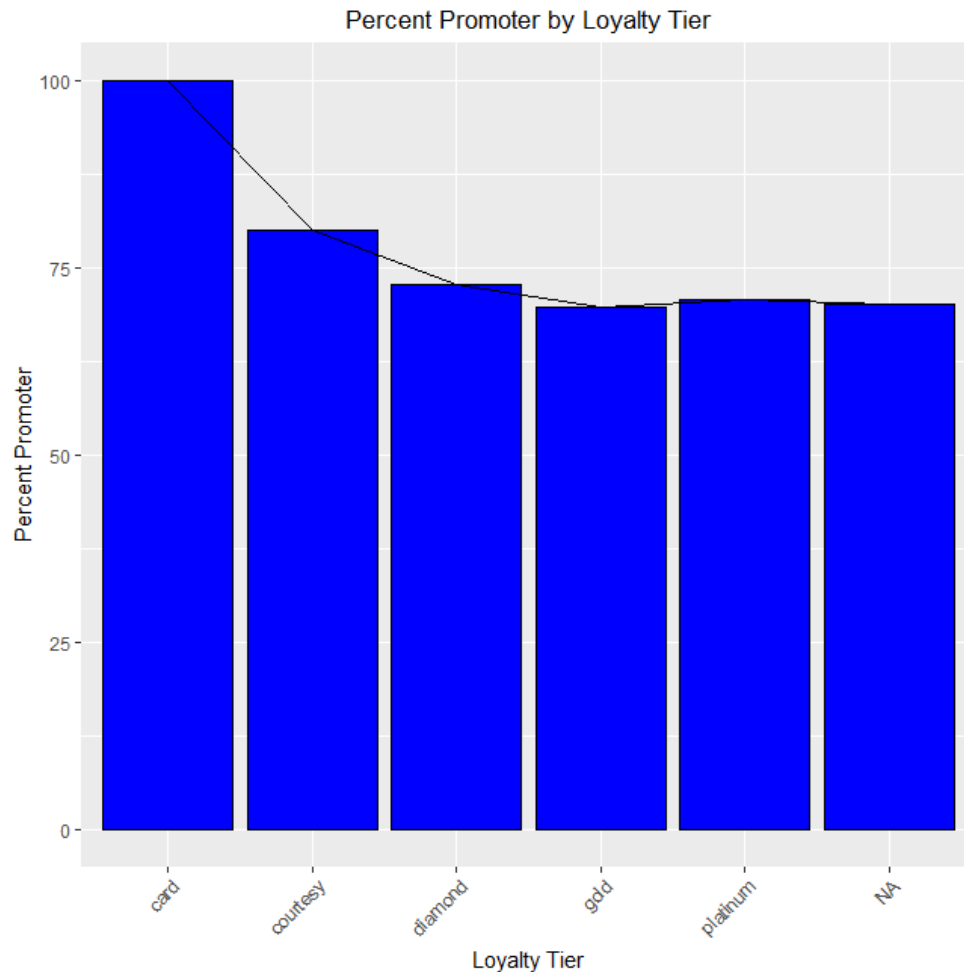
Percent Promoter by Age Range



This shows us that the age group with the highest rate of promoters is the 76+ group (84%), and that the lowest defined group was the 26-35 group (66%). The group with the most promoters overall was the 46-55 block, and, despite having the highest promoter rate, the 76+ bracket had the smallest number of promoters. If the 18-25 age group is dismissed as an anomaly produced by relatively small sample size, we can see that there is a correlation between guests' age and their overall satisfaction with their stay.

Tier Status breakdown of NPS promoters:

	GPTier	Promoter	n	tn	p
6	card	1	1	1	100.00000
5	courtesy	1	8	10	80.00000
4	diamond	1	469	645	72.71318
3	platinum	1	1477	2084	70.87332
2	<NA>	1	3818	5441	70.17092
1	gold	1	6452	9258	69.69108

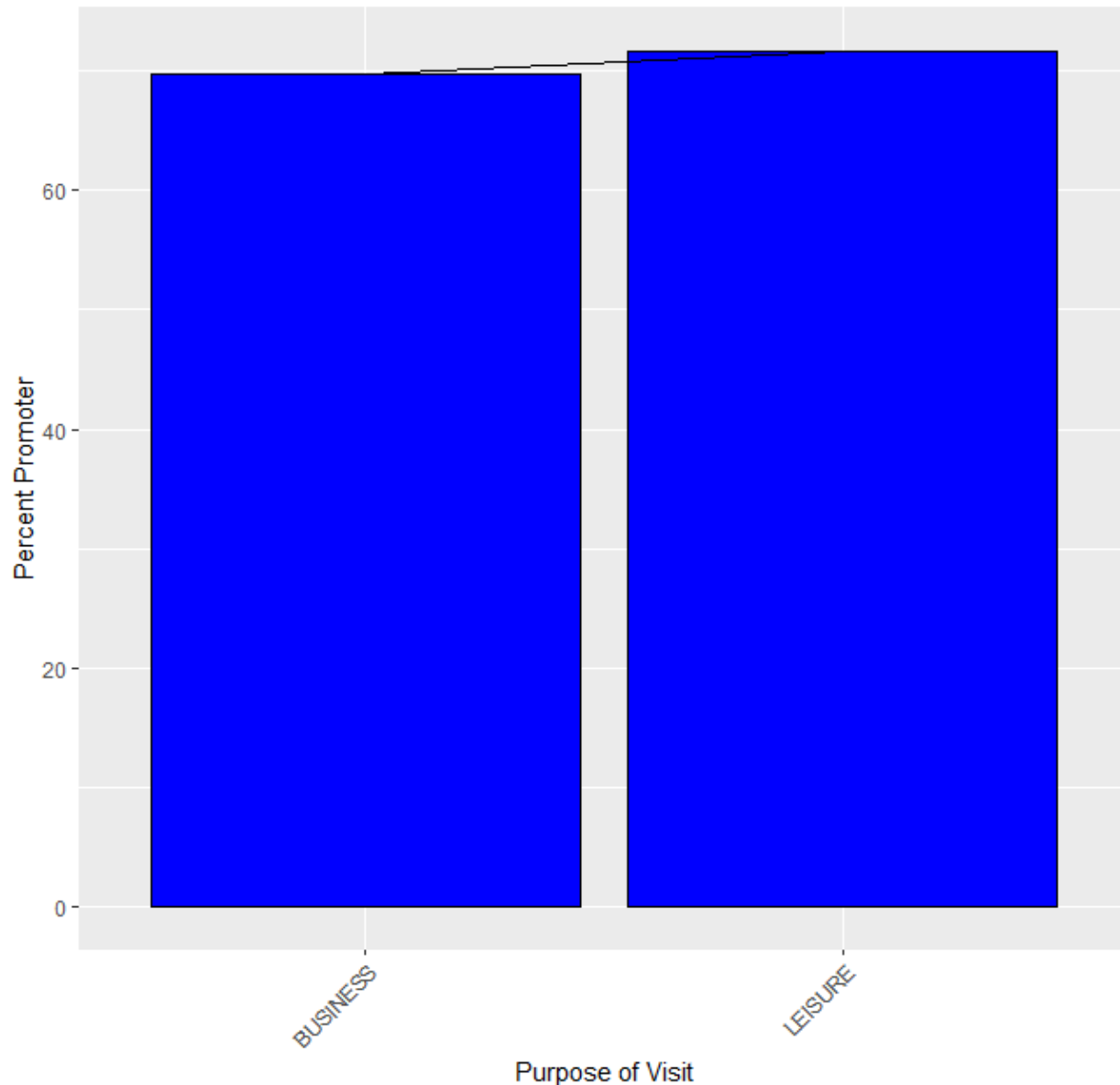


By conducting analysis on Hyatt's loyalty program (as it relates to NPS), we see that NPS types across the three status tiers represented are relatively similar. Additionally, the promoter rate among survey-takers who did not participate in the loyalty program was similar as well. That said, Hyatt is missing opportunities to enroll guests in its loyalty program. We see that the higher a guest's status, the more satisfied they are with their experience; why not enroll as many guests as possible? Additionally, if "card" represents a guest who holds a Hyatt-branded credit card, Hyatt can do much better in its promotion of the card, given that there is only one observation in the entire dataset.

