1 EDA warm up! (13 points)

(6) a. To honor 4th of July (i.e. today!), a popular US news magazine wants to write an article on how much Americans know about capitals of their states. They devise a test that lists 100 cities and each respondent must guess the state in which the city can be found. Each correct answer earns one point, for a maximum of 100. The random sample of 5000 people had a distribution of scores that was *normally* distributed with mean 62 and standard deviation 12.

(2)i. How many countries can identify correctly the central 95% of the people in this sample?

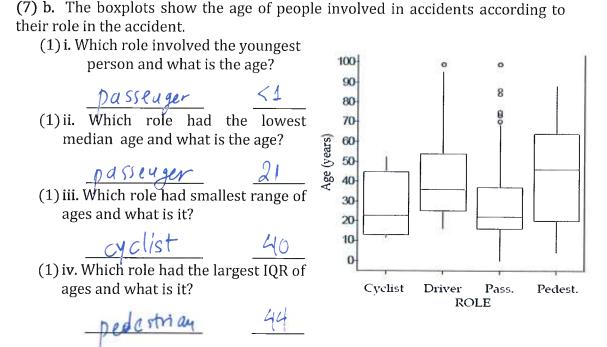
Normal model: 95% of the ausmer are in the luterral: [4-20,4125] ->
[62-2x12]62+2x12] ->
[39,86]
95% of people can identify between 38 and 16 countorier

(2)ii. A journalist claims that a score of 45 (and below) should be considered poor. Do you agree or not? Justify your answer scientifically.

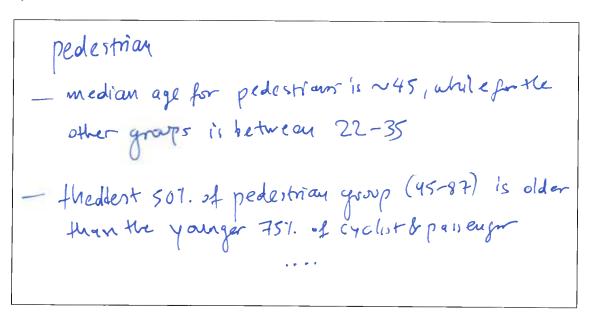
 $\frac{2}{2} = \frac{x - \mu}{\sigma} = \frac{45 - 62}{12} = -1.42$ $\frac{1}{2} = \frac{1}{2} = \frac{45 - 62}{12} = -1.42$ $\frac{1}{2} = \frac{1}{2} =$

(2)iii. Can you decide from the given information whether there are any outliers or not (either too smart or not that smart people)? What extra diagram/sketch (except the actual scores) would you need and how you were going to use it for answering?

No I need a boxplot



(3) v. A journalist asks you the tricky question: "Which role generally involves the oldest people?" Justify your answer scientifically in order to impress the journalist.



5 Modeling homelessness (15 points)

Homelessness is a problem in many large U.S. cities. To better understand the problem, a multiple regression was used to model the rate of homelessness based on several explanatory variables. The following data were collected for 50 large U.S. cities. The regression results appear below.

Homeless

number of homeless people per 10,000 in a city

Poverty

percent of residents with income under the poverty line

Unemployment

percent of residents unemployed

Temperature

average yearly temperature (in degrees F.)

Vacancy

percent of housing that is unoccupied

Rent Control

indicator variable, 1 = city has rent control, 0 = it has not

Dependent variable is Homeless. R squared (R^2) = 38.4%, Adjusted R^2 = 31.5%

Variable	Coefficient	SE (Coeff.)	t-ratio	p-value
Intercept	-4.275	3.465	-1.23	0.2239
Poverty	0.0823	0.0823	1.00	0.3228
Unemployment	0.159	0.218	0.73	0.4699
Temperature	0.135	0.0587	2.30	0.0262
Vacancy	-0.247	0.138	-1.79	0.0809
Rent Control:1	2.944	1.37	2.15	0.0373

(2) a. Which variables are associated with the number of homeless people in a city (using a 95% significance level)?

temperature & rent-control (p-value < 005)

NB: vacancy is >005

(4) b. Explain the meaning of the coefficients: **Temperature** and **Rent Control** in the *context of this problem*.

for cities with fixed parerty, we employment, etc.

