Fall 2020 Midterm I SOLUTIONS

- 1. [0.5 point] The overall rate of diabetes in county A is greater than in county B. Therefore, the rate of diabetes for each age group in county A must be greater than the rate of diabetes in the corresponding age group in county B. Is this statement true or false?
- (a) True
- (b) False

SOLUTION: (b) False

2. [1 point total] Use the data below on a study looking at the effects of anger (measured by the Speilbere Trait Anger Scale test) and coronary heart disease (CHD) to answer the following questions.

	Low Anger	Moderate Anger	High Anger	Total
CHD	53	110	27	190
No CHD	3057	4621	606	8284
Total	3110	4731	633	8474

- 2. i) [0.5 point] What percent of those individuals who were classified as "High Anger" developed CHD?
- (a) Approximately 14%
- (b) Approximately 4%
- (c) Approximately 0.3%
- (d) Approximately 7%

SOLUTION: (b) Approximately 4%

- 2. ii) [0.5 point] This percent is part of the
- (a) Marginal distribution of CHD
- (b) Conditional distribution of anger given CHD
- (c) Marginal distribution of anger
- (d) Conditional distribution of CHD given high anger

SOLUTION: (d) Conditional distribution of CHD given high anger

3. [0.5 point] You are given a dataset, covid_data which has 6 columns (id, county, state, num_deaths, population and num_uninsured). Which line of code could you run so that there are exactly 5 columns in the output data frame?

```
(a) covid_data %>% rename(county_name = county)
```

- (b) covid_data %>% select(-num_uninsured)
- (c) covid_data %>% filter(state == "California")
- (d) covid_data %>% select(county, population)

SOLUTION: (b) covid_data %>% select(-num_uninsured)

- 4. [0.5 point] With covid_data %>% arrange(state, -population), how will this line of code sort the data?
- (a) Sort state in descending order first, then population in ascending order
- (b) Sort state in ascending order first, then population in ascending order
- (c) Sort state in ascending order first, then population in descending order
- (d) Sort population in ascending order first, then state in descending order

SOLUTION: (c) Sort state in ascending order first, then population in descending order

- 5. [1 point] What functions are necessary to visualize the distribution of a categorical variable? Choose all that apply.
- (a) geom_histogram()
- (b) ggplot()
- (c) geom_point()
- (d) $geom_bar()$
- (e) aes()
- (f) geom_cat()

SOLUTION: (b) ggplot(), (d) geom_bar(), (e) aes()

6. [2.5 points total] In your job as an analyst, your supervisor asks you to analyze data from the National Survey on Drug Use and Health from the Substance Abuse and Mental Health Data Archive.

Each row in the dataset drug_dat corresponds to an age group, with variables summarizing drug use across ages. The variable heroin_use gives the percentage of heroin use for the corresponding age group. Here are the first six rows of age and heroin_use

```
drug_dat <- read_csv("./drug_use_by_age_shaziap1.csv")
drug dat %>% select(age, heroin use) %>% head()
```

#	A .	tibb	ole:	6	х	2
	age	е	heroin_use			ıse
	<c]< td=""><td>hr></td><td></td><td colspan="3"><dbl></dbl></td></c]<>	hr>		<dbl></dbl>		
1	12			().()25
2	13			().()3
3	14			().()5
4	15			().()4
5	16			().()3
6	17			().1	L
	# 1 2 3 4 5 6	<pre># A</pre>	<pre># A tibb age <chr> 1 12 2 13 3 14 4 15 5 16 6 17</chr></pre>	<pre># A tibble: age hero <chr> 1 12 2 13 3 14 4 15 5 16 6 17</chr></pre>	<pre># A tibble: 6 age heroin <chr> 1 12 0 2 13 0 3 14 0 4 15 0 5 16 0 6 17 0 </chr></pre>	<pre># A tibble: 6 x age heroin_u <chr> <dh 0.0="" 0.1="" 1="" 12="" 13="" 14="" 15="" 16="" 17="" 2="" 3="" 4="" 5="" 6="" <="" pre=""></dh></chr></pre>

6. i) [0.5 point] What type of variable is Heroin Usage (%)? Select all that apply.

