STATISTICAL ANALYSIS In [1]: import pandas as pd In [18]: df=pd.read_csv("heart.csv") In [20]: df.head() Out[20]: age sex cp trtbps chol fbs restecg thalachh exng oldpeak slp caa thall output **0** 63 1 3 145 233 1 150 2.3 0 0 1 **1** 37 1 2 130 250 0 3.5 0 0 2 **2** 41 0 1 130 204 0 0 172 1.4 2 0 2 **3** 56 1 1 120 236 0 178 0.8 2 0 2 **4** 57 0 0 120 354 0 163 0.6 2 0 2 In [21]: df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 303 entries, 0 to 302 Data columns (total 14 columns): # Column Non-Null Count Dtype --- ----- -----0 age 303 non-null int64 1 sex 303 non-null int64 303 non-null int64 ср trtbps 303 non-null int64 303 non-null int64 chol fbs 303 non-null int64 restecg 303 non-null int64 thalachh 303 non-null int64 303 non-null int64 9 oldpeak 303 non-null float64 10 slp 303 non-null int64 11 caa 303 non-null int64 12 thall 303 non-null int64 13 output 303 non-null dtypes: float64(1), int64(13) memory usage: 33.3 KB **DESCRIPTIVE STATISTICS** In [3]: print("Mean:\n", df.mean()) print("\nMedian:\n", df.median()) print("\nMode:\n", df.mode().iloc[0]) print("\nStandard Deviation:\n", df.std()) print("\nVariance:\n", df.var()) print("\nRange:\n", df.max() - df.min()) print("\nSkewness:\n", df.skew()) print("\nKurtosis:\n", df.kurt()) Mean: 54.366337 age 0.683168 sex ср 0.966997 trtbps 131.623762 246.264026 chol fbs 0.148515 restecg 0.528053 thalachh 149.646865 0.326733 exng oldpeak 1.039604 1.399340 slp caa 0.729373 2.313531 thall 0.544554 output dtype: float64 Median: age 55.0 1.0 sex 1.0 ср 130.0 trtbps 240.0 chol fbs 0.0 restecg 1.0 153.0 thalachh 0.0 exng 0.8 oldpeak slp 1.0 0.0 caa 2.0 thall 1.0 output dtype: float64 Mode: 58.0 age 1.0 sex 0.0 ср 120.0 trtbps 197.0 chol fbs 0.0 1.0 restecg 162.0 thalachh exng 0.0 oldpeak 0.0 slp 2.0 0.0 caa 2.0 thall output 1.0 Name: 0, dtype: float64 Standard Deviation: 9.082101 age 0.466011 sex 1.032052 ср 17.538143 trtbps chol 51.830751 fbs 0.356198 restecg 0.525860 22.905161 thalachh 0.469794 exng oldpeak 1.161075 slp 0.616226 caa 1.022606 thall 0.612277 0.498835 output dtype: float64 Variance: 82.484558 age 0.217166 sex 1.065132 ср 307.586453 trtbps chol 2686.426748 0.126877 fbs restecg 0.276528 thalachh 524.646406 0.220707 oldpeak 1.348095 slp 0.379735 1.045724 caa 0.374883 thall 0.248836 output dtype: float64 Range: 48.0 age 1.0 sex 3.0 ср 106.0 trtbps chol 438.0 fbs 1.0 2.0 restecg 131.0 thalachh 1.0 exng oldpeak 6.2 2.0 slp 4.0 caa thall 3.0 output 1.0 dtype: float64 Skewness: -0.202463 age -0.791335 sex 0.484732 ср 0.713768 trtbps chol 1.143401 fbs 1.986652 0.162522 restecg thalachh -0.537410 0.742532 exng oldpeak 1.269720 -0.508316 slp 1.310422 caa -0.476722 thall -0.179821 output dtype: float64 Kurtosis: -0.542167 age -1.382961 sex -1.193071 ср trtbps 0.929054 4.505423 chol fbs 1.959678 restecg -1.362673 thalachh -0.061970 -1.458317 exng 1.575813 oldpeak slp -0.627521 0.839253 caa 0.297915 thall -1.980783 output dtype: float64 INFERENTIAL STATISTICS In [24]: **from** scipy **import** stats trtbps_values = df['trtbps'] population_mean = 130 t_stat, p_value = stats.ttest_1samp(trtbps_values, population_mean) print(f"T-Statistic: {t_stat}") print(f"P-Value: {p_value}") T-Statistic: 1.6116108638133604 P-Value: 0.10809121014912249 INTERPRETATION • t-statistic represents that sample mean is 1.61 standard errors below the population mean • p-value is larger than 0.05 so,we fail to reject the null hypothesis **CONFIDENCE INTERVAL** In [5]: **import** numpy **as** np from scipy import stats # Sample mean and standard error for TRTBPS sample_mean = np.mean(trtbps_values) standard_error = stats.sem(trtbps_values) # Compute 95% confidence interval for TRTBPS confidence_interval = stats.norm.interval(0.95, loc=sample_mean, scale=standard_error) print(f"95% Confidence Interval for TRTBPS: {confidence_interval}") 95% Confidence Interval for TRTBPS: (129.64902030398173, 133.59850444849354) **REGRESSION ANALYSIS** In [7]: import statsmodels.api as sm # Define independent variable (add constant for intercept) X = sm.add_constant(df['trtbps']) # Define dependent variable y = df['age'] # Fit linear regression model model = sm.OLS(y, X).fit()# Print model summary print(model.summary())