(x11 raniables), an increase in 1°F leads to

a discrease of 0.14/ hondors per 10.000

Rentcontrol: for cities with fixed other variables,

cities with rentcontrol have 2.94 howeless pepla

under those after witheten

(2) c. Do the results suggest that having rent control laws in a city causes higher levels of homelessness? Explain.

No.

this is an observational study, so I cannot establish causal relations.

(4) d. If we created a new model by adding several more explanatory variables, which statistic should be used to compare them—the R² or the adjusted R²? Explain.

Adjusted & because it controls for madels with different that predictors

(3) e. Using the plot of residuals vs predicted values below, can you conclude that the above regression is a good fit (in terms of distribution of residuals)? Justify your choice.

6 PCA-phobia (10 points)

Ten participants are given a battery of personality tests, comprising the following items: Anxiety; Agoraphobia; Arachnophobia; Adventure; Extraversion; and Sociability (with a scoring range of 0 to 100). The purpose of this project is to ascertain whether the correlations among the six variables can be accounted for in terms of comparatively few latent variables or factors.

Part	Anx	Agora	Arach	Adven	Extra	Soci
1	71	68	80	44	54	52
2	39	30	41	77	90	80
3	46	55	45	50	46	48
4	33	33	39	57	64	62
5	74	75	90	45	55	48
6	39	47	48	91	87	91
7	66	70	69	54	44	48
8	33	40	36	31	37	36
9	85	75	93	45	50	42
10	45	35	44	70	66	78

(3) i. Provide some descriptive statistics on the dataset (mean, st. deviations for each item) and justify whether normalization is needed for this problem.

agora:	52.8	sd: 19.04 sd: 18.08	allcloreso
avach:	564	sd; 22.24 sd; 17.99 sd; 17.70	might not be ucessary to normalize
Extra:	59.3 58.5	Sd: 18.44	Without C

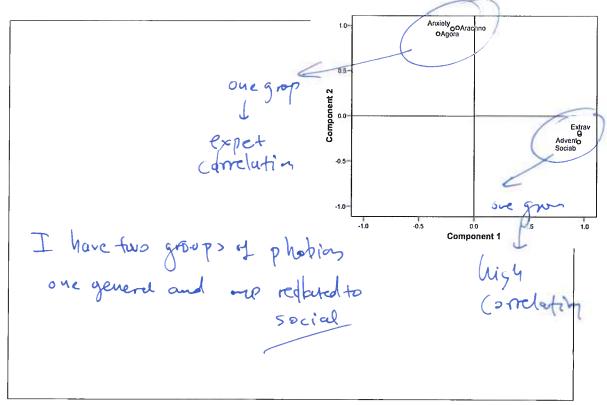
(2) ii. Running PCA on this (small) dataset gives the following results in terms of the principal components (NB: here the full results with 6 components are presented), along with the eigenvalues and the % of variance explained by each eigenvalue. As an experienced data scientist how many eigenvalues would you pick? Justify your choice.

Component	Total	% of Variance
1	4.164	69.397
2	1.612	26.862
3	.144	2.396
4	.052	.867
5	.023	.383
6	.006	.095

First tour PCAs explain = 96.3%. A He variance in data. They also are > 1 (eigender)

so we pick 2

(4) iii. The plot for two PCs is given below. How would you communicate to a non-data-scientist your final conclusion from this dataset by using this figure and the results from ii.)? Can you predict (without computing) what would be the correlations between the initial 6 variables?



7 (fast) regularization check questions (6 points)

Explain in one or two sentences why the statements are true (or false).

(2)a. L2 (Ridge) regularization is more robust to outliers than L1 (Lasso).

there is no difference.

(2)b. In terms of feature selection, L2 (Ridge) regularization is preferred (to L1-Lasso) since it comes up with sparse solutions.

(10), 1950 > feature selection (spane) because many coefficient jo to 0

(2)c. Nearly all the algorithms we have been through this course, have a tuning parameter which is performing regularization (even if we didn't call it that). In the concept of using PCA as a dimension reduction technique (i.e. reduce training data down to k dimensions with PCA and use these as features), one claims that higher k means less regularization. Do you agree or not?

Yes, 1 k more parameters => less generalitation less reguralitation

Bonus question (1 point)

What is Jerry's favorite color?

Answer: