Final Project

Team 3

IST 687 Fall 2022

Data Set Content

- Variables
 - Age
 - Sex
 - BMI
 - Children
 - Smoker
 - Region
 - Charges
- Data Source Medical Cost Personal Datasets | Kaggle
- Observations 1338

Original Business Questions

- What are the demographics of our patient population?
- What variables have the most effect on insurance charges?
- How can we predict an individual's insurance cost based on their variables/ how accurately can we forecast costs for each patient by looking at the data?
- What variables in the dataset can an individual control to potentially reduce their insurance charges, and by how much?
- Other than behavior of individuals such as smoking, what can cause increase of medical cost?
- What variables have the strongest correlation?

Demographics

Average Age: 39.21 years

• Range: 18-64

Average BMI: 30.66

• Range: 15.96-53.13

Average Children: 1.09

• Range: 0-5

Average Charges: \$13,270.42

• Range: \$1,121.87-\$63,770.43

Gender:

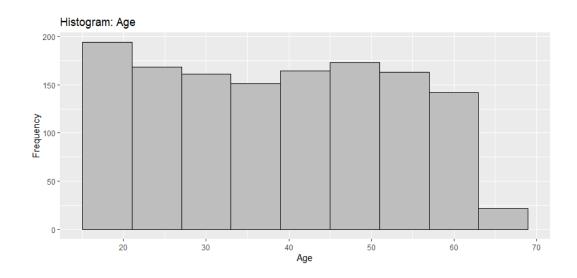
- Female 48.48%
- Male 50.52%

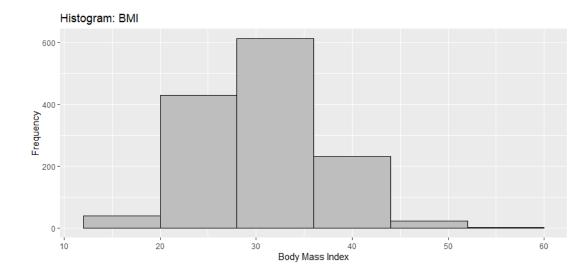
Percentage Smokers:

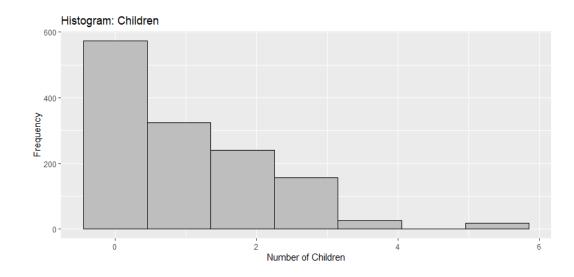
- Smoker 20.48%
- Non-smoker 79.52%

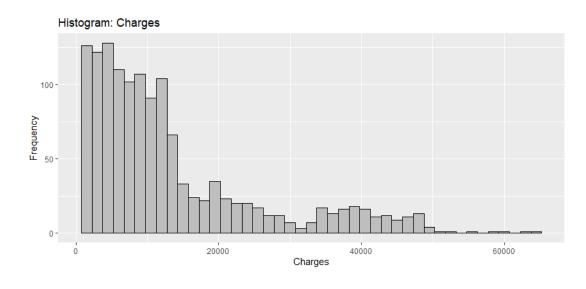
Population by Region:

- Southwest 325
- Southeast 364
- Northwest 325
- Northeast 324









Multivariate Linear Regression Model

- Initial Regression Model
 - charges = age + bmi + children + female + smoker + southwest + southeast + northwest
 - F-statistic = < 2.2e-16
 - Adjusted R-squared: 74.94%

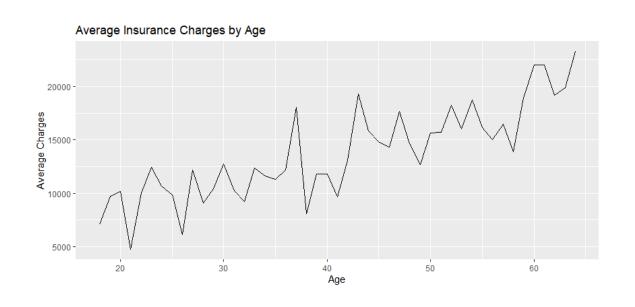
- Statistically Significant Model
 - charges = age + bmi + children + smoker
 - F-statistic = <2.2e-16
 - Adjusted R-squared = 74.89%

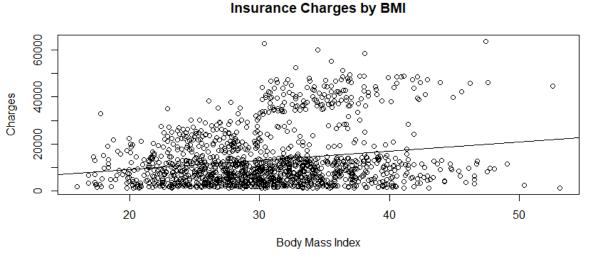
MLR Model Drilled Down

- Remove Age
 - charges = bmi + children + smoker
 - Adjusted R-squared: 66.07%
- Remove Children
 - charges = age + bmi + smoker
 - Adjusted R-squared: 74.69%