Date:	Mon	, 16 Sep 202	4 Prob	(F-statistic):	7.76e-07	
Time:		17:23:2	6 Log-I	Likelihood:		-1085.	
No. Observations:		30	3 AIC:	AIC:		2175.	
Df Residua	ls:	30	1 BIC:			2183.	
Df Model:			1				
Covariance	Type:	nonrobus	t				
=======	coef	std err	t	P> t	[0.025	0.975]	
const	35.3255	3.806	9.283	0.000	27.837	42.814	
trtbps	0.1447	0.029	5.048	0.000	0.088	0.201	
Omnibus:		 1.86	5 Durbi	in-Watson:		2.015	
Prob(Omnibus):		0.39	4 Jarqı	ue-Bera (JB):		1.782	
Skew:		-0.10	8 Prob	(JB):		0.410	
Kurtosis:		2.69	3 Cond.	. No.		1.01e+03	

OLS Regression Results ______

Least Squares F-statistic:

0.078 0.075

25.48

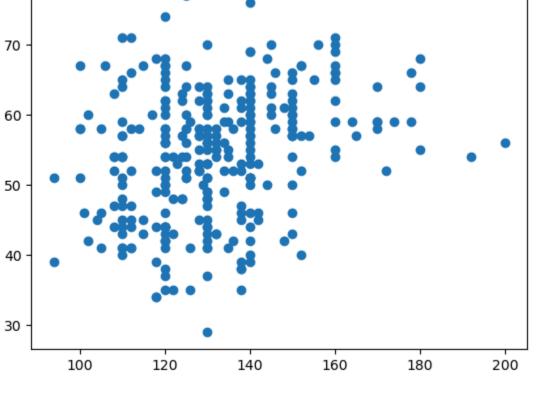
Dep. Variable: age R-squared:
Model: OLS Adj. R-squared:

Method:

Notes: [1] Standard Errors assume that the covariance matrix of the errors is correctly specified. [2] The condition number is large, 1.01e+03. This might indicate that there are strong multicollinearity or other numerical problems.

In [10]: import matplotlib.pyplot as plt In [22]: plt.scatter(df['trtbps'],y) plt.show()

70 -	•	
70 -		
60 -		•••



CONCLUSION we have R-SQUARED = 0.078