Purpose of visit breakdown of NPS promoters:

Purpose of Visit	Promoter	n	tn	p
2	LEISURE	1 2404	3353	71.69699
1	BUSINESS	1 9821	14086	69.72171

Percent Promoter by Purpose of Visit

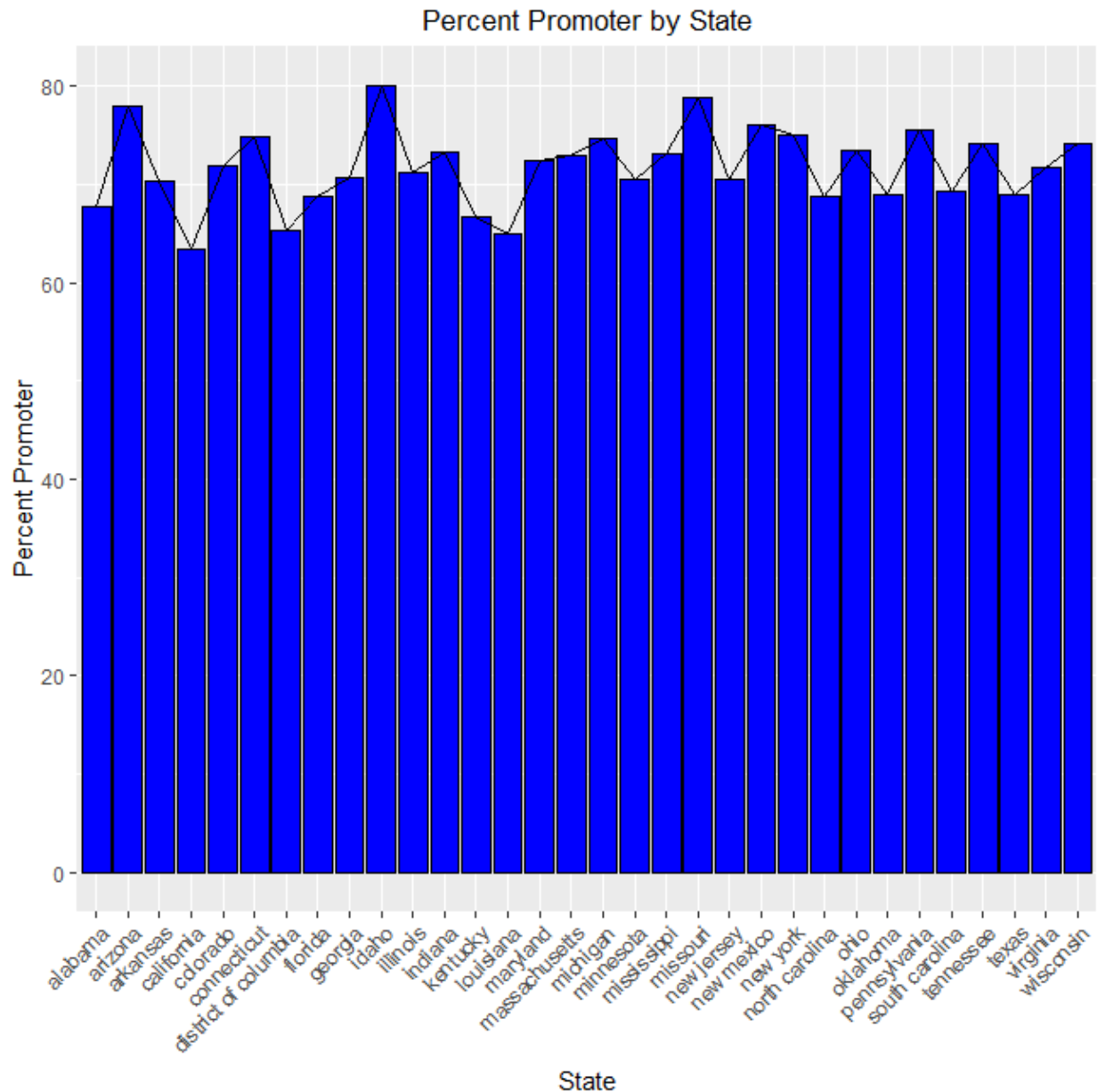


In looking at the number of promoters and rate of promoters among business and leisure travelers, we see that there is little difference between the two. Nonetheless, the business travelers promoted at a rate that was about two percentage points lower than leisure travelers. Notwithstanding, the sample size of the business travelers was more than four times that of the pool of leisure travelers.



Hotel state breakdown as it relates to NPS promoters:

	stateName	Promoter	n	tn	p
28	idaho	1	92	115	80.00000
29	missouri	1	78	99	78.78788
20	arizona	1	163	209	77.99043
17	new mexico	1	212	279	75.98566
18	pennsylvania	1	206	273	75.45788
25	new york	1	99	132	75.00000
13	connecticut	1	321	429	74.82517
7	michigan	1	505	677	74.59380
11	tennessee	1	390	526	74.14449
22	wisconsin	1	126	170	74.11765
12	ohio	1	380	517	73.50097
19	indiana	1	176	240	73.33333
26	mississippi	1	95	130	73.07692
21	massachusetts	1	146	200	73.00000
24	maryland	1	108	149	72.48322
8	colorado	1	429	597	71.85930
9	virginia	1	421	587	71.72061
4	illinois	1	896	1259	71.16759
5	georgia	1	750	1060	70.75472
14	new jersey	1	294	417	70.50360
27	minnesota	1	93	132	70.45455
32	arkansas	1	26	37	70.27027
16	south carolina	1	214	309	69.25566
23	oklahoma	1	118	171	69.00585
1	texas	1	1926	2795	68.90877
2	florida	1	1508	2191	68.82702
10	north carolina	1	392	570	68.77193
15	alabama	1	283	418	67.70335
30	kentucky	1	76	114	66.66667
3	district of columbia	1	975	1492	65.34853
31	louisiana	1	52	80	65.00000
6	california	1	675	1065	63.38028



After grouping the hotels into the states where they are located and honing in on their promoter rates, there appears to be a great deal of disparity between the top states and those further down the list. While Idaho's 80% promoter rate can be explained away by its relatively small sample size, other states with considerable sample sizes still performed much higher than others. For instance, Ohio, Tennessee, Michigan, and Connecticut all had over 400 survey takers yet still cleared the 73.5% mark for promoter rate. Meanwhile, California's promoter rate was around 63% over a sample size of 1,065.



Hotel brand breakdown for NPS promoters:

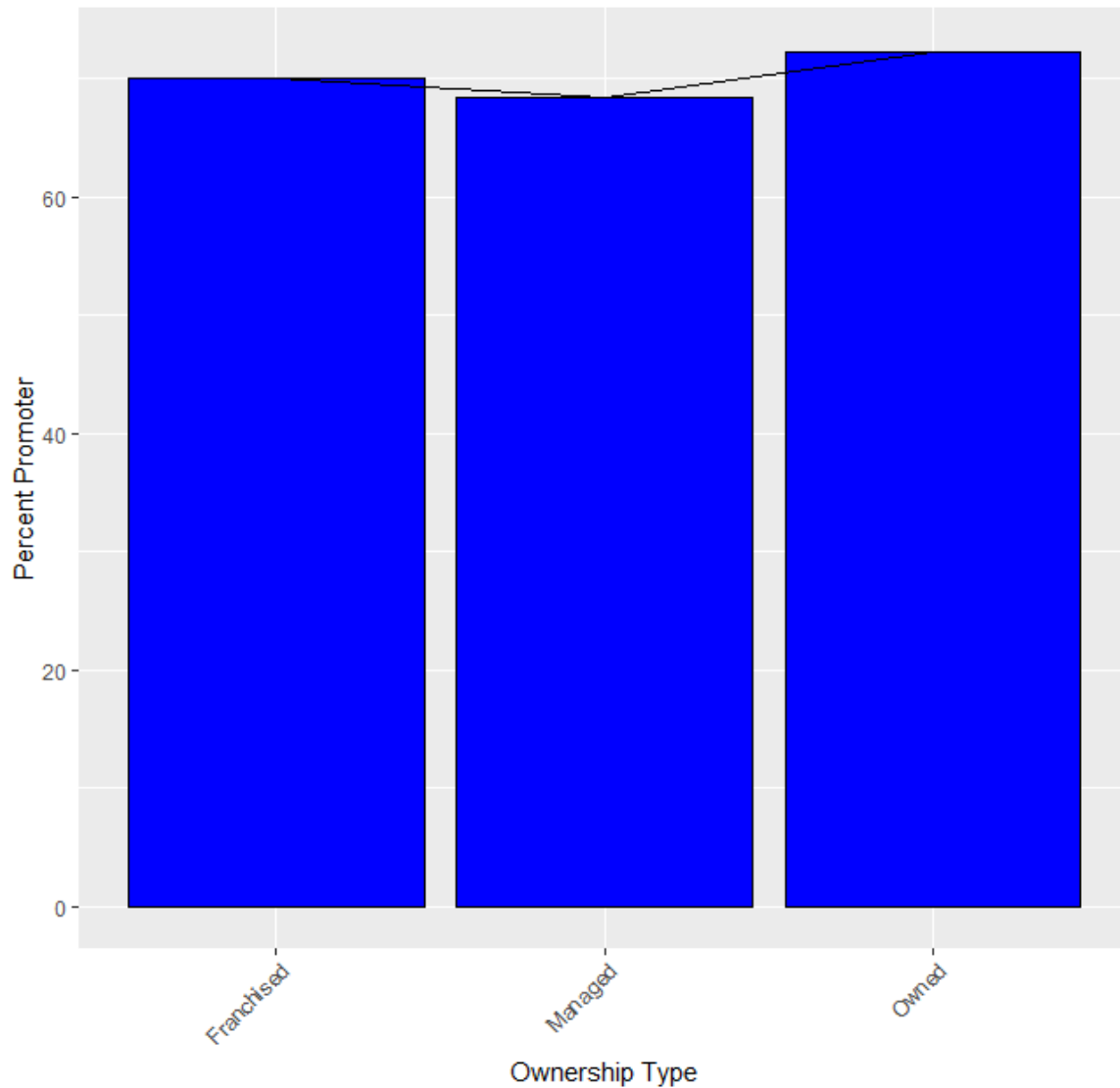
	HotelBrand	Promoter	n	tn	p
6	Park Hyatt	1	145	179	81.00559
4	Grand Hyatt	1	524	731	71.68263
1	Hyatt Place	1	7603	10630	71.52399
3	Hyatt House	1	1556	2198	70.79163
2	Hyatt Regency	1	2091	3194	65.46650
5	Hyatt	1	306	507	60.35503

