Kaggle Master-1-Final Test

Total points 100/100



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Sevdanur GENÇ

✓ Q1- Assume you have a dataset named museum_data and its index is

Date column. When you run "museum_data.head()" statement you get
the following://lmage in Cell D4Which code below achieves the following
requirement?"In October 2018, how many more visitors did Avila Adobe
receive than the Firehouse Museum?"Note: All dates are in the format of:
YYYY-MM-01 (Day is always 01 for every year and month) *

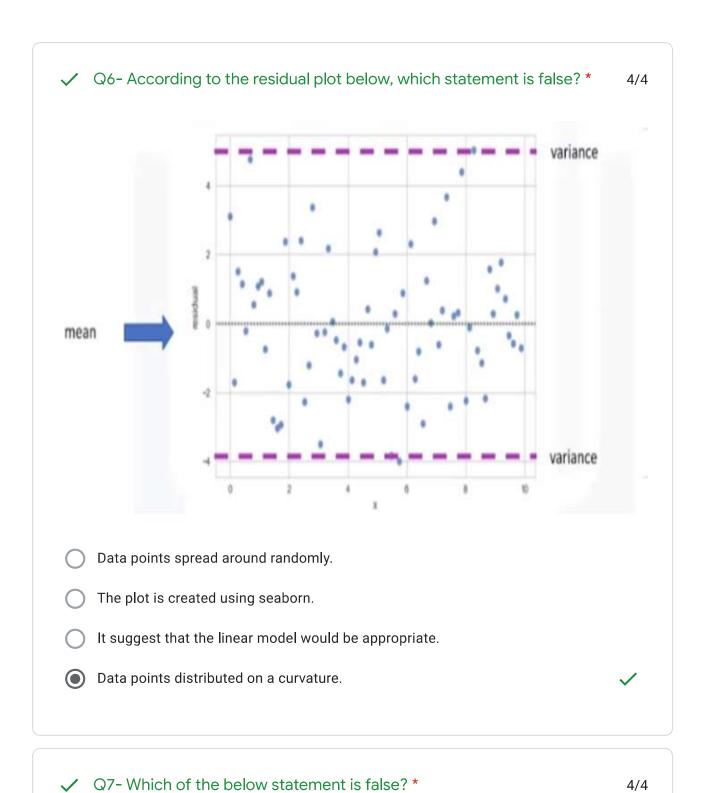
	Avila Adobe	Firehouse Museum	Chinese American Museum	America Tropical Interpretive Center
Date				
2014-01-01	24778	4485	1581	6602
2014-02-01	18976	4172	1785	5029
2014-03-01	25231	7082	3229	8129
2014-04-01	26989	6756	2129	2824
2014-05-01	36883	10858	3676	10894

- museum_data[museum_data.index.isin(['2018-10-01'])]['Avila Adobe'].sum() museum_data[museum_data.index.isin(['2018-10-01'])]['Firehouse
 Museum'].sum()
- museum_data[museum_data.index.isin(['2018-10-01'])]['Firehouse Museum'].sum() museum_data[museum_data.index.isin(['2018-10-01'])]['Avila Adobe'].sum()

pupoum data[muooum data index ioin(['2010 10 01'])]['Eirahouge Museum'] cum()

museum_data[museum_data.index.isin(['2018-10-01'])]['Chinese American Museum'].sum()	V
museum_data[museum_data.index.isin(['2019-10-01'])]['Firehouse Museum'].sum museum_data[museum_data.index.isin(['2020-10-01'])]['Avila Adobe'].sum()	() -
Q2- What is the aim of the below code pieces? *	4/4
<pre>from sklearn.metrics import mean_absolute_error</pre>	
<pre>predicted_home_prices = melbourne_model.predict(X)</pre>	
<pre>mean_absolute_error(y, predicted_home_prices)</pre>	
For splitting the data as test and train	
For data modelling	
For interpreting the data description	
For summarizing model quality	~
Q3- Which one is false about overfitting and underfitting? *	4/4
Training on too much epoch and batch size causes overfitting.	
Splitting dataset as train and test datasets will always be enough to prevent overfitting, no need for validation datasets.	✓
Insufficient training (less epoch less batch size), causes underfitting.	
In overfitting accuracy will be very good at train data but will be very bad at unseed data.	en

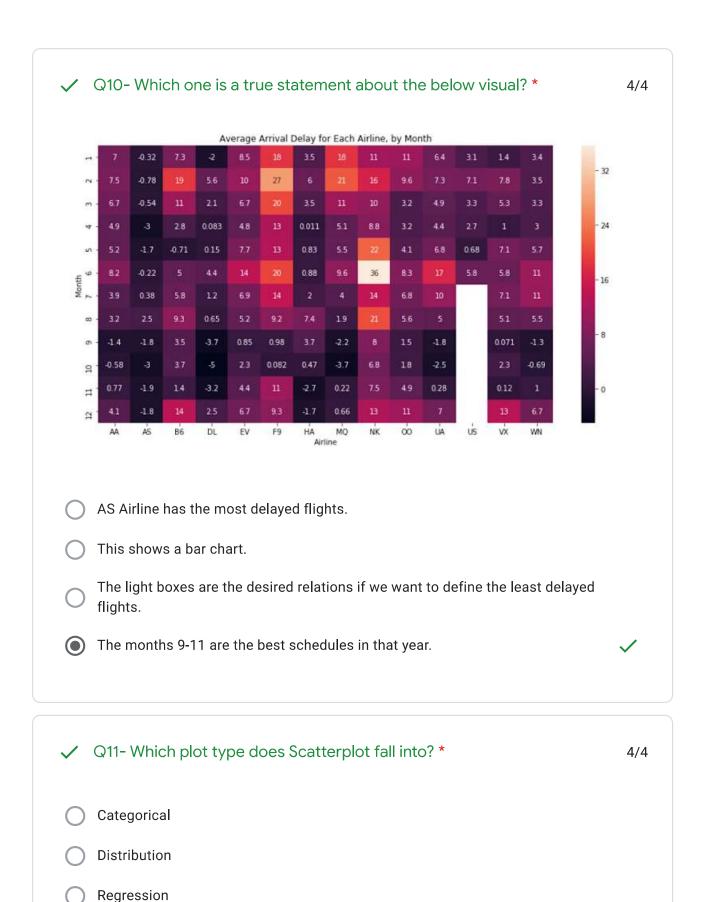
```
Q4- What do the highlighted code pieces mean? *
                                                                                  4/4
X_train_plus = X_train.copy()
X valid plus = X valid.copy()
for col in cols with missing:
    X_train_plus[col + '_was_missing'] = X_train_plus[col].isnull()
    X_valid_plus[col + '_was_missing'] = X_valid_plus[col].isnull()
my imputer = SimpleImputer()
imputed X train plus = pd.DataFrame(my imputer.fit transform(X train plus))
imputed X valid_plus = pd.DataFrame(my_imputer.transform(X_valid_plus))
imputed X_train_plus.columns = X_train_plus.columns
imputed X valid plus.columns = X valid plus.columns
     To make copy to avoid changing original data
     For imputation
 To put removed column names back
     To make new columns indicating what will be imputed
✓ Q5- Which of the following is not an approach that we can use to
                                                                                  4/4
     prepare our categorical data? *
     One-Hot Encoding
     Normalization
     Label Encoding
     Drop categorical values
```



sns.distplot - Show the distribution of a single numerical variable

sns.heatmap - Used to find color-coded patterns in tables of numbers

sns.Implot - Useful for drawing multiple regression lines, if the scatter plot contains multiple, color-coded groups. V Q8- Which of the following statements are true about "max_depth" 4/4 hyperparameter in Random Forest? * Lower is better parameter in case of same validation accuracy Higher is better parameter in case of same validation accuracy Increase the value of max_depth may overfit the data Increase the value of max_depth may underfit the data II, III I, IV II, IV Q9- Let assume, we have a data set called home_data with 3 features names; LotArea, YearBuilt, PoolArea. How do you define non-missing values for the feature LotArea? * non_missings = home_data['LotArea'].count() non_missings = home_data('LotArea'].mean() non_missings = home_data('LotArea').mean() non_missings = home_data('LotArea').mean()	•	groups	~
hyperparameter in Random Forest? * I- Lower is better parameter in case of same validation accuracy Higher is better parameter in case of same validation accuracy III- Increase the value of max_depth may overfit the data IV- IV- II, III	0		ns
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III- III- Increase the value of max_depth may overfit the data Increase the value of max_depth may underfit the data II, III II, III I, IV II, IV V Q9- Let assume, we have a data set called home_data with 3 features names; LotArea, YearBuilt, PoolArea. How do you define non-missing values for the feature LotArea? * non_missings = home_data["LotArea"].count() non_missings = home_data["LotArea"].mean()		Tryperparameter in Nariaon in orest.	
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I, IV II, IV Q9- Let assume, we have a data set called home_data with 3 features names; LotArea, YearBuilt, PoolArea. How do you define non-missing values for the feature LotArea? * non_missings = home_data["LotArea"].count() non_missings = home_data["LotArea"].mean()	0	II, III	
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non_missings = home_data["LotArea"].mean()	•	non_missings = home_data["LotArea"].count()	✓
	0	non_missings = home_data.count()	
<pre>non_missings = home_data.mean()</pre>	0	non_missings = home_data["LotArea"].mean()	
	0	non_missings = home_data.mean()	



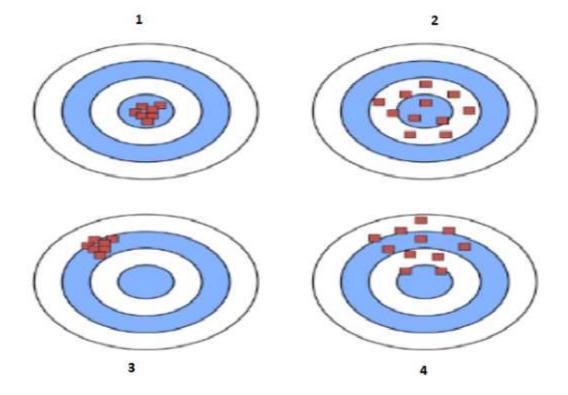
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	Q12-You will build a model to predict housing prices. The model will be deployed on an ongoing basis, to predict the price of a new house when a description is added to a website. Here are four features that could be
	used as predictors. Which of the features is most likely to be a source of leakage? *
0	Whether the house has a basement
0	Latitude and longitude of the house
•	Average sales price of homes in the same neighborhood
0	Size of the house (in square meters)
/	Q13- How is the Gradient Boosting cycle proceed? Please choose the correct order from the mixed statements below. *
- - 	o .
0	11-1-111
•	II-III-I
	I-II-III
0	

Relational

Q14- What do you think about train_X when line 1 and line 2 are executed 4/4 separately? The rest of the code is exactly the same. *
Line 1. train_X, val_X, train_y, val_y = train_test_split(X, y, random_state = 2,shuffle=False) Line 2. train_X, val_X, train_y, val_y = train_test_split(X, y, random_state = 1,shuffle=False)
They generate different random number ,but the train_X is equal to each other.
They generate different random number so the train_X is equal to each other.
They generate different random number so the train_X differs from each other.
They generate different same number and the train_X is equal to each other.

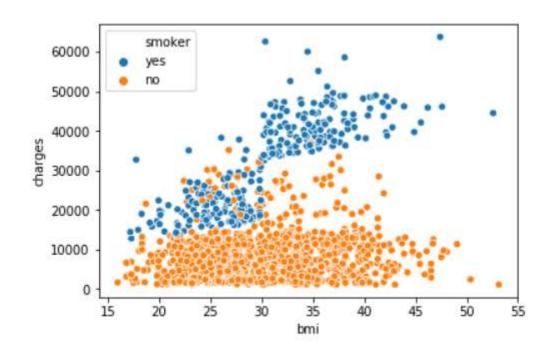
✓ Q15- According to the shooting clusters scheme above, for each figure 4/4 which statements are true? Notice that, shooting targets are the centers.



- 1:High Bias- Low Variance 2: High Bias-High Variance 3:Low Bias-Low Variance 4:Low Bias-High Variance
- 1:High Bias- High Variance 2:High Bias-Low Variance 3:Low Bias-High Variance 4:Low Bias-Low Variance
- 1:Low Bias- High Variance 2:Low Bias-Low Variance 3:High Bias-High Variance 4: High Bias-Low Variance
- 1:Low Bias- Low Variance 2:Low Bias-High Variance 3:High Bias-Low Variance 4:
 High Bias-High Variance

- Q16- Which of the statements below is incorrect for ensemble learning 4/4 and its techniques? *
- Its techniques use a combination of learning algorithms to optimize better predictive performance.
- It makes the model more robust.
- Typically, It reduces overfitting in the data.
- Typically, it reduces underfitting in the data.

✓ Q17- Which one is the correct option for the below visualization? * 4/4



- sns.lmplot(x="bmi", y="charges", hue="smoker", data=insurance_data)
- sns.scatterplot(x=insurance_data['bmi'], y=insurance_data['charges'])
 - sns.scatterplot(x=insurance_data['bmi'], y=insurance_data['charges'],

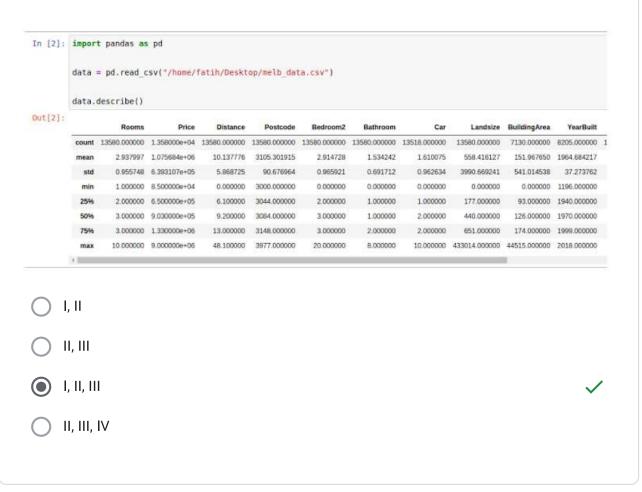
sns.regplot(x=insurance_data['bmi'], y=insurance_data['charges'])	
✓ Q18- Which of the below is/are nominal variable(s)? *	4/4
I - Gender	
II - Genotype	
III - Religious preference	
IV- IQ	
V - Income earned in a week.	
O I, II	
I, II, III	~
II, III, IV	
O I, III, IV	
Q19- What is the function of parameter ci? *	4/4
Defining the plot type	
To make subplot	
To highlight the classes of data points	
	,
Defining to show confidence interval	~

hue=insurance_data['smoker'])

✓	Q20- Imagine that you have a company and you would like to create a plot which shows the sales based on date. Which one of the following plot function is less likely suitable for this job? *	4/4
0	sns.lineplot	
•	sns.heatmap .	/
0	sns.barplot	
0	sns.scatterplot	
✓	Q21- Which of the following statement is inconsistent with pipelines? *	4/4
0	You won't need to manually keep track of your training and validation data at each step with a pipeline.	
0	With a pipeline, we can use the cross-validation technique easily.	
0	With pipelines, there is less probability to forget a preprocessing step.	
•	It's hard to productionize a model with pipelines.	/

Q22- Which of the following statements are true ab of cross-validation? *	out the intended use 4/4
I - To reduce randomness while measuring model pe	erformance.
II - To get a better measure of model performance.	
III - To increase model's training performance.	
IV - To increase MAE (mean absolute error) or MSE (mean squared error).
I, II	✓
O II, III	
O II, IV	
O I, IV	

✓ Q23- Which of the below can be said definitely according to the results 4/4 table taken from the data.describe() method? I. 75% of the values in the Rooms column are greater than 2. II. There are some houses with a land size of 0. III. There are missing values in the BuildingArea column. IV. There is no house with 9 rooms in the data set *



I-They help us to deal with categorical values. II-Label Encoding assigns each value to a different integer whether it is unique or not. III-One Hot Encoding creates new column for every possible value in the original data. IV-For large number of categorical variable count value (such as 15 different values) it is not good to use One Hot Encoder generally. I, II, III II, III, IV All of them Q25- Which statement is not true with the following code? museum_data = pd.read_csv(museum_filepath,index_col="Date",parse_dates=True) * If we did not use parse_dates, the type of the Date column would not change to datetime. (Assuming that the original type of the column is not datetime) Above code means that creating a new Date column that is not in csv and this column is defined as the index of the museum_data. When parse_dates = True, the type of Date column in museum_data becomes datetime The index value of the museum_data is the Date column in the csv file.	Q24- Which of the following statements are true about LabelEncoder and OneHotEncoder? *	4/4
II, III, IV I, III, IV All of them General Casv(museum_filepath,index_col="Date",parse_dates=True) * If we did not use parse_dates, the type of the Date column would not change to datetime. (Assuming that the original type of the column is not datetime) Above code means that creating a new Date column that is not in csv and this column is defined as the index of the museum_data. When parse_dates = True, the type of Date column in museum_data becomes datetime	II-Label Encoding assigns each value to a different integer whether it is unique or not. III-One Hot Encoding creates new column for every possible value in the original data. IV-For large number of categorical variable count value (such as 15 different values) it is not good to us	
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	museum_data =	4/4

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