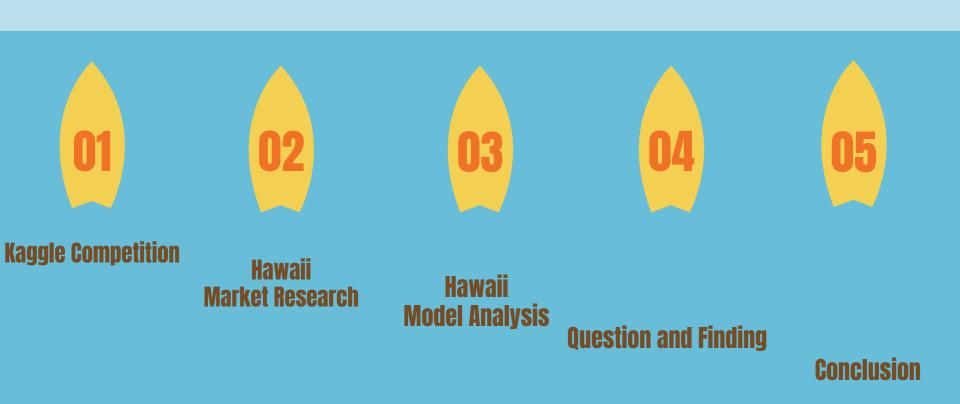


# BUDT 758T - Hawaii Identifying Buying Targets for AirBnB Investment Property

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## **Kaggle Competition**

#### Data processing steps:

- Convert variables into numeric or factor types & fill NA values.
- Construct four models: Logistic regression, Bagged Trees, Random forest and XGBoost.
- Compare AUC and choose the best model.

```
# Function to convert to numeric
ConvertNumeric = function(dfCol, NAValue){
  val <- (gsub("\\$", "", dfCol))
  val <- (gsub("\\", "", val))
  val <- suppressWarnings(as.numeric(gsub("\\,", "", val)))
  val[is.na(val)] = NAValue
  return(val)
}

# Function to convert to factor
ConvertFactor = function(dfCol, GoodVector, NAValue){
  dfCol[!(dfCol %in% GoodVector)] = NAValue
  dfCol = as.factor(dfCol)
  return(dfCol)
}</pre>
```

Model	AUC
Logistic	0.8765
Bagged Trees	0.9092
Random Forest	0.9318
XGBoost	0.9399

## **Variables used**

Numeric Variables: accommodates, availability\_30,availability\_90,availability\_365,bathrooms,bedrooms,beds,a menities, cleaning\_fee, extra\_people, guests\_included, host\_listings\_count, host\_response\_rate, host\_verifications, maximum\_nights, minimum\_nights, monthly\_price, price, review\_scores\_accuracy, review\_scores\_checkin, review\_scores\_cleanliness, review\_scores\_communication, review\_scores\_location, review\_scores\_rating, review\_scores\_value, security\_deposit, weekly\_price

Factor Variables: bed\_type, cancellation\_policy, host\_identity\_verified, host\_has\_profile\_pic, host\_is\_superhost, host\_response\_time, host\_since, instant\_bookable, is\_location\_exact, market, property\_type, require\_guest\_phone\_verification, require\_guest\_profile\_picture, requires\_license, room\_type



## **Quick Look on Hawaii Market**







Most to Least Visited Places: Oahu 49.6% > Maui 24.4% > Big Island 14.5% > Kauai 11.5%

## **Quick Look on Hawaii Market**

## **Visiting Purpose**

- Pleasure/Vacation
  - **Honey Moon** 
    - Wedding
    - Conference

## **Total Visitor Spending**

- **Increasing since** 2003 (to 2019)
- Lodging, largest spending category









## **Length of Stay**

- First time visitors: 31%, 8.24 days
- Repeating visitors: 68%, 9 days



## **Total Property**

14247

## **Kaggle Part (Team 2)**









#### Data Processing

- Convert selected variables into either numeric variables or factor variables
- Replace NAs for numeric variables with the average value or the highest frequency value
- Substitute NAs for factor variables with "Other" or the highest frequency value

#### Final model

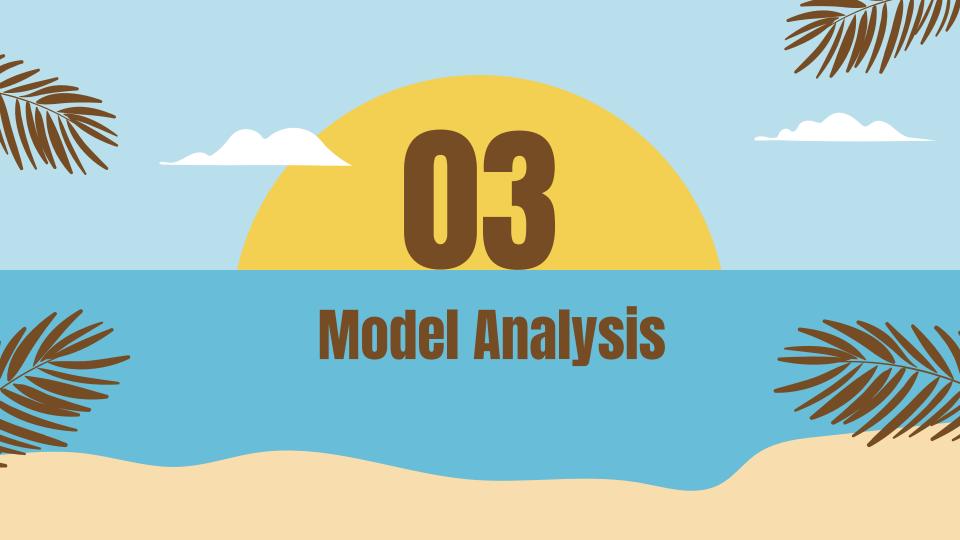
XGboost
Parameters:
Trees = 1000,
set\_engine("xgboost"),
set\_mode("classification")

#### AUC

0.9399

#### Variables

Numeric variables: accommodates. availability 30, availability 60, availability 90, availability 365, b athrooms, bedrooms, beds, amenities, cleaning\_fee, extra\_people, quests\_included, host\_listings\_count, host\_response\_rate, host\_verifications, maximum\_nights, minimum\_nights, monthly price, price, review scores accuracy, review\_scores\_checkin, review\_scores\_cleanliness, review scores communication, review scores location, review scores rating, review scores value, security deposit, weekly\_price Factor variables: bed\_type, cancellation\_policy, host identity verified, host has profile pic, host is superhost, host\_response\_time, host\_since, instant\_bookable, is location exact, market, property type, require\_guest\_phone\_verification, require quest profile picture, requires license, room type



## **Models - variables and performance**

#### Kaggle variables

recipe1 <recipe(high\_booking\_rate ~ ., data = dfa\_analysis) %%
step\_rm(id, access, city, description, host\_about, host\_acceptance\_rate, host\_location, host\_neighbourhood, house\_rules, interaction,
is\_business\_travel\_ready, latitude, longitude, neighborhood\_overview, neighbourhood, notes, space, state, transit, zipcode, square\_feet,
`{randomControl}`)

.metric	.estimator	.estimate
<chr></chr>	<chr></chr>	<dbl></dbl>
roc_auc	binary	0.9319582

#### Hawaii model with updated variables

.metric <chr></chr>	.estimator <chr></chr>	.estimate
roc_auc	binary	0.9374196

**Models: XGboost** 

#### **Variables Selection**

- Weekly\_discount replace weekly\_price
  - = 1 weekly\_price/(daily\_price\*7)
- Add specific amenity columns (True/False): Family-kid friendly, parking, washer, dryer, hair dryer, Wifi

## **Models - Cutoff and Cost Matrix**

#### Costs at different cutoff

Cutoff	Cost
0.3	2318
0.4	1832
0.5	1463
0.6	1187
0.7	992

**Models: XGboost** 

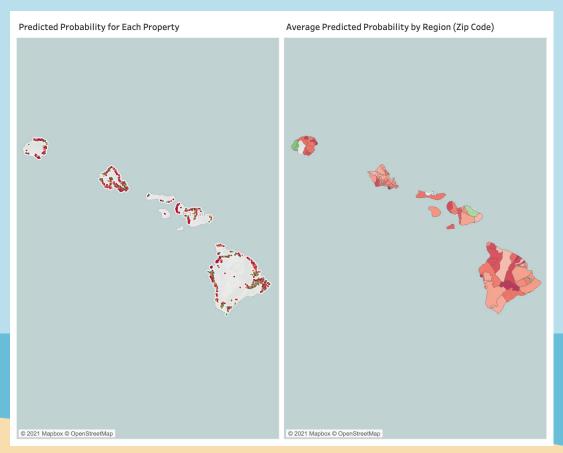
#### **Cut-off determination**

This is a business advice for property purchasing, so we'd like lower False Positive. In cost matrix, we set cost of False positive 3 times higher than true positive. For the cost matrix, we found that cutoff = 0.7 has lowest cost.

