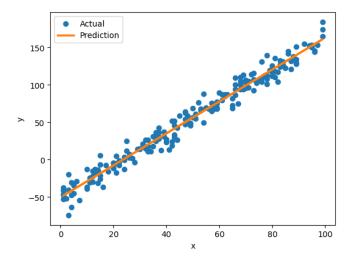
```
# Do NOT modify this block of code
import numpy as np
import numpy.typing as npt
import matplotlib.pyplot as plt
# Do NOT modify this block of code
n = 200
rand_x = np.random.RandomState(0)
x = rand_x.randint(1, 100, (n, 1))
rand_noise = np.random.RandomState(10)
y = 2.13 * x - 50.84 + 10*rand_noise.randn(n, 1)
\tt def \ fit\_regression\_line(x: \ npt.NDArray[np.float64], \ y: \ npt.NDArray[np.float64]) \ -> \ npt.NDArray[np.float64]: \ 
             ## Your code starts here ##
             #finding the mean for \boldsymbol{x} and \boldsymbol{y}
             x_{mean} = np.mean(x)
           y_mean = np.mean(y)
            beta_1 = np.sum((x - x_mean) * (y - y_mean)) / np.sum((x - x_mean)**2)
            beta_0 = y_mean - beta_1 * x_mean
            ## end ##
            return np.array([beta_0, beta_1])
# Do NOT modify this block of code.
# If this code cell run without any 'AssertionError', then your function 'fit_regression_line' is returning a numpy array with the correct shape.
assert fit_regression_line(x, y).shape == (2, )
\# Do NOT modify this block of code
def rmse(x: npt.NDArray[np.float64], y: npt.NDArray[np.float64]) -> np.float64:
            beta_0, beta_1 = fit_regression_line(x, y)
           y_hat = beta_1 * x + beta_0
             return np.sqrt(np.mean((y - y_hat)**2))
print(f"RMSE score is: {np.round(rmse(x, y), 4)}")
                RMSE score is: 9.7621
# Do NOT modify this block of code
\label{local_def} \mbox{def $r$\_squared(x: npt.NDArray[np.float64], y: npt.NDArray[np.float64]) $$\rightarrow$ np.float64: $$ \mbox{def $r$\_squared(x: npt.NDArray[np.float64], y: npt.NDArray[np.float64]) $$\rightarrow$ np.float64: $$ \mbox{def $r$\_squared(x: npt.NDArray[np.float64], y: npt.NDArray[np.float64]) $$\rightarrow$ np.float64: $$ \mbox{def $r$\_squared(x: npt.NDArray[np.float64], y: npt.NDArray[np.float64]) $$\rightarrow$ np.float64: $$\mbox{def $r$\_squared(x: npt.NDArray[np.float64], y: npt.NDArray[np.float6], y: npt.NDArray[np.float6], y: npt.NDArray[np.float6], y: npt.NDArray[np.float6], y: npt.NDArray
            beta_0, beta_1 = fit_regression_line(x, y)
            y_hat = beta_1 * x + beta_0
             return (np.corrcoef(y_hat, y, rowvar=False)[0, 1])**2
print(f"R-Squared score is: {np.round(r squared(x, y), 4)}")
                R-Squared score is: 0.9737
# Do NOT modify this block of code
\label{local_predictions} \texttt{def plot\_predictions}(\texttt{x: npt.NDArray[np.float64]}) \ \ \textbf{-> None:}
            beta_0, beta_1 = fit_regression_line(x, y)
            y_hat = beta_1 * x + beta_0
             plt.scatter(x.flatten(), y.flatten(), color=plt.cm.tab10(0), label='Actual')
            plt.plot(x.flatten(), y_hat.flatten(), color=plt.cm.tab10(1), linewidth=2.5, label='Prediction')
            plt.legend()
             plt.xlabel("x")
            plt.ylabel("y")
             plt.show()
plot_predictions(x, y)
```



Do NOT modify this block of code

import pandas as pd

Q2(a)

 $\mbox{\tt\#}$ Do NOT modify this block of code

import seaborn as sns

iris_df = sns.load_dataset('iris')
print(iris_df.shape)
iris_df.head()

(150, 5)

(, - ,						
	sepal_length	sepal_width	petal_length	petal_width	species	Ħ
0	5.1	3.5	1.4	0.2	setosa	ıl.
1	4.9	3.0	1.4	0.2	setosa	
2	4.7	3.2	1.3	0.2	setosa	
3	4.6	3.1	1.5	0.2	setosa	
4	5.0	3.6	1.4	0.2	setosa	

Your code starts here

#selecting the numeric column of data type float
numeric_columns = iris_df.select_dtypes(include=[float]).columns

#to iterate over the columns
for col in numeric_columns:

#calculating the mean
col_mean = iris_df[col].mean()

#calculating the standard deviation
col_sd = iris_df[col].std()

#calculating the z_score
iris_df[col + '_z'] = (iris_df[col] - col_mean) / col_sd

end

Q2(b)

Your code starts here

#creating a list of all columns with z-scores, that have the string '_z' z_score_col = [col for col in iris_df.columns if '_z' in col]

#calculating the mean of the z-scores, row-wise as shown using "axis=1"
iris_df['mean_z_score'] = iris_df[z_score_col].mean(axis=1)

end

√ Q2(c)

Your code starts here

###lambda: create small anonymous fn @runtime
#grouping the df using the species col
#finding abs then finding the max val of it
#sorting the data in the col in descending order (ascending=False)
max_abs_mean_z_score = iris_df.groupby('species')['mean_z_score'].apply(lambda x: x.abs().max()).sort_values(ascending=False)

end