

**BMAN73701 Programming in Python for Business Analytics 2023-24 1st Semester**[Course Content](#) [Mock Tests 2023](#) [Review Test Submission: Mock test 2023/24 - Second assessed online test](#)

## Review Test Submission: Mock test 2023/24 - Second assessed online test

User	Rakshit Yadav
Course	BMAN73701 Programming in Python for Business Analytics 2023-24 1st Semester
Test	Mock test 2023/24 - Second assessed online test
Started	10/11/23 04:17
Submitted	30/11/23 06:22
Status	Completed
Attempt Score	Mark not available.
Time Elapsed	482 hours, 5 minutes
Results Displayed	All Answers, Submitted Answers, Correct Answers, Feedback

**Question 1**

9.0909 out of 10 points

Let us assume 'sales' is a Numpy matrix that contains amount sold per department (rows) per year (columns).

Fill in the missing blanks so that each line of code matches the comment the precedes the line.

Every input to a blank is exactly one word, symbol or number. NO spaces should be used (either before or after a word/symbol).

```
# 1. Calculate total number of sales of each department
```

```
total = np.[sum](sales, axis=[axis1])
```

```
# 2. Find the store with the largest total sales
```

```
best_store = np.[argmax](total)
```

```
# 3. Find if any store in any year did not have sales :
```

← OK

```
np.[all](sales > 0)
```

```
# 4. Find which stores had at least one year without sales:
```


```
np.[any](sales == 0, axis=[axis_1])
```


```
# 5. Find the largest sale amounts per year:
```

```
np.[max](sales, axis=[axis_0])
```


```
# 6. For each store, calculate its maximum sales of all  
years, then find index of the store with the smallest  
maximum.
```


```
np.[argmin](np.[max2](sales, axis=[axis_1]))
```


Specified Answer for: sum  sum


Specified Answer for: axis1  1


Specified Answer for: argmax  max


Specified Answer for: all  all


Specified Answer for: any  any


Specified Answer for: axis\_1  1

Specified Answer for: max  max


Specified Answer for: axis\_0  0

Specified Answer for: argmin  argmin

Specified Answer for: max2  max

Specified Answer for: axis\_11  1

#### Correct Answers for: sum

Evaluation Method	Correct Answer	Case Sensitivity
 Exact Match	sum	

#### Correct Answers for: axis1

Evaluation Method	Correct Answer	Case Sensitivity
 Exact Match	1	

#### Correct Answers for: argmax

Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	argmax	
<b>Correct Answers for: all</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	all	Case Sensitive
<b>Correct Answers for: any</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	any	Case Sensitive
<b>Correct Answers for: axis_1</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	1	Case Sensitive
<b>Correct Answers for: max</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	max	Case Sensitive
<b>Correct Answers for: axis_0</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	0	Case Sensitive
<b>Correct Answers for: argmin</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	argmin	Case Sensitive
<b>Correct Answers for: max2</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	max	Case Sensitive
<b>Correct Answers for: axis_11</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	1	

## Question 2

10 out of 10 points

Assuming that 'A' is a Numpy matrix with 1000 rows and 500 columns, match each line of code with the most appropriate description.

Question	Correct Match	Selected Match
Select all rows and the first column	✔ 3. <code>A[:, :1]</code>	✔ 3. <code>A[:, :1]</code>
Select all columns and the first row	✔ 2. <code>A[:1, :]</code>	✔ 2. <code>A[:1, :]</code>

Select all columns and the last row

✔ 4.

`A[-1:, 0:]`

✔ 4.

`A[-1:, 0:]`

Select all rows and the last column

✔ 5.

`A[:-1, -1:]`

✔ 5.

`A[:-1, -1:]`Select all columns except the last one  
and all rows

✔ 1.

`A[:-1, :-1]`

✔ 1.

`A[:-1, :-1]`

All Answer Choices

1. `A[:-1, :-1]`2. `A[1, :]`3. `A[:, 1]`4. `A[-1:, 0:]`5. `A[:-1, -1:]`

### Question 3

10 out of 10 points

You are given a Numpy matrix P that contains N rows and M columns. Each row of the matrix P represents one data point (i.e., you have N data points) and each column one numerical feature (i.e., you have M numerical features). You wish to apply Standardization (Z-score normalization) to this data, however, you missed the lecture about Data Preprocessing and you don't know of any function to do this. You decide to implement it yourself only using Numpy. To perform Standardization, you need to do the following steps:

1. For all values of each feature  $X$ , that is,  $X = \{x_1, x_2, \dots, x_N\}$ , you need to do the following operation:

$$x'_i = \frac{x_i - \bar{X}}{\sigma_X}$$

where  $\bar{X}$  and  $\sigma_X$  are, respectively, the mean and standard deviation of all

values of  $X$ . This transformation must be applied to each of the M features of the matrix.

2. If the standard deviation of a feature before preprocessing is less than 1e-5, that feature is constant for all purposes and it must be removed from the data.

Select the correct lines of Python for this task.

Selected



Answer:

```
tmp = P[:, P.std(axis=0) >= 1e-5]
P_norm = (tmp - tmp.mean(axis=0)) /
tmp.std(axis=0)
```

Answers:

```
tmp = (P - P.mean(axis=1)[:,np.newaxis]) /
P.std(axis=1)[:,np.newaxis]
P_norm = tmp[tmp.std(axis=1) >= 1e-5, :]
```

```
P_norm = (P[:, P.std(axis=0) >= 1e-5] -
P.mean(axis=0)) / P.std(axis=0)
```



```
tmp = P[:, P.std(axis=0) >= 1e-5]
P_norm = (tmp - tmp.mean(axis=0)) /
tmp.std(axis=0)
```

```
if P.std() >= 1e-5:
    P_norm = P - P.mean() / P.std()
```

```
P_norm = (P[P.std(axis=1) >= 1e-5, :] -
P.mean(axis=1)[:,np.newaxis]) / (P.std(axis=1)
[:,np.newaxis])
```

```
tmp = P[:, P.std(axis=0) < 1e-5]
P_norm = (tmp - tmp.mean(axis=0)) /
tmp.std(axis=0)
```

```
tmp = P[P.std(axis=1) < 1e-5, :]
P_norm = (tmp - tmp.mean(axis=1)[:,np.newaxis]) /
(tmp.std(axis=1)[:,np.newaxis])
```

```
tmp = (P - P.mean(axis=0)) / P.std(axis=0)
P_norm = tmp[:, tmp.std(axis=0) >= 1e-5]
```











```
tmp = P[P.std() >= 1e-5]
P_norm = (tmp - tmp.mean(tmp)) / tmp.std()
```

```
tmp = P[P.std(axis=1) >= 1e-5, :]
P_norm = (tmp - tmp.mean(axis=1)[:,np.newaxis]) /
(tmp.std(axis=1)[:,np.newaxis])
```

#### Question 4

4 out of 10 points

Assuming that x is a numpy vector of shape (10,) and A is a numpy matrix of shape (10,5), match each erroneous expression with the correct explanation of the error.

Question	Correct Match	Selected Match
<code>A + x</code>	 2. Gives an error because the number of columns of A is incompatible with the shape of x.	 1. There is no error. The code is correct.
<code>A.reshape(10, 5) + x.reshape(10, 5)</code>	 5. Gives an error because x does not have enough elements to fill the shape requested.	 6. Gives an error because x has only one dimension, i.e., x is a vector not a matrix.
<code>A.reshape(10, ) + x.reshape(10, 1)</code>	 3. Gives an error because A has too many elements to fill the shape requested.	 3. Gives an error because A has too many elements to fill the shape requested.
<code>A.reshape(5, 10) + x.reshape(10, 1)</code>	 4. Gives an error because the number of rows of A is incompatible with the shape of x.	 4. Gives an error because the number of rows of A is incompatible with the shape of x.
<code>A.sum(axis=1) + x.sum(axis=1)</code>	 6. Gives an error because x has only one dimension, i.e., x is a vector not a matrix.	 2. Gives an error because the number of columns of A is incompatible with the shape of x.

---

#### All Answer Choices

1. There is no error. The code is correct.
2.  
Gives an error because the number of columns of A is incompatible with the shape of x.
3.  
Gives an error because A has too many elements to fill the shape requested.
4.  
Gives an error because the number of rows of A is incompatible with the shape of x.
5.  
Gives an error because x does not have enough elements to fill the shape requested.

6.

Gives an error because x has only one dimension, i.e., x is a vector not a matrix.

## Question 5

7.14285 out of 10 points

Given the following two data frames:

Sales_Company_A		
"Date"	"Profit"	"Phone"
01-01-2021	2.2	"iPhone"
01-01-2021	2.5	"Android"
02-01-2021	2.4	"iPhone"
02-01-2021	2.9	"Android"

Sales\_Company\_B

"Date"	"iPhone"	"Android"
01-01-2021	2.2	2.5
02-01-2021	2.4	2.9
03-01-2021	2.5	3.1
04-01-2021	2.6	2.0

Fill the blanks of the following sentences:

- The shape of Sales\_Company\_B is **[wide]** because **[at\_least\_one\_var]** appears **[two\_columns]**.
- The dataset Sales\_Company\_**[A]** is tidy because it is **[long\_two]** and **[each\_obs]** appears **[different\_row]**.

Selected  
Answer:

Given the following two data frames:

Sales_Company_A		
"Date"	"Profit"	"Phone"
01-01-2021	2.2	"iPhone"
01-01-2021	2.5	"Android"
02-01-2021	2.4	"iPhone"
02-01-2021	2.9	"Android"

Sales\_Company\_B

"Date"	"iPhone"	"Android"
01-01-2021	2.2	2.5

02-01-2021	2.4	2.9
03-01-2021	2.5	3.1
04-01-2021	2.6	2.0

Fill the blanks of the following sentences:

- The shape of Sales\_Company\_B is **wide** because **at least one variable** appears **in two columns**.
- The dataset Sales\_Company\_ **A** is tidy because it is **long** and **each variable** appears **in a different column**.

Answers: Given the following two data frames:

Sales_Company_A		
<b>"Date"</b>	<b>"Profit"</b>	<b>"Phone"</b>
01-01-2021	2.2	"iPhone"
01-01-2021	2.5	"Android"
02-01-2021	2.4	"iPhone"
02-01-2021	2.9	"Android"

Sales\_Company\_B

<b>"Date"</b>	<b>"iPhone"</b>	<b>"Android"</b>
01-01-2021	2.2	2.5
02-01-2021	2.4	2.9
03-01-2021	2.5	3.1
04-01-2021	2.6	2.0

Fill the blanks of the following sentences:

- The shape of Sales\_Company\_B is **wide** because **at least one variable** appears **in two columns**.
- The dataset Sales\_Company\_ **A** is tidy because it is **long** and **each observation** appears **in a different row**.

#### All Answer Choices



- wrong
- tidy
- wide
- long
- each variable
- each observation
- in a different row
- in a different column
- at least one observation
- at least one variable
- in two columns
- in two rows
- A
- B

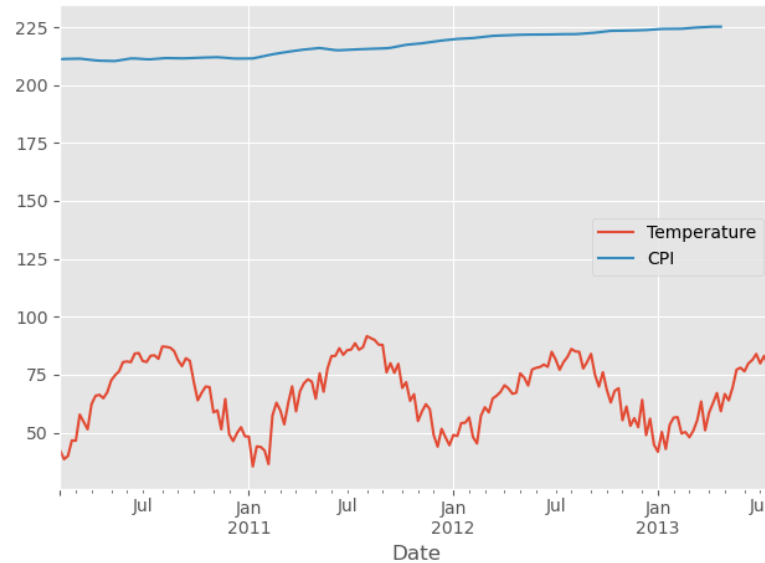
### Question 6

10 out of 10 points

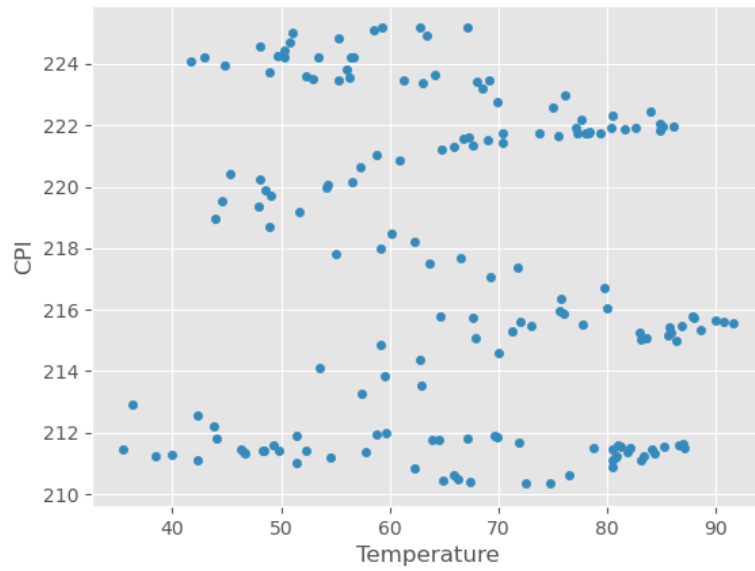
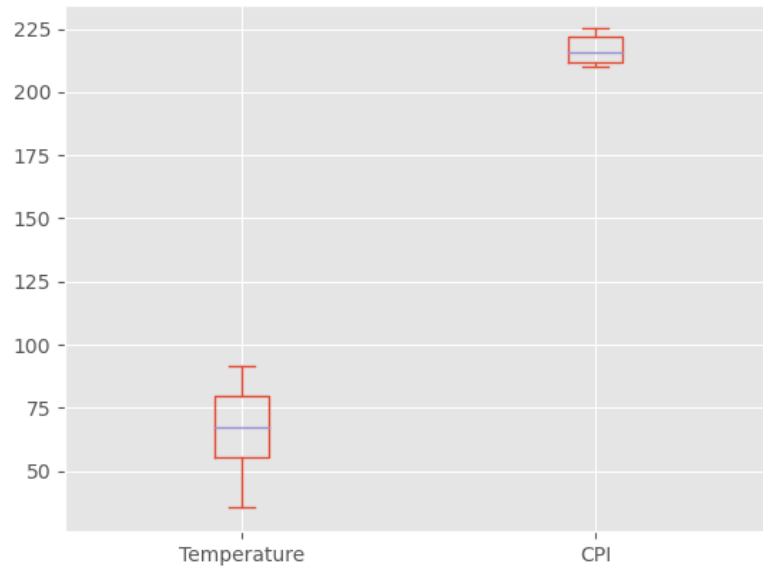
Select the plot that most closely matches the expected output of this line of Python code:

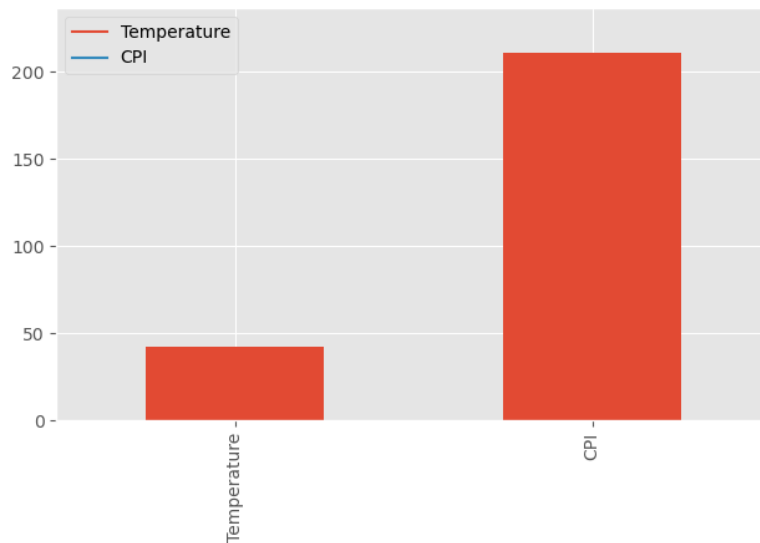
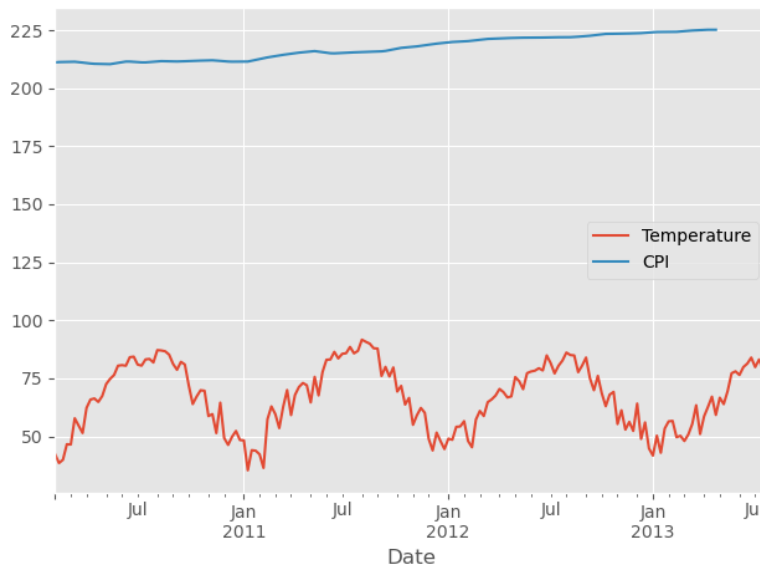
```
df[['Temperature', 'CPI']].plot()
```

Selected  
Answers:



Answers:





### Question 7

0 out of 10 points

Which of the following statements are true regarding DataFrames in Pandas?  
Mark all that apply. Incorrect answers penalize.

Selected



Answers:

Automatic index alignment between DataFrames only happens after we set the index of the DataFrames to the values of some column.



Setting the index of a DataFrame to a column changes the labels of the rows to the values of the column chosen.



When summing two columns of two different DataFrames, for example  
`df1['col1'] + df2['col2']`  
the result is a new column where each row is the sum of the values of the corresponding rows of each DataFrame.

Answers:



When summing two columns of two different DataFrames, for example  
`df1['col1'] + df2['col2']`  
the result depends on the index of each DataFrame.

Automatic index alignment between DataFrames only happens after we set the index of the DataFrames to the values of some column.



Setting the index of a DataFrame to a column changes the labels of the rows to the values of the column chosen.

When summing two columns of two different DataFrames, for example  
`df1['col1'] + df2['col2']`  
If one DataFrame has fewer rows than the other, then we will get missing values.

When summing two columns of two different DataFrames, for example  
`df1['col1'] + df2['col2']`  
the result is a new column where each row is the sum of the values of the corresponding rows of each DataFrame.

## Question 8

4 out of 10 points

Mark all the sentences that are true regarding matplotlib. Incorrect answers penalise.

Selected

Answers:



In matplotlib, when you place an axes within an axes you create a plot on top of another plot.



In matplotlib, a figure may contain several axes but axes do not contain figures.



In matplotlib, plots are objects that have their own methods (functions), which may be used to modify the plots.



If you want to create a large number of similar plots, it is better to use **plt.show()** after creating each plot to save it to a file.

Answers:



In matplotlib, when you place an axes within an axes you create a plot on top of another plot.

Plots created with Pandas and with Matplotlib are incompatible because they use different ways of plotting.



In matplotlib, a figure may contain several axes but axes do not contain figures.



Pandas uses matplotlib to create plots.

The Synder GUI allows creating more complex plots than matplotlib.

In matplotlib, when you place a figure within a figure you create a plot on top of another plot.

In matplotlib, a figure may contain several axes and an axes may contain several figures.



In matplotlib, plots are objects that have their own methods (functions), which may be used to modify the plots.



If you want to create a large number of similar plots, it is better to NOT use **plt.show()** after creating each plot.

If you want to create a large number of similar plots, it is better to use **plt.show()** after creating each plot to save it to a file.

## Question 9

8 out of 10 points

Assuming that 'X' is a Numpy matrix and 'df' is a Pandas DataFrame, fill the blanks in the following code to match the comments.

Every input to a blank is exactly one word, number or symbol. NO spaces should be used (either before or after a word/symbol).