Here we see that the two brands with the smallest sample sizes (Hyatt and Park Hyatt) fell well outside of the other brands' promoter rates. That said, the three brands with large sample sizes (Hyatt Regency, Hyatt House, and Hyatt Place) did experience some disparity, with 65%, 71%, and 72% promoter rates, respectively.

Hotel ownership type breakdown for NPS promoters:

	Relationship	Promoter	n	tn	p
3	Owned	1	2781	3844	72.34651
1	Franchised	1	5952	8494	70.07299
2	Managed	1	3492	5101	68.45717

Percent Promoter by Ownership Type

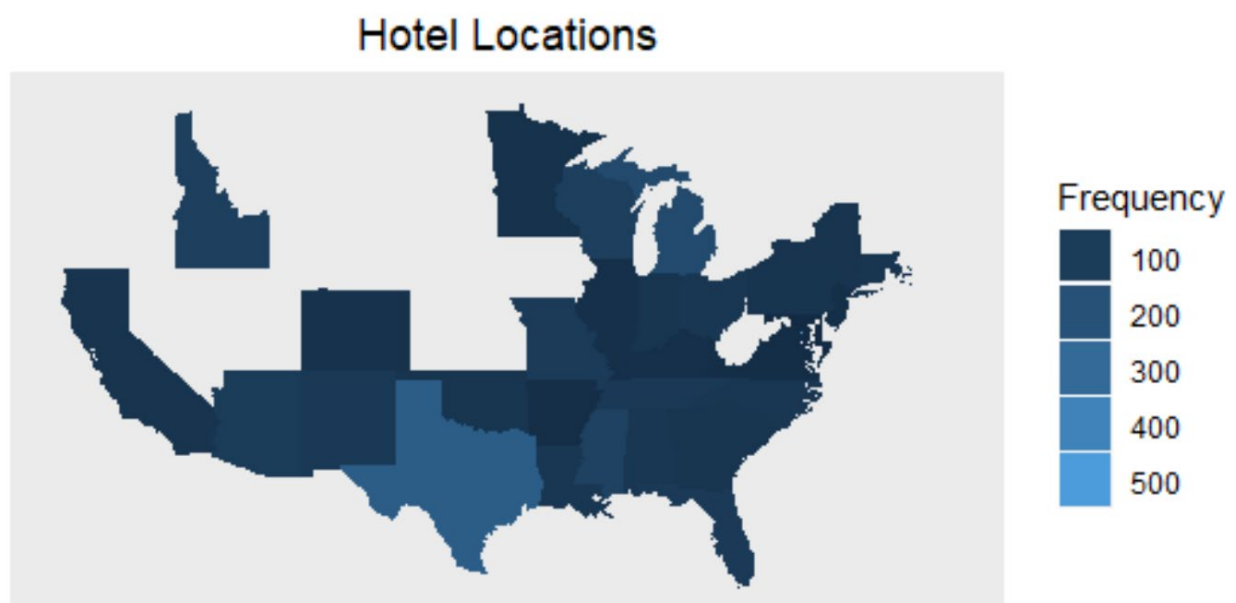


The above chart seems to suggest that the Hyatt brand's stewardship of hotels tends to be more effective than independent ownership. The hotels with the lowest scores tended to be those that were managed by third party companies.

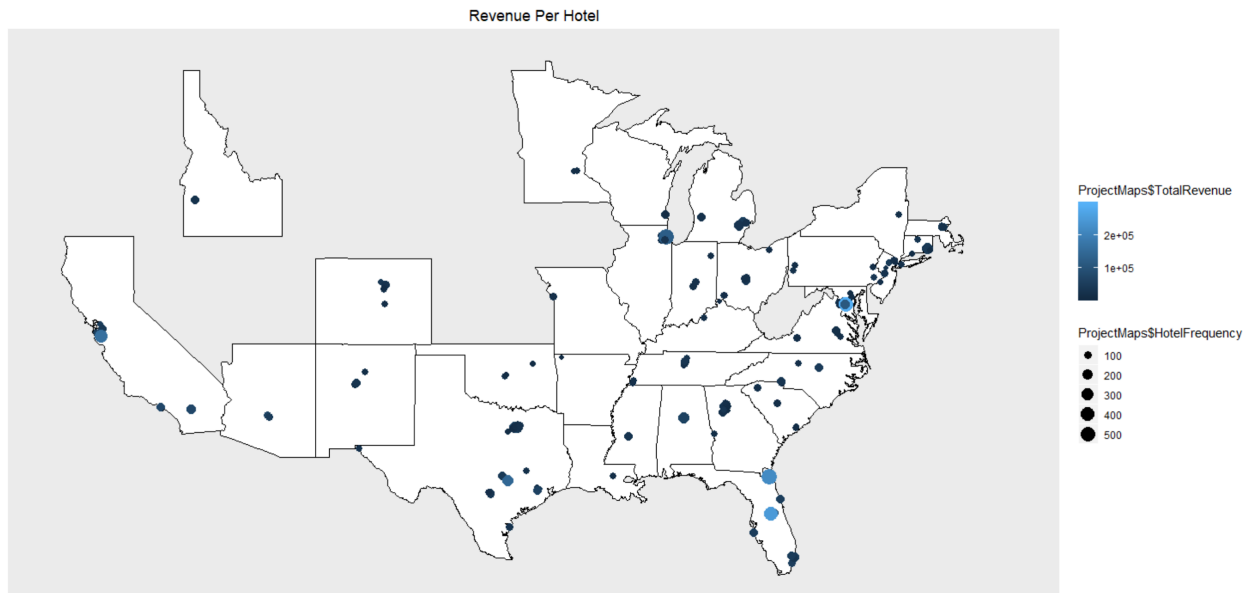
Maps

Orange plotted the 170 unique hotels to better understand the location, revenue and frequency of stay for each hotel.

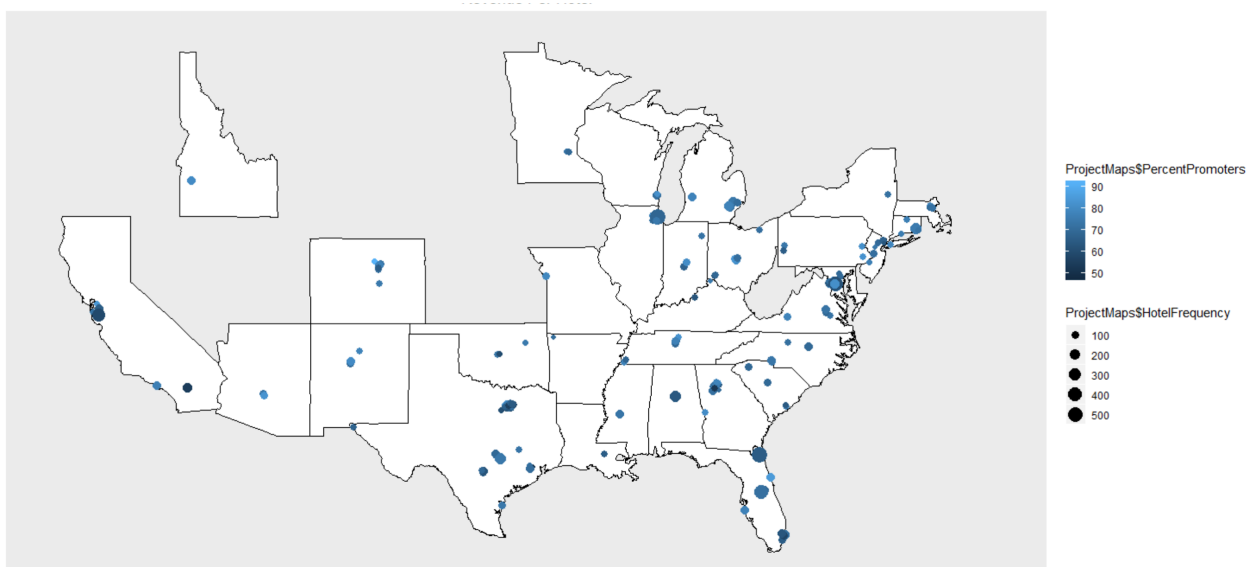
We are able to see that Texas has the most frequently visited hotels. No major patterns begin to emerge with this view in terms of frequency, but we are able to see a large swath of the country (in the Midwest and West) that is not represented in the data set. This information can be key for planning future hotels.



The image below allows us to understand the relationship between frequency and revenue generated for each hotel. The visual suggests what we expected: More visits to a hotel will result in greater revenue as seen by the bigger and brighter circles. The only exception is in California and Illinois. Due to the data being collected in February it makes sense to see Chicago not producing much revenue despite a relatively large circle if we make the assumption that the Chicago hotels were discounted during the colder months. The same assumption would likely not hold true for hotels in a warmer climate such as California.



The following map allows us to see the relationship between Promoter Percentage (Percentage of Promote vs. Non Promote) and the frequency at which guests visit the hotels. There appears to be a strong correlation here due to the larger circles being brighter. This may suggest that guests are providing a Promote ranking and returning as a repeat guest.





Top and Bottom 10 Profiles

Orange wanted to determine the differentiating factors between the Top 10 and Bottom 10 hotels. We ranked hotels based on the percentage of guests that submitted a Promoter survey, and leveraged numerical and factorial variables to gain insights.

The following code and function can be used as surveys are populated so that Hyatt can have live updates of the Top and Bottom 10 Hotels.

```
HotelPromotePercent <- function()
{
  hotelPromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
    group_by(HotelAbbrev = fct_inorder(HotelAbbrev)) %>%
    summarise(Percent = percent(mean(promoter)))
  return(hotelPromoter[order(desc(hotelPromoter$Percent)),])
}

HotelRankings <- HotelPromotePercent()

Top10 <- as.data.frame(HotelRankings[1:10,])
Top10 <- merge(Top10, ProjectSurveyUSA, by = "HotelAbbrev")

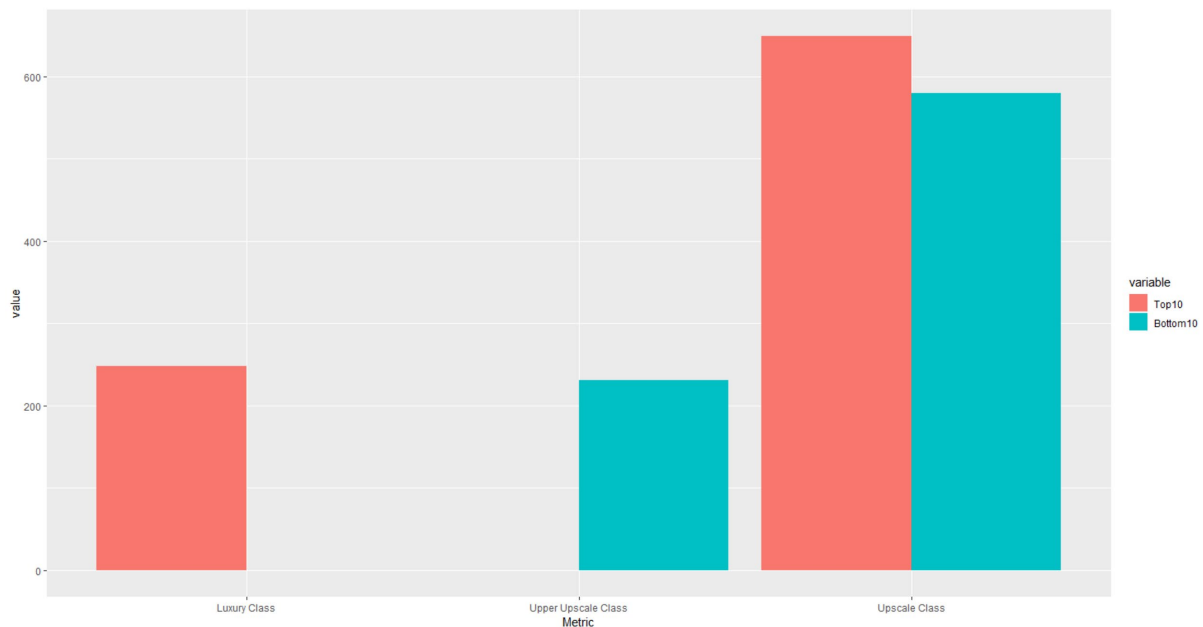
Bottom10 <- as.data.frame(HotelRankings[(dim(HotelRankings)[1]-9):dim(HotelRankings)[1],])
Bottom10 <- merge(Bottom10, ProjectSurveyUSA, by = "HotelAbbrev")
```

In order to compare the two categories, charts and visuals were generated.

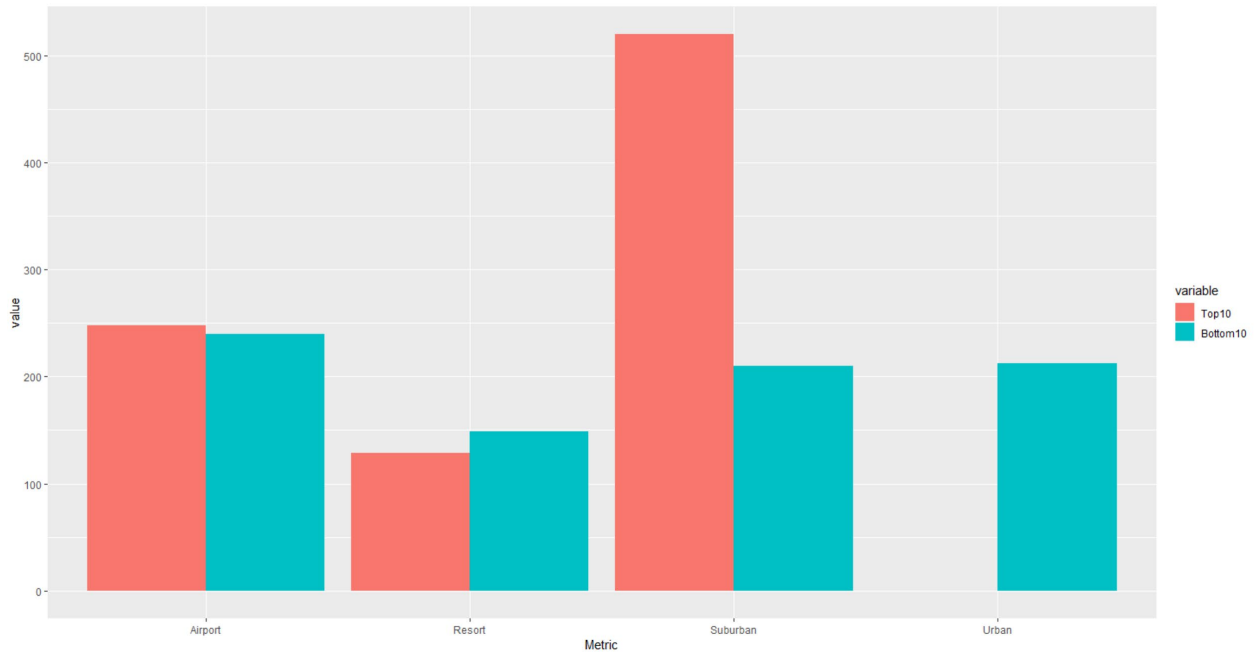
As a result of this comparison we are able to see that guest scores for the Top 10 hotels are higher in nearly every category. Perhaps the most interesting piece is that Quoted Rates and Revenue are higher with the Top 10 hotels. This suggests that guests are more concerned about the experience during their stay than they are with saving money. This insight is critical as Hyatt attempts to convert Non Promoters to Promoters.