- (a) Categorical
- (b) Quantitative
- (c) Nominal
- (d) Ordinal
- (e) Continuous
- (f) Discrete

SOLUTION: (b) Quantitative, (e) Continuous

SOLUTION:

blank 1: age

blank 2: x

blank 3: scatter (or line)

blank 4: relationship is positive (increasing)

7. [3.5 points total] The dataset food_data includes percentage of food intake for different categories of food, with a row of data for each of 170 countries. The dataset also includes the proportion of the country's population who are obese, the proportion undernourished, and and the % of COVID-19 cases.

```
## Parsed with column specification:
## cols(
## .default = col_double(),
## Continent = col_character(),
## Country = col_character(),
## Undernourished = col_character(),
## 'Unit (all except Population)' = col_character()
## )
```

See spec(...) for full column specifications.

head(food_dat)

##	#	A tibble:	6 x 33				
##		Continent	Country	'Alcoholic	Bever~	'Animal fats'	'Animal Product~
##		<chr></chr>	<chr></chr>		<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
##	1	Asia	Afghanistan		0.0014	0.197	9.43
##	2	Europe	Albania		1.67	0.136	18.8
##	3	Africa	Algeria		0.271	0.0282	9.63
##	4	Africa	Angola		5.81	0.056	4.93
##	5	N America	Antigua and Barbuda		3.58	0.0087	16.7
##	6	S America	Argentina		4.27	0.223	19.3
##	#	with 2	28 more variables: Aq	uatic Produ	icts, Ot	cher <dbl>,</dbl>	
##	#	Cereals	- Excluding Beer <db< th=""><th>ol>, Eggs <d< th=""><th>lbl>, Fi</th><th>lsh, Seafood <d< th=""><th>lbl>,</th></d<></th></d<></th></db<>	ol>, Eggs <d< th=""><th>lbl>, Fi</th><th>lsh, Seafood <d< th=""><th>lbl>,</th></d<></th></d<>	lbl>, Fi	lsh, Seafood <d< th=""><th>lbl>,</th></d<>	lbl>,
##	#	Fruits -	- Excluding Wine <dbl< th=""><th>>, Meat <db< th=""><th>ol>, Mil</th><th>lk - Excluding</th><th>Butter <dbl>,</dbl></th></db<></th></dbl<>	>, Meat <db< th=""><th>ol>, Mil</th><th>lk - Excluding</th><th>Butter <dbl>,</dbl></th></db<>	ol>, Mil	lk - Excluding	Butter <dbl>,</dbl>
##	#	Miscella	aneous <dbl>, Offals</dbl>	<dbl>, Oilc</dbl>	rops <ċ	lbl>, Pulses <d< th=""><th>lbl>,</th></d<>	lbl>,
##	#	Spices <	dbl>, Starchy Roots	<dbl>, Stim</dbl>	ulants	<dbl>,</dbl>	
##	#	Sugar &	Sweeteners <dbl>, Su</dbl>	ıgar Crops <	dbl>, 1	<pre>Ireenuts <dbl>,</dbl></pre>	,
##	#	Vegetabl	le Oils <dbl>, Vegeta</dbl>	ables <dbl>,</dbl>	Vegeta	al Products <db< th=""><th>ol>,</th></db<>	ol>,

Use this histogram to answer parts a-d.





5

7. i) [1 point] Describe the distribution.

SOLUTION: unimodal, skewed to the right

- 7. ii) [0.5 point] Pick the sentence that is most correct
- (a) The mean is approximately equal than 5
- (b) The mean is larger than 5
- (c) The mean is smaller than 5
- (d) Not enough information to choose

SOLUTION: (b) The mean is larger than 5

- 7. iii) [1 point] Select all true statements based on the histogram and knowledge you've gained in this class.
- (a) mean = median
- (b) mean > median
- (c) mean < median
- (d) The mean is resistant to outliers.
- (e) The median is resistant to outliers.

SOLUTION: (b) mean > median, (e) The median is resistant to outliers.

7. iv) [1 point] What is the binwidth for this distribution?

SOLUTION: 2.5

8. [2.5 points total] The data set named diabt contains information about diabetic and non-diabetic patients. In particular, the variable diabetes equals 0 for individuals without diabetes, equals 1 for individuals with type 1 diabetes and equals 2 for individuals with type 2 diabetes.

Here is some information about these data:

dim(diabt)

[1] 18 6

head(diabt)

##	#	A tibb]	Le: 6 x 6					
##		nameid	height_cm	weight_kg	sex	race	diabetes	
##		<chr></chr>	<dbl></dbl>	<dbl></dbl>	<fct></fct>	<chr></chr>	<dbl></dbl>	
##	1	ADF	160	75	1	white	2	
##	2	PUD	186	78	1	white	0	
##	3	HYD	155	49	1	blakc	0	
##	4	RFD	150	64	1	blakc	1	
##	5	UDF	172	72	1	white	0	
##	6	USR	174	123	1	blakc	1	

8. i) [0.5 points] What type of variable is diabetes? Choose the best answer.

- (a) continuous
- (b) discrete
- (c) categorical
- (d) ordinal

SOLUTION: (c) categorical

8. ii) [2 points] Write code to make a chart (histogram or bar) for the distribution of the types of diabetes, where there is a separately colored bar for men and women; these bars are next to each other, within each type of diabetes.

SOLUTION:

```
ggplot(diabt, aes(x=diabetes)) + geom_bar(aes(fill = sex), position = "dodge")
```

8. iii) [2 points] The formula to calculate BMI is $\frac{weight(kg)}{height^2(m)}$. Add a variable named **bmi** to the **diabt** data frame.

SOLUTION: diabt <- diabt %>% mutate(bmi = weight_kg / (height_cm / 100)^2)

9. [0.5 points] True or False: Correlations for average measures are usually stronger than correlations based on individual data.

(a) True

(b) False

SOLUTION: (a) True



10. [0.5 points] The Pearson's correlation coefficient for this graph is likely close to:

- a) 0.2
- b) 0.4
- c) 0.6d) 0.8
- e) You should not calculate Pearson's correlation for this relationship

SOLUTION: e) You should not calculate Pearson's correlation for this relationship

11. [6 points total] You are interested in visualizing the relationship between the number of stem cell divisions and one's lifetime risk of different types of cancer. To investigate, you have a dataset called cancer_data, with a row of data for each of various types of cancers:

head(cancer_data)

##	#	A tibble: 6 x 3		
##		disease	lifetime_risk	<pre>stem_cell_divisions</pre>
##		<chr></chr>	<dbl></dbl>	<dbl></dbl>
##	1	AM leukemia	0.0041	130.
##	2	Basal Cell	0.3	3550
##	3	CL Leukemia	0.0052	130.
##	4	Colorectal	0.048	1168
##	5	FAP Colorectal	1	1168
##	6	Lynch Colorectal	0.5	1168

You create a scatterplot of lifetime_risk versus stem_cell_divisions:



Relationship between stem cell divisions and lifetime risk of d

11. i) [2 points] In 1-3 *brief* sentences max, how would you describe these data? Would you want to use a linear model to summarize this relationship?

SOLUTION: Most of the points are clumped in the bottom left corner. You would not want to use a linear model.

You decide to transform your data and make a plot of the transformed relationship as shown below.



Transformed relationship between stem cell divisions

ii) [3 points] What transformation did you likely perform on the the explanatory and/or the response 11. variables to produce the second scatterplot? To make sure you picked the correct function, apply the transformation to your best guess of the x and y values for the point labeled A in the first plot and show that it roughly corresponds to a point on the second plot.

SOLUTION: Log transformed both x and y. Take the datapoint at 1100 (approximately) and $0.5. \log(1100) = 7, \log(0.5) = -0.69.$

11. iii) [1 point] A classmate says that according to this plot, it is clear that the number of stem cell divisions directly affects the lifetime risk of cancer. What is one concept you learned about in class that provides an alternate explanation for the linear relationship between these variables?

SOLUTION: Confounding or lurking variable.

```
## Parsed with column specification:
## cols(
##
     MCV = col_double(),
##
     ALKPHOS = col_double(),
     SGPT = col_double(),
##
##
     SGOT = col_double(),
##
     GAMMAGT = col double(),
     DRINKS = col_double(),
##
##
     disorder = col_double()
## )
```

12. [8 points total] The dataset bupa contains information about liver disorders. It contains data on 345 individuals' blood test results and liver disorder status. The following table shows the first six rows of bupa. The variables SGPT and GAMMAGT are both measurements of the patients' liver condition in the unit IU/L.

##	#	A tibl	ole: 6 x	7				
##		MCV	ALKPHOS	SGPT	SGOT	GAMMAGT	DRINKS	disorder
##		<dbl></dbl>						
##	1	85	92	45	27	31	0	1
##	2	85	64	59	32	23	0	2
##	3	86	54	33	16	54	0	2
##	4	91	78	34	24	36	0	2
##	5	87	70	12	28	10	0	2
##	6	98	55	13	17	17	0	2

12. i) [2 points] You made a scatter plot of SGPT vs. GAMMAGT and based on the plot, decide it might be better to build a linear regression model using the natural log transformed variables log_SGPT and log_GAMMAGT:

log(SGPT) = a + b * log(GAMMAGT). Write code that adds two new variables to bupa, and fits the linear model (saved as bupa_model).

SOLUTION:

bupa <- bupa %>% mutate(log_SGPT = log(SGPT), log_GAMMAGT = log(GAMMAGT))
bupa_model <- lm(log_SGPT ~ log_GAMMAGT, data = bupa)</pre>

12. ii) [1 point] The summary of the fitted linear model is:

##	#	A tibble: 2	x 5			
##		term	estimate	std.error	statistic	p.value
##		<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
##	1	(Intercept)	2.06	0.0986	20.9	6.05e-63
##	2	log_GAMMAGT	0.369	0.0290	12.7	1.50e-30

Interpret the slope parameter.