# 1. Calculate the mean value of each column of X taking into account missing

values.

```
result = np.[mean](X, axis=[axis0])
```

# 2. Calculate the mean value of each column of X ignoring missing values.

```
result = np.[nanmean](X, axis=[axis0_])
```

# 3. Replace missing values in df with the value 1

```
df = df.[fillna](1)
```

# 4. Remove columns with missing values


```
df = df.[dropna](axis=[axis1_])
```


# 5. Remove rows with missing values


```
df = df.[dropna2](axis=[axis0_])
```


# 6. Calculate the mean value of each column of df ignoring missing values.


```
result = df.[mean2]()
```


Specified Answer for: mean  nanmean


Specified Answer for: axis0  0


Specified Answer for: nanmean  mean


Specified Answer for: axis0\_  0


Specified Answer for: fillna  fillna

Specified Answer for: dropna  dropna


Specified Answer for: axis1\_  1

Specified Answer for: dropna2  dropna

Specified Answer for: axis0\_\_  0

Specified Answer for: mean2  mean

#### Correct Answers for: mean

Evaluation Method	Correct Answer	Case Sensitivity
 Exact Match	mean	

#### Correct Answers for: axis0

Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	0	
<b>Correct Answers for: nanmean</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	nanmean	
<b>Correct Answers for: axis0_</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	0	
<b>Correct Answers for: fillna</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	fillna	Case Sensitive
<b>Correct Answers for: dropna</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	dropna	Case Sensitive
<b>Correct Answers for: axis1__</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	1	
<b>Correct Answers for: dropna2</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	dropna	Case Sensitive
<b>Correct Answers for: axis0__</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	0	
<b>Correct Answers for: mean2</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	mean	

**Question 10**

8 out of 10 points

Fill the blanks in the following python code to match the comments.

Every input to a blank is exactly one word, number or symbol.

NO spaces should be used (either before or after a word/symbol).

Assume that any required **'import'** or preprocessing step has already been done earlier in the code.

**# 1. Split the data into 75% training and 25% test data sets**

**x\_train, x\_test, y\_train, y\_test = [train\_test\_split]**


```
(data.drop('answer' axis=1), data.answer, test_size = [.75])
```


```
# 2. Train a decision tree on the training data to predict
when the 'answer' is 0 or 1 (column 'answer' of the data).
```

```
dt = [DecisionTreeClassifier].[fit2] ([X_train], [y_train])
```


```
# 3. Calculate the scores for 5-fold cross-validation
```


```
scores = [cross_val_score](dt, [X_train2], [y_train2], cv = [five])
```


Specified Answer for: train\_test\_split  train\_test\_split


Specified Answer for: .75  0.25


Specified Answer for: DecisionTreeClassifier  DecisionTreeClassifier


Specified Answer for: fit2  fit

Specified Answer for: X\_train  X\_train

Specified Answer for: y\_train  y\_train


Specified Answer for: cross\_val\_score  crass\_val\_score

Specified Answer for: X\_train2  X\_train


Specified Answer for: y\_train2  y\_train

Specified Answer for: five  5


#### Correct Answers for: train\_test\_split

Evaluation Method	Correct Answer	Case Sensitivity
 <i>Exact Match</i>	train_test_split	Case Sensitive


#### Correct Answers for: .75

Evaluation Method	Correct Answer	Case Sensitivity
 <i>Contains</i>	.75	


#### Correct Answers for: DecisionTreeClassifier

Evaluation Method	Correct Answer	Case Sensitivity
 <i>Exact Match</i>	DecisionTreeClassifier	

#### Correct Answers for: fit2

Evaluation Method	Correct Answer	Case Sensitivity
 <i>Exact Match</i>	fit	Case Sensitive

#### Correct Answers for: X\_train

Evaluation Method	Correct Answer	Case Sensitivity
 <i>Exact Match</i>	X_train	

#### Correct Answers for: y\_train



Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	y_train	
<b>Correct Answers for: cross_val_score</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	cross_val_score	Case Sensitive
<b>Correct Answers for: X_train2</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	X_train	
<b>Correct Answers for: y_train2</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	y_train	
<b>Correct Answers for: five</b>		
Evaluation Method	Correct Answer	Case Sensitivity
✔ <i>Exact Match</i>	5	

Thursday, 30 November 2023 06:23:54 o'clock GMT