	Metric	Top10	Bottom10
1	AwardCat	2.1917503	1.9038224
2	CI_Score	9.3717343	9.0000275
3	CS_Score	9.3740826	8.5392483
4	FBE_Score	8.8602026	8.4486094
5	FB_Frequency	1.4891112	1.4782775
6	FutureResvDays	124.7190635	133.8249075
7	GR_Score	9.2922595	8.1971542
8	HC_Score	9.4344979	8.1405404
9	IS_Score	8.8932721	8.7039651
10	LOS	2.7775648	2.6626004
11	LR_Score	9.2307692	7.8051788
12	OS_Score	9.1036789	7.8865312
13	Promoter	0.8338907	0.5425401
14	QuotedRate	145.6821962	131.6670530
15	Revenue	378.2024080	346.0555857
16	SC_Score	9.1474880	8.7163978
17	T_Score	9.0359993	8.3930850

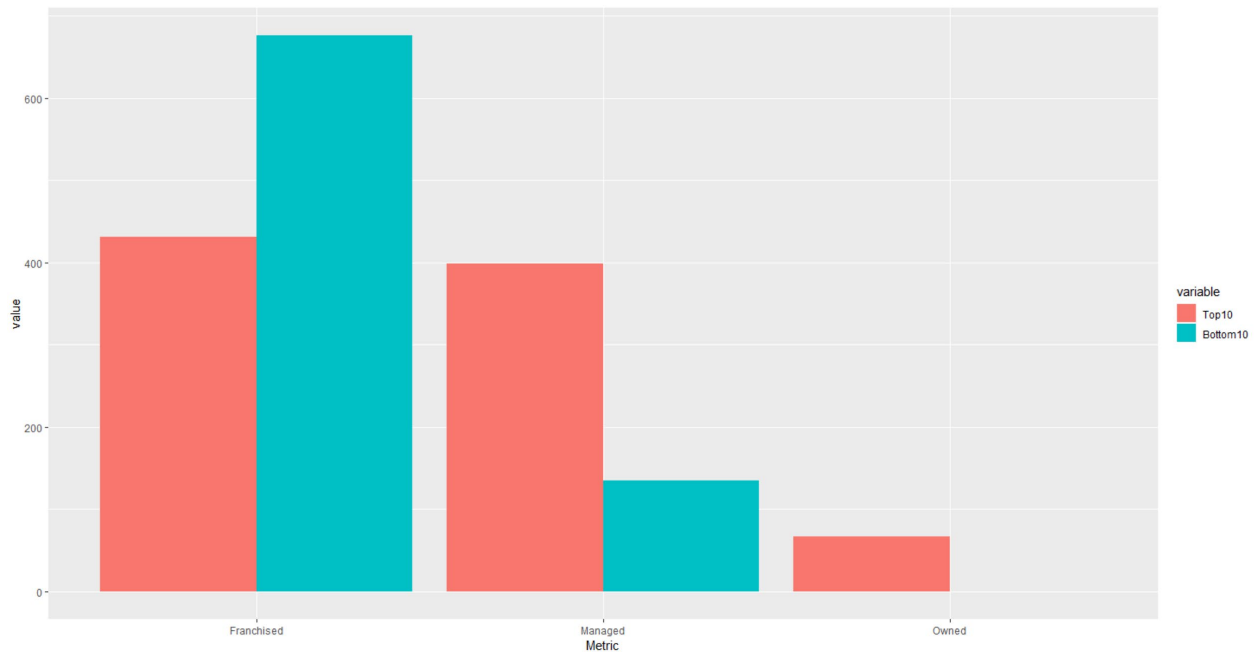
Our observations about spending seem to hold true given the image below. Luxury hotels are absent from the Bottom 10, but formed an important part of the Top 10. While the Bottom 10 does focus include Upper Upscale properties, the surveys resulting in promoters still fall short of the Top 10's Luxury totals. A shift in focus and marketing might better serve the Bottom 10.



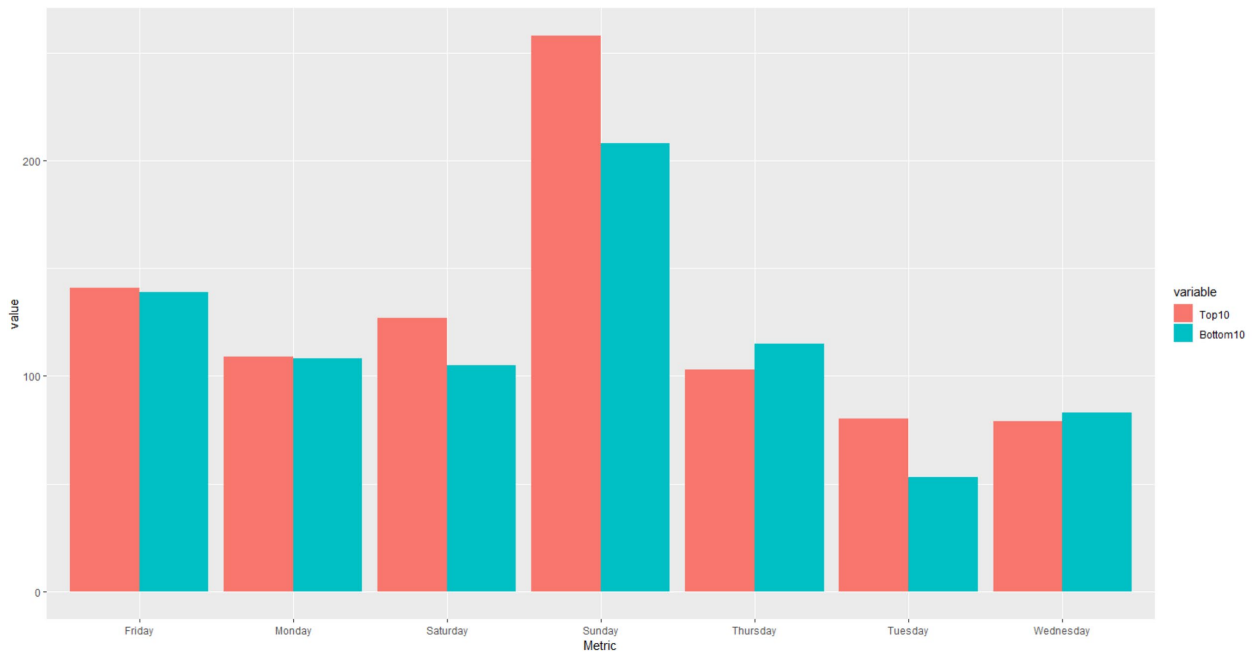
If Hyatt does consider expansion to the Midwest and West it may want to consider building hotels in the suburbs, as the Top 10 hotels capitalized on this location when compared to the Bottom 10. Of note: Urban locations did not contribute to the success of the Top 10 hotels.



Hyatt should also consider the types of owners operating the hotels as franchisees are connected more to the Bottom 10 while Hyatt-owned properties are associated exclusively with the Top 10.



Orange also recognized a trend on check out day for the Top and Bottom 10. The Top 10 hotels appeared to have more guests checkout on the weekends when compared to the Bottom 10. Perhaps the Bottom 10 hotels can offer promotions that incentivize guests to stay over for the weekend, which will hopefully result in more Promoter surveys for the Bottom 10.



Models

In order to determine the factors that are the most significant contributors to the NPS, we utilized our knowledge and understanding from the previous sections.