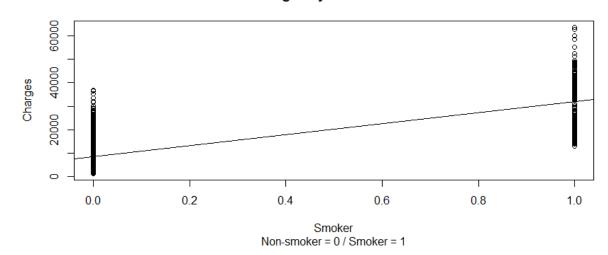
- Remove BMI
 - charges = age + children + smoker
 - Adjusted R-squared: 72.31%
- Remove Smoker
 - charges = age + bmi + children
 - Adjusted R-squared: 11.81%

Significant Variables Effects on Charges

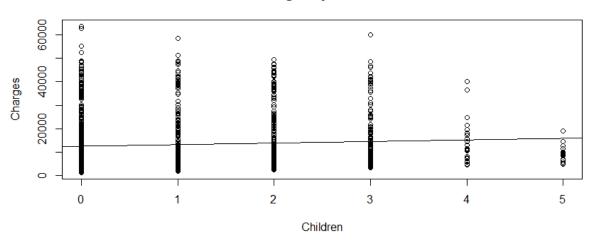




Insurance Charges by Non-smoker vs. Smoker

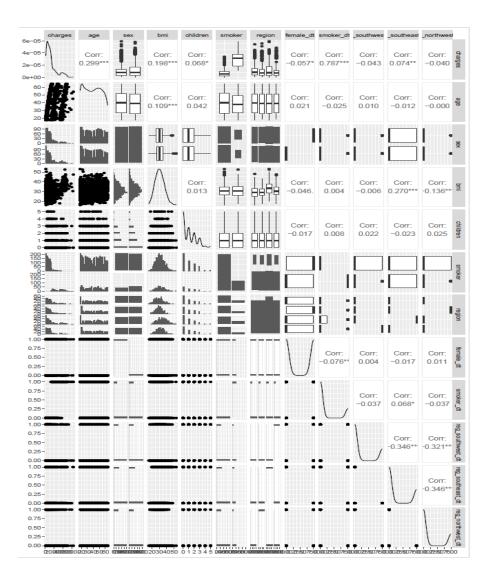


Insurance Charges by Number of Children



Correlation of Variables

- Most significant correlation
 - Charges and Smoker: 78%
 - Charges and Age: 30%
 - Southest Region and BMI: 27%



Prediction Model/Function for Business App

```
charges_function <- function(age,bmi,children,smoker) {
   model_charges <- lm(charges ~ age + bmi+ children + smoker_dt, data = ins_df)
   pred_df <- data.frame(age = age, bmi = bmi, children = children, smoker_dt = smoker)
   pred_charges <- predict(model_charges,pred_df, type = "response")
   return(pred_charges)
}
charges_function(27,32,1,0)|</pre>
```

Output: \$5,631.92

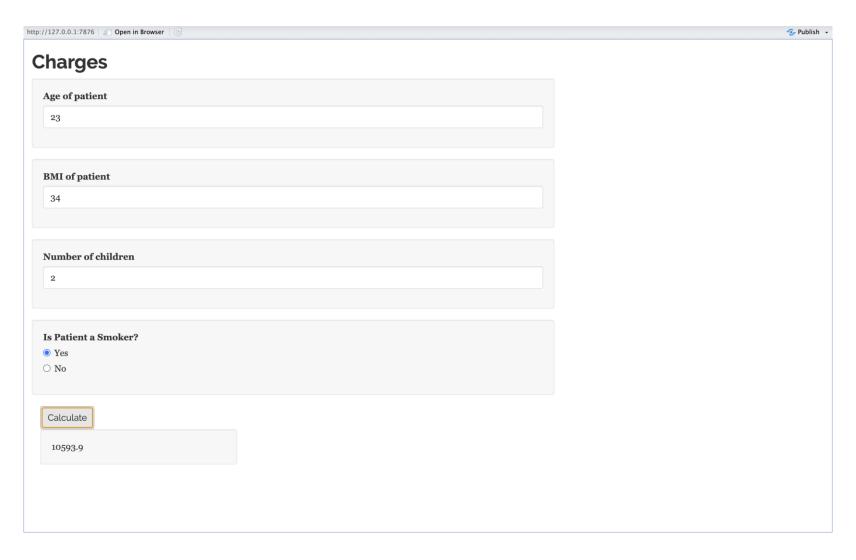
Additional Business Questions

- Controllable Variables
 - BMI
 - Smoking
 - Number of Children*
- Uncontrollable Variables
 - Age

R-squared: 74.89%	F-Statistic: <2.2e-16	
Variable	Coefficient	P-Value
Intercept	-12102.77	< 2e-16
Age	257.85	< 2e-16
ВМІ	321.85	< 2e-16
Children	473.50	0.000608
Smoker	23811.40	< 2e-16

^{*}Assumes children aren't already born

Shiny App for Web User



Limitatations of the Dataset

- Several large outliers that may affect assumptions made in the linear regression analysis
- Inabilty to address the high average cost of healthcare
- Dataset is from one insurance company
- Prediction models failing due to global events that disrupt healthcare system, e.g. the COVID-19 pandemic

Future Topics to Address The Limitations

- Dataset from various insurance companies and cross analysis of the cost across healthcare systems
- Dataset from other sources including government sources on the number of unisured patients and the role they play in the cost of healthcare
- Transparancy in medical bills including administrative cost and hidden fees to address discrepancies in the cost