SOLUTION: A one unit increase in the logarithm of GAMMAPT(IU/L) is associated with an increase of 0.369 in the logarithm of SGPT(IU/L).

12. iii) [1 point] Here is some more R output:

glance(bupa_model)

A tibble: 1 x 12 ## r.squared adj.r.squared sigma statistic p.value AIC BIC df logLik ## <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> ## 1 0.320 0.318 0.420 161. 1.50e-30 1 -189. 384. 396. ## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>

Interpret the r-squared value. Be specific.

SOLUTION: r_squared = 0.320 means that 32% of the variation in log_SGPT can be explained by the variations in log_GAMMAGT.

12. iv) [2 points] Recall that you applied the log transformation on both variables. How would the r-squared value for the relationship between SGPT and GAMMAGT compare to the r-squared value for the relationship between log_SGPT and log_GAMMAGT and why? For reference, here is the scatter plot based on the transformed data:



SOLUTION: The original plot didn't look linear, so if we calculated the r-squared value for the non-linear relationship, we suspect it would be lower than for this clearly linear relationship.

12. v) [2 points] Explain why it is not a good idea to make a prediction of SGPT given GAMMAGT=5000 using the current data and model. Provide a calculation based on any information provided above to support your reasoning.

SOLUTION: Extrapolation. $\log(5000) = 8.52$, while the maximum value of $\log_GAMMAGT$ in the current dataset is 5.69. It is far from the bulk of the data and cannot be accurately predicted.

13. [1 point] You have a dataset called **diet** that contains information on diet and the incidence of coronary heart disease (CHD) of individuals.

For reference, the variables in this dataset include: id: subject identifier, numeric job: occupation, that can take the values Driver, Conductor, and Bank worker energy: total energy intake (kCal per day/100), numeric height: in cm, numeric weight: in kg, numeric fat: fat intake (g), numeric

chd: CHD event, where the value 1 implies this individuals has had a CHD event, and 0 implies this individuals has had no CFD event

Write one line of code to create a new data frame called diet_subset, which only contains individuals who are drivers and have fat intakes larger than 100g.:

SOLUTION: diet_subset <- diet %>% filter(job == "Driver", fat > 100)

14. [6 points total] The following data looks at the relationship between endometriosis and hypertension. A third variable included in this analysis is the genotype each woman has of a particular gene. The three levels are GG, GT, and TT.

##	#	A tibble: 6	x 6				
##		endo_status	genotype	count	count_with_ht	genotype_prop	$percent_ht$
##		<chr></chr>	<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
##	1	Endo	GG	55	24	0.056	43.6
##	2	Endo	GT	768	344	0.784	44.8
##	3	Endo	TT	156	62	0.159	39.7
##	4	No Endo	GG	2401	1121	0.261	46.7
##	5	No Endo	GT	4393	2028	0.478	46.2
##	6	No Endo	TT	2395	1007	0.261	42.0

14. i) [1 point] Using the data, fill in the blanks of the following two-way table.

	Hypertension	No Hypertension	Total
Endo	430	А	979
No Endo	В	\mathbf{C}	9189
			<u> </u>
Total	4418	D	10168

SOLUTION:

A: 549

B: 3988

C: 5201

D: 5750

14. ii) [1 point] What is the marginal distribution of endometriosis in this population? Round your answer to 2 decimal places.

SOLUTION: The percentage of women who have endometriosis is 979/10168 is 9.63%. The percentage of women who do not have endometriosis is 90.37%.

14. iii) [1 point] What is the conditional distribution of hypertension among women with endometriosis?

SOLUTION: The percentage of women who have hypertension given they have endometriosis is 43.92%. The percentage of women who do not have hypertension given they have endometriosis is 56.08%.

- 15. iv) [1 point] Which group has the highest overall rate of hypertension?
- (a) Endo
- (b) No endo

SOLUTION: (a) Endo has the highest overall rate of hypertension.



attr(,"class") ## [1] "labels"

14. v) [2 points] From the visualization above, it is evident that within each genotype, there is a higher incidence of hypertension in the group without endo than the group with endo. In 1-3 *brief* sentences and using your answer in Part E, identify the cause of this phenomenon, and explain why that is the cause. Hint: Look at the variable genotype_prop in your dataframe.

SOLUTION: Overall, there is a higher incidence of hypertension in the endo group but, within each strata of genotype, there is a higher incidence of hypertension in the non-endo group. This is because genotype is a confounding variable for the relationship between endo and hypertension. There is a different distribution of genotype among women with endo and women without leading us to believe that genotype has an effect on both endo and hypertension.