Linear Model

The following code was used to predict Detractor, Passive or Promoter based on a training data set and then tested against the test data set.

```
syModel <- polr(NPS ~ CS_Score + GR_Score + HC_Score + OS_Score + QuotedRate + Revenue + SC_Score + Relationship + CheckOutDay + T_Score + Type + Location, data = trainData, Hess = TRUE)
```

This resulted in a model with 89.97% accuracy as seen in the matrix below.

testData...59.	Detractor	Passive	Promoter
Detractor	511	109	20
Passive	60	768	199
Promoter	13	182	3951

```
Accuracy <- round((511+768+3951)/((511+60+13+109+768+182+20+199+3951)*100,2)
Accuracy
```

Association Rules Mining

The purpose is to find the rules which help the team predict the occurrence of specific characteristics based on the other variables in the data set

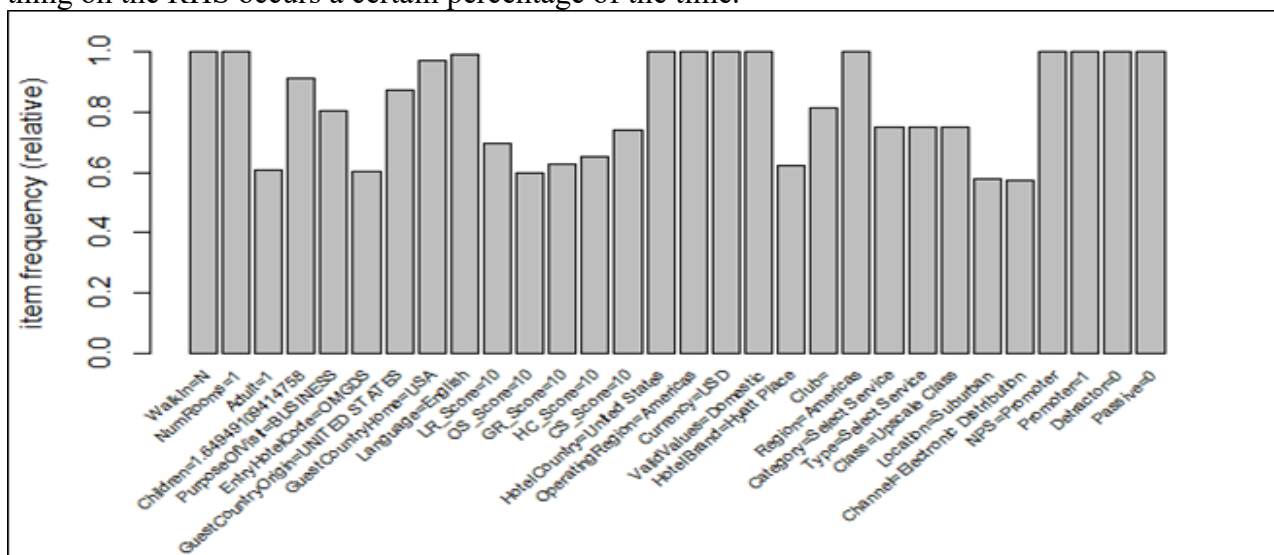
We used the criteria support and confidence to identify the most important relationships.

The support is an indication of how frequently the items appear in the data.

We originally looked at the overall data set. To do this, we converted the data-frame to transaction data by converting numbers or strings to factors in the data set, then convert ultimately the data-frame to a transaction data set. We also restricted the data to exclusively look at promoters.

At the first level we have one-item sets (commonly found). The next level is the two-items sets - they need to have the property that each of their subsets must be frequent enough to include (sufficient support).

Each rule states that when the thing or things on the left hand side of the equation occurs, the thing on the RHS occurs a certain percentage of the time.



Using a generic support of 2% and a confidence of 50%, the set of rules is 198, which is a lot of rules to examine manually. Hence, it was plotted using the arulesViz package.

The lift is a measure of the performance of association rule at predicting with respect to the population.

Lift serves as a measure of interestingness (from textbook) and the higher the lift the more attention it gets.

- More favorable are rules that have high support and high confidence, Some rules with high lifts seem to have low support, Support is an indication of how frequently the items appear in the data.

```
> # creating rules with respect to NPS_type
> Rules.NPS <- apriori(NPS.df, parameter = list(support = 0.2, confidence = 0.5),
+                   appearance = list(rhs=c("NPS=Promoter", "NPS=Detractor",
+                   "NPS=Passive"), default="lhs"))
Apriori

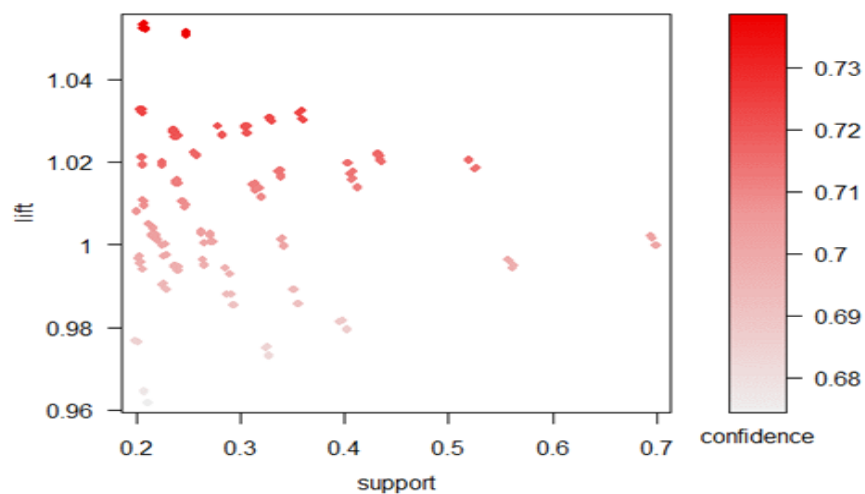
Parameter specification:
confidence minval smax arem aval originalsupport maxtime support minlen maxlen target ext
0.5 0.1 1 none FALSE TRUE 5 0.2 1 10 rules FALSE

Algorithmic control:
filter tree heap memopt load sort verbose
0.1 TRUE TRUE FALSE TRUE 2 TRUE

Absolute minimum support count: 3487

set item appearances ...[3 item(s)] done [0.00s].
set transactions ...[61 item(s), 17439 transaction(s)] done [0.01s].
sorting and recoding items ... [20 item(s)] done [0.00s].
creating transaction tree ... done [0.01s].
checking subsets of size 1 2 3 4 5 6 7 done [0.01s].
writing ... [198 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
```

Scatter plot for 198 rules



We re-ran the good rule with a lift of 1.05 and got rules with the highest lift; this amounted to twelve rules.

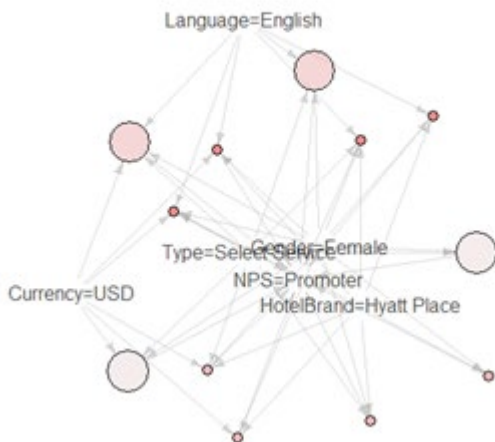
set of 12 rules

```
> inspect(Good.Rules)
```

