Group 11 Final Presentation

Tom Tribe, Ken MacIver, Jundi Yang, Mei Huang

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Group 11: Diamonds Dataset



Group Members (photos)



Group Members (name, email, ORCID)

Tom Tribe

- tom.tribe2016@gmail.com
- **>** 0000-0002-5002-8066

Ken Maclver

- ▶ ken.maciver68@gmail.com
- 0000-0001-8999-4598

Jundi Yang

- ▶ ivyli112358@gmail.com
- 0000-0003-0888-9564

Mei Huang

- huangmei139@gmail.com
- 0000-0003-2401-0679

The Diamonds dataset

- ➤ This large dataset has 53940 rows (diamonds) of ten variables (approx 540,000 values)
- Slow to process!
- Nine of the variables are various measures of diamond size and quality, while the tenth is the price
- We selected diamonds because it was simple to understand what each variable was measuring, and to have the opportunity to work with a large dataset
- Particularly interested in which variables are most predictive of diamond price

The Variables

red font = categorical variable

- carat: the diamond's weight
- cut: a measure of quality (4 levels)
- color: a measure of colour quality (7 levels)
- clarity: a measure of clearness (6 levels)
- x: length in mm
- y: width in mm
- z: depth in mm
- depth: total depth percentage
- table: width of top of diamond relative to widest point
- price: the price of the diamond in US dollars

(List adapted from list at kaggle.com).

Summary of Numeric Variables

	carat	depth	table	price	X	у	z
sample size	53940	53940	53940	53940	53940	53940	53940
minimum	0.20	43.00	43.00	326.00	0.00	0.00	0.00
first	0.40	61.00	56.00	950.00	4.71	4.72	2.91
quartile							
median	0.70	61.80	57.00	2401.00	5.70	5.71	3.53
mean	0.80	61.75	57.46	3932.80	5.73	5.73	3.54
third	1.04	62.50	59.00	5324.25	6.54	6.54	4.04
quartile							
maximum	5.01	79.00	95.00	18823.0	0.74	58.90	31.80
IQR	0.64	1.50	3.00	4374.25	5 1.83	1.82	1.13
standard	0.47	1.43	2.23	3989.44	1.12	1.14	0.71
deviation							
skewness	1.12	-0.08	0.80	1.62	0.38	2.43	1.52
kurtosis	4.26	8.74	5.80	5.18	2.38	94.21	50.08

Cateogrical Summary

Cut

Fair

Good

	Count 1610		1610	4960		120	82	2155	51	13791	
_	Color	-	D	Е	F	=	G	H	<u> </u>	l J	
	Count	67	775 9	797	9542	2 112	92	8304	542	2 2808	_
Cla	arity	l1	IF	,	SI1	SI2	V	S1	VS2	VVS1	VVS2
Сс	unt	741	1790	13	3065	9194	81	71	12258	3655	5066

Very Good

Ideal

Premium

Pairs Plot

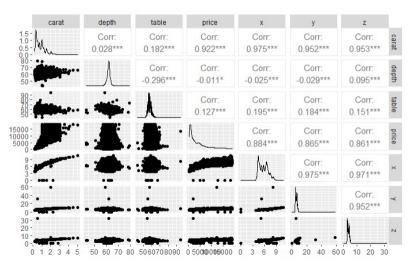
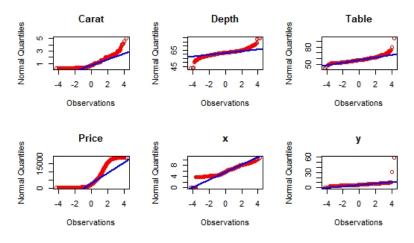
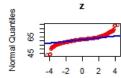


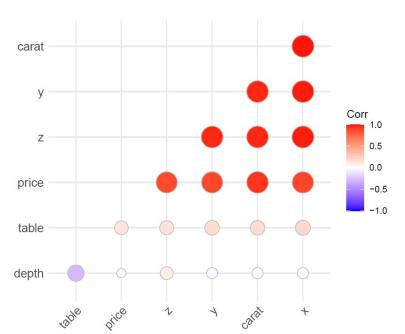
Figure 1: Pairs plot

Normal QQ Plots





Correlation Plot



Price by Cateogrical

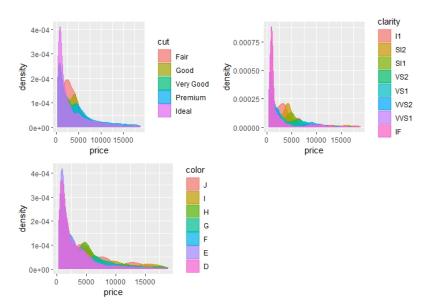


Figure 4: Price by Categorical

Leading Question 1

- ► How can we best predict diamond price?
- ► We intend to use the following techniques to investigate this question:
- Stepwise Regression, Principal Components Analysis, Principal Components Regression and an Exploratory Factor Analysis

Multiple Regression

- Starting with the full model we used a stepwise regression procedure to find the best model for predicting diamond price.
- According to AIC the best model was:
- price ~ carat + cut + color + clarity + depth + table + x
- ► All variables excluding y and z are significant in the model
- ▶ The 'best' model had an Adjusted R² of 91.98%

Regression Assumptions

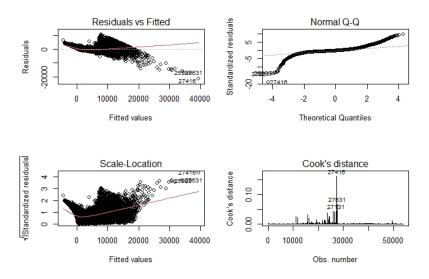
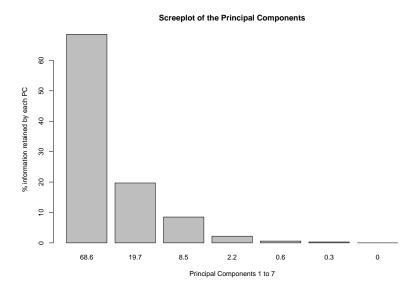


Figure 5: Regression Diagnostics

Principal Components Analysis: Screeplot



Principal Components Analysis: Values

```
## PC1 PC2

## carat 0.45032907 -0.049484287

## depth -0.03619535 -0.722832026

## table 0.10580678 0.678011337

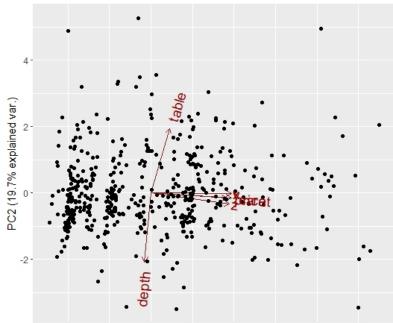
## price 0.42566082 -0.052989386

## x 0.45214098 -0.003202394

## y 0.45214716 -0.004959419

## z 0.44118352 -0.111906516
```

Biplot



Principal Components Regression

- We conducted a Principal Components Regression with diamond price as the response variable
- ► We found that all six principal components were significant in the model for predicting price
- ► However when we only used the first two principal components we were still able to explain over 80% of the variation in price

Factor Analysis

- We hypothesized that the variation in the data might be able to be explained by two factors
- ► These were: "Dimension x Price" and "Light Conductance"
- We tested this hypothesis with a Factor Analysis