Cost Matrix		Actual	Class
		0	1
Predicted	0	0	1
Class	1	3	1

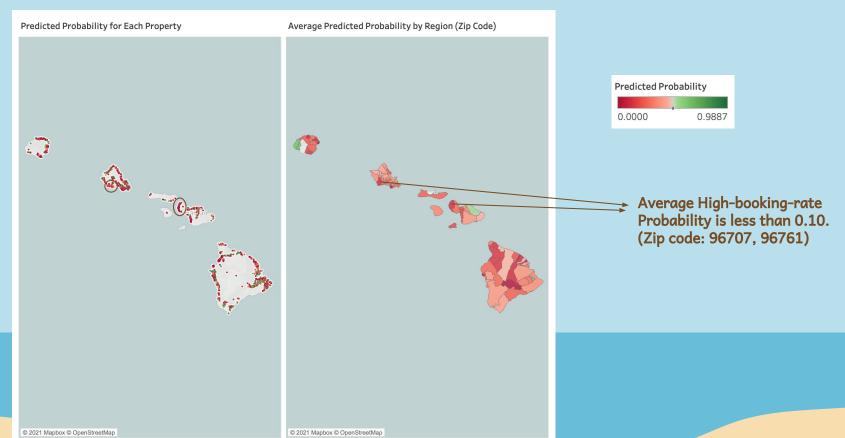


## **Q1: Which region is not good for investment?**



Predicted Probability	
0.0000	0.9887

## **Q1: Which region is not good for investment?**



## **Q2: How does weekly price discount affect probability?**

Weekly Discount Interval	Number	High-booking-rate Probability
>= 30%	92	0.4546
20% ~ 30%	127	0.6417
10% ~ 20%	197	0.6730
0% ~ 10%	183	0.6021
No discount	177	0.5967

High-booking-rate Probability for property cannot stay higher than a week = **0.2124** 

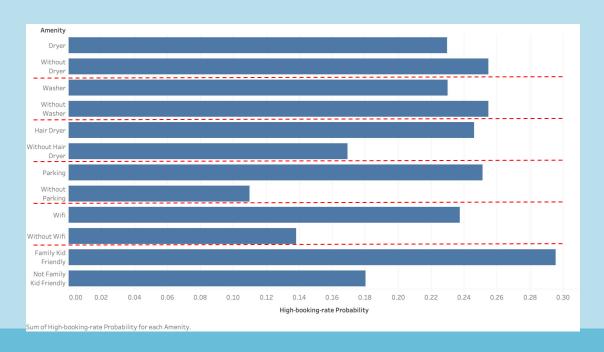
Two-sample t test for unequal sample size (unequal variance):

- Difference between 10% ~ 20% and no discount is significant (p-value = 0.01).
- Difference between 10% ~ 20%
   and 20% ~ 30% is not significant
   (p-value = 0.3623).
- Difference between 10% ~ 20% and 0 ~ 10% is significant (p-value = 0.02).

#### Management insights:

- → Allow customers to book more than a week (maximum\_nights >= 7).
- → Choose 10% to 30% weekly discount as the price management strategy!

## **Q3: How do provided amenities affect probability?**



#### Management insights:

- → Add Wifi and Hair Dryer equipments into the property.
- → Try to add parking service.
- → Washer & Dryer not important.
- → Provide family-kid friendly service and equipments.



## **Conclusion**

	DESCRIPTION
Problem	The airbnb/ short rental market is one of the most competitive and ever-changing markets for real estate investors.
Goal	Screen & Identify the right properties to invest
Method	A projection model to identify the properties with high booking rates and thus high income generating potential.
AUC	94%
Limitations	?

# Investor's Goal

High Booking Rate, Avg. Length of Stay

**Operating Income** 

Occupancy, Rental Pricing

**Operating Cost** 

Net Operating Income



Property Age, Property Price (& Mortgage)



- Investor's Budget
- Property Location

## References

Major Cities/Locations: Honolulu, Haleiwa, Kailua-Kona, Hawi, Captain Cook, Kihe Paia, Princeville, and Kapaa (location)

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u-and-how-much.html

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recreational interests of visitors are changing: hiking> golfing

https://www.civilbeat.org/2019/07/9-charts-that-show-how-hawaii-tourism-is-chang

ing/

Illegal vacation rentals problems

Airbnb horror stories analysis

https://www.asherfergusson.com/airbnb/

Buying Property for Airbnb (location, type, accessibility, neighborhood)

https://www.fool.com/millionacres/real-estate-investing/rental-properties/buying-property-airbnb/

# THANK YOU Q&A