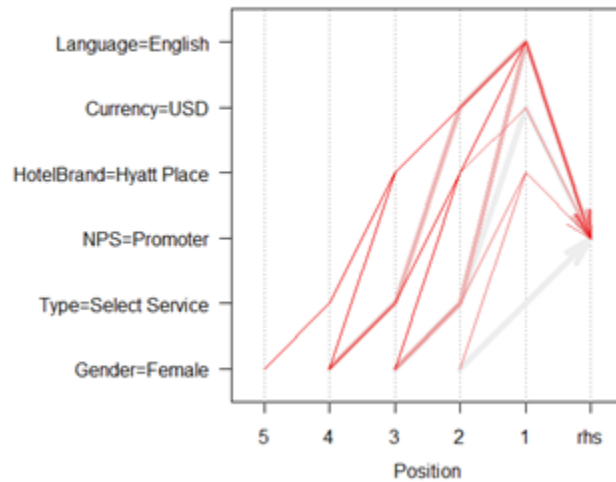
	lhs	rhs	support	confidence	lift	count
[1]	{HotelBrand=Hyatt Place,Gender=Female}	=> {NPS=Promoter}	0.2073513	0.7376581	1.052272	3616
[2]	{Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2476633	0.7365280	1.050659	4319
[3]	{HotelBrand=Hyatt Place,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2073513	0.7376581	1.052272	3616
[4]	{Language=English,HotelBrand=Hyatt Place,Gender=Female}	=> {NPS=Promoter}	0.2068926	0.7382852	1.053166	3608
[5]	{Currency=USD,HotelBrand=Hyatt Place,Gender=Female}	=> {NPS=Promoter}	0.2073513	0.7376581	1.052272	3616
[6]	{Language=English,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2466311	0.7371037	1.051481	4301
[7]	{Currency=USD,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2476633	0.7365280	1.050659	4319
[8]	{Language=English,HotelBrand=Hyatt Place,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2068926	0.7382852	1.053166	3608
[9]	{Currency=USD,HotelBrand=Hyatt Place,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2073513	0.7376581	1.052272	3616
[10]	{Language=English,Currency=USD,HotelBrand=Hyatt Place,Gender=Female}	=> {NPS=Promoter}	0.2068926	0.7382852	1.053166	3608
[11]	{Language=English,Currency=USD,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2466311	0.7371037	1.051481	4301
[12]	{Language=English,Currency=USD,HotelBrand=Hyatt Place,Type=Select Service,Gender=Female}	=> {NPS=Promoter}	0.2068926	0.7382852	1.053166	3608

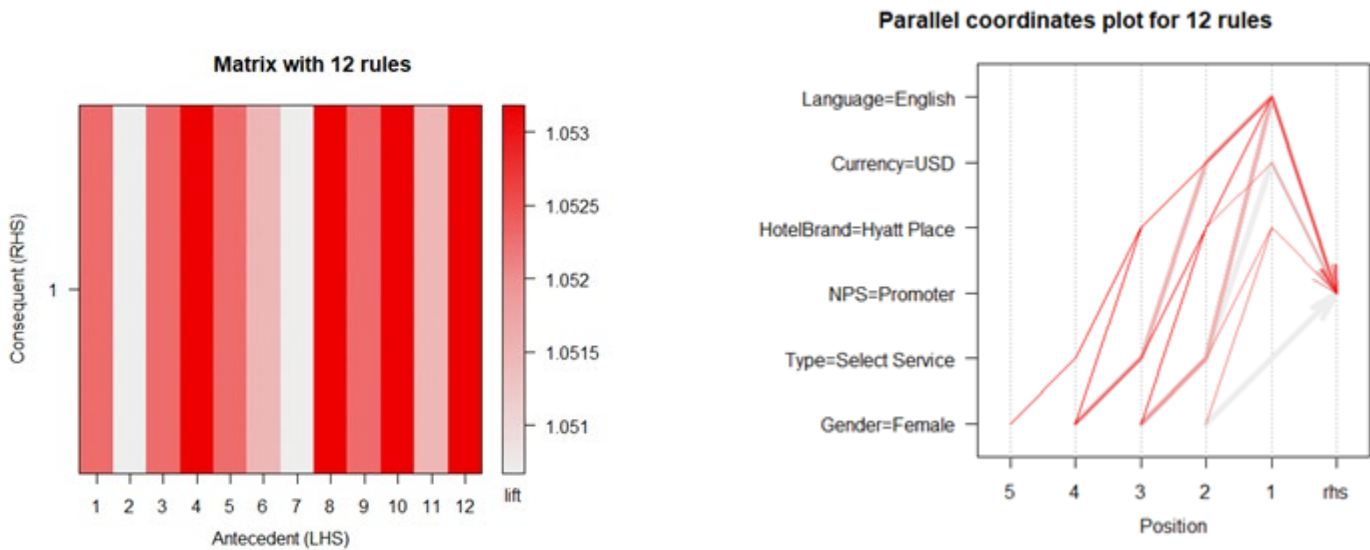
Graph for 12 rules

size: support (0.207 - 0.248)
color: lift (1.051 - 1.053)



Parallel coordinates plot for 12 rules





Support Vector Machine

Our Support Vector Machine utilized several more variables and produced an accuracy of 99.9%. This accuracy was higher, due in part, to the Passive and Detractors aggregated into a NonPromoter bucket. We took this approach as Hyatt's goal should be to convert all non-promoters to become promoters.

```
> syr <- read.csv("ProjectSurveyUSA_12.10.18v2.csv")
>
> library(kernlab)
> table(syr$NPS)
```

```
NonPromoter    Promoter
      5214      12225
```

```
> randindex <- sample(1:dim(syr)[1])
> summary(randindex)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
      1    4360    8720    8720   13080   17439
```

```
> length(randindex)
```

```
[1] 17439
```

```
> cutPoint2_3 <- floor((2/3)*dim(syr)[1])
```

```
>
```

```
> #Training set
```

```
> trainData <- syr[randindex[1:cutPoint2_3],]
```

```
>
```

```
> #Test set
```

```
> testData <- syr[randindex[(cutPoint2_3+1):dim(syr)[1]],]
```

```
>
```

```
>
```

```
> #Build model based on Training
```

```
> svmOutput <- ksvm(NPS ~., data = trainData, kernel = 'rbfdot',
+                   kpar = 'automatic', C=50, cross = 3, prob.model = TRUE)
```

```
>
```

```
> #Make a prediction
```

```
> svmPred <- predict(svmOutput, testData, type = 'votes')
```

```
>
```

```
> compTable <- data.frame(testData[,35], svmPred[1,])
```

```
> table(compTable)
```

```

               sumPred.1...
testData...35.    0     1
  NonPromoter    2 1711
   Promoter    4094     6
```


Best Model

The Ordered Logistic model provides more detail with its prediction but the Support Vector Machine achieves the most important goal for Hyatt: Identify Promoters and Non-Promoters. The Non-Promoters should remain Hyatt's biggest focus with the assumption that Promoters will remain in this category. If all hotels only improve then it should be reasonable to assume that Promoters will only become more satisfied and Non-Promoters will slowly convert. Yes, there may be some Promoters that do not like some of the changes, but the net gain will be positive for Promoters.

Business Question Insights

- What is the overall NPS?
 - Provide an overview of the count and percentage for each category of Promoter, Passive and Detractor

```
> #Calculate the total and percentage of Promoter, Passive and Detractor
> NPSsummary <- as.data.frame(table(ProjectSurveyUSA$NPS))
> colnames(NPSsummary) <- c("Category", "Freq")
> NPSsummary$Percent <- percent((NPSsummary$Freq / sum(NPSsummary$Freq)))
> NPSsummary
```

	Category	Freq	Percent
1	Detractor	2006	11.5%
2	Passive	3208	18.4%
3	Promoter	12225	70.1%

- What is the NPS amongst different demographics
 - By US Region

```
> #Calculate the NPS by Region
> ProjectSurveyUSA$stateRegion <- ifelse(ProjectSurveyUSA$stateName %in% c("washington", "oregon", "california", "idaho", "nevada", "montana", "wyoming", "utah", "arizona", "colorado"), "West",
+                                       ifelse(ProjectSurveyUSA$stateName %in% c("new mexico", "oklahoma", "texas", "arkansas", "louisiana", "mississippi", "alabama"), "South",
+                                       ifelse(ProjectSurveyUSA$stateName %in% c("north dakota", "south dakota", "nebraska", "kansas", "minnesota", "iowa", "missouri", "wisconsin", "illinois", "michigan", "indiana", "kentucky", "ohio"), "Midwest",
+                                       ifelse(ProjectSurveyUSA$stateName %in% c("maine", "new hampshire", "vermont", "massachusetts", "rhode island", "connecticut", "new york", "new jersey", "pennsylvania", "delaware", "maryland", "district of columbia"), "Northeast",
+                                       ifelse(ProjectSurveyUSA$stateName %in% c("west virginia", "virginia", "tennessee", "north carolina", "south carolina", "georgia", "florida"), "Southeast",
+                                       "ERROR"))))
```



```
> #View the values populated in stateRegion. ERROR should not be included
> unique(ProjectSurveyUSA$stateRegion)
[1] "Southeast" "Northeast" "South" "Midwest" "West"
> #If ERROR appears from the previous line then run the code below to see the row where the error is occurring. If there is not an error you should get integer(0) as output
> which(grepl("ERROR", ProjectSurveyUSA$stateRegion))
integer(0)
>
> #Get final view of the dataset to make sure the column was added
> View(ProjectSurveyUSA)
>
> #Create a Numerator and Denominator for each region and divide
> regionPromote <- function(region)
+ {
+   n <- length(which(ProjectSurveyUSA$NPS == "Promoter" & ProjectSurveyUSA$stateRegion == region))
+   d <- length(which(ProjectSurveyUSA$stateRegion == region))
+   return(percent(n/d))
+ }
>
> NEpromote <- regionPromote("Northeast")
> SEpromote <- regionPromote("Southeast")
> MWpromote <- regionPromote("Midwest")
> Spromote <- regionPromote("South")
> Wpromote <- regionPromote("West")
>
> rp <- data.frame(NEpromote, SEpromote, MWpromote, Spromote, Wpromote)
>
> colnames(rp) <- c("NorthEastPromote", "SouthEastPromote", "MidWestPromote", "SouthPromote", "WestPromote")
>
> rp
```

	NorthEastPromote	SouthEastPromote	MidWestPromote	SouthPromote	WestPromote
1	69.5%	70.1%	72.6%	69.4%	68.4%

- By State

- 31 states + Washington D.C.

```
> StatePromotePercent <- function()
+ {
+   StatePromo <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+   group_by(stateName = fct_inorder(stateName)) %>%
+   summarise(Percent = percent(mean(promoter)))
+   StatePromo <- as.data.frame(StatePromo)
+   return(StatePromo)
+ }
> StatePromotePercent()
```



	stateName	Percent
1	florida	68.8%
2	district of columbia	65.3%
3	new mexico	76.0%
4	georgia	70.8%
5	texas	68.9%
6	tennessee	74.1%
7	massachusetts	73.0%
8	louisiana	65.0%
9	south carolina	69.3%
10	illinois	71.2%
11	kentucky	66.7%
12	ohio	73.5%
13	north carolina	68.8%
14	colorado	71.9%
15	new jersey	70.5%
16	virginia	71.7%
17	california	63.4%
18	missouri	78.8%
19	wisconsin	74.1%
20	minnesota	70.5%
21	oklahoma	69.0%
22	pennsylvania	75.5%
23	arizona	78.0%
24	arkansas	70.3%
25	maryland	72.5%
26	new york	75.0%
27	connecticut	74.8%
28	alabama	67.7%
29	idaho	80.0%
30	michigan	74.6%
31	indiana	73.3%
32	mississippi	73.1%

- By Hotel (provides same insight as by city)

```
> HotelPromotePercent <- function()
+ {
+   hotelPromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+   group_by(HotelAbbrv = fct_inorder(HotelAbbrv)) %>%
+   summarise(Percent = percent(mean(promoter)))
+   hotelPromoter <- as.data.frame(hotelPromoter)
+   return(hotelPromoter)
+ }
> HotelPromotePercent()
      HotelAbbrv Percent
1      HR Grand Cypress  71.5%
2      HY Arlington    62.0%
3      GH Washington   66.5%
4      PH Washington   81.0%
5      HR Crystal City  61.7%
6      HR Washington   60.8%
7      HP Albuquerque Airport 69.3%
8      HP Albuquerque Uptown 80.0%
9      HP Windward     78.0%
10     HP Johns Creek  80.3%
```

- By Gender

```
> GenderPromotePercent <- function()
+ {
+   genderPromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+   group_by(Gender = fct_inorder(Gender)) %>%
+   summarise(Percent = percent(mean(promoter)))
+   genderPromoter <- as.data.frame(genderPromoter)
+   return(genderPromoter)
+ }
> GenderPromotePercent()
      Gender Percent
1      Female  72.2%
2      Male   69.1%
3              52.6%
4 Prefer not to answer 56.6%
```

- By Purpose of Visit

```
> PurposePromotePercent <- function()
+ {
+   purposePromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+   group_by(PurposeOfVisit = fct_inorder(PurposeOfVisit)) %>%
+   summarise(Percent = percent(mean(promoter)))
+   purposePromoter <- as.data.frame(purposePromoter)
+   return(purposePromoter)
+ }
> PurposePromotePercent()
      PurposeOfVisit Percent
1      BUSINESS   69.7%
2      LEISURE    71.7%
```

- What days of the week are most popular amongst guests?
 - Check-in Day



```
> CheckInPromotePercent <- function()
+ {
+   CheckInPromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+     group_by(CheckInDay = fct_inorder(CheckInDay)) %>%
+     summarise(Percent = percent(mean(promoter)))
+   CheckInPromoter <- as.data.frame(CheckInPromoter)
+   return(CheckInPromoter[order(desc(CheckInPromoter$Percent)),])
+ }
> CheckInPromotePercent()
  CheckInDay Percent
3   Saturday  72.4%
5   Thursday  71.2%
1    Friday   70.4%
7  Wednesday  70.1%
2    Monday   69.5%
4    Tuesday  68.1%
6    Sunday   67.6%
```

○ Check-out Day

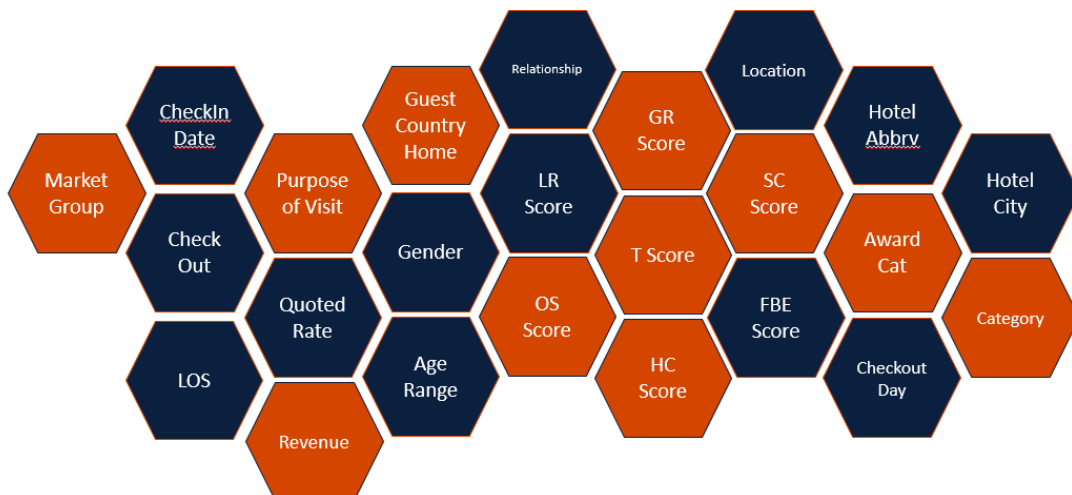
```
> CheckOutPromotePercent <- function()
+ {
+   CheckOutPromoter <- ProjectSurveyUSA %>% mutate(promoter = ifelse(NPS == "Promoter", 1, 0)) %>%
+     group_by(CheckOutDay = fct_inorder(CheckOutDay)) %>%
+     summarise(Percent = percent(mean(promoter)))
+   CheckOutPromoter <- as.data.frame(CheckOutPromoter)
+   return(CheckOutPromoter[order(desc(CheckOutPromoter$Percent)),])
+ }
> CheckOutPromotePercent()
  CheckOutDay Percent
2   Saturday  72.0%
1    Tuesday  71.7%
6    Sunday   71.5%
4    Monday   70.9%
5    Friday   68.5%
3   Thursday  67.9%
7  Wednesday  67.1%
```

- What attributes are best at predicting if a guest will be a promoter?

MarketGroup
CheckInDate
CheckOut
LOS
PurposeOfVisit
QuotedRate
Revenue
GuestCountryHome
Gender
AgeRange
LR_Score
OS_Score
GR_Score



T_Score
HC_Score
CS_Score
SC_Score
IS_Score
CI_Score
FB_Frequency
FBE_Score
HotelAbbrev
AwardCat
HotelCity
HotelBrand
Club
Category
Type
Class
Location
Relationship
GPTier
Channel
NPS
CheckInDay
CheckOutDay





Recommendations

Orange recommends Hyatt gather more data and places a stronger emphasis on making fields required to avoid the large number of NAs that were present in the data. This can be achieved by offering the surveys online. While this may reduce the number of surveys it will at least provide better data, which will ultimately lead to better insights and decision making.

If Hyatt is interested in expanding its reach, Orange recommends expansions to the Midwest and West with an understanding and objective to place hotels in suburban settings. Franchising these hotels should be avoided as it appears franchised hotels have an association to the Bottom 10 hotels.