# **Problem 1**

1. Mean: 1.0489703904839585 Variance: 5.427220681881726 Skewness: 0.8806086425277363 Kurtosis: 23.12220078998973

2. Mean: 1.0489703904839585 Variance: 5.427220681881727 Skewness: 0.8806086425277364 Kurtosis: 23.122200789989723

3. I use the hypothesis testing with the null hypothesis that the statistical package functions are not biased. Since the p-values of 4 moments are not all greater than alpha (0.05), we need to reject the hypothesis that the statistical package functions are not biased.

# Problem 2

1.

# OLS Regression Results

Dep. Variable:	у	R-squared:	0.346
Model:	0LS	Adj. R-squared:	0.342
Method:	Least Squares	F-statistic:	104.6
Date:	Sat, 14 Sep 2024	<pre>Prob (F-statistic):</pre>	5.59e-20
Time:	01:26:30	Log-Likelihood:	-284.54
No. Observations:	200	AIC:	573.1
Df Residuals:	198	BIC:	579 <b>.</b> 7
Df Model:	1		

Covariance Type: nonrobust

	0.0874					
	0.7753	0.071 0.076	-1.222 10.226	0.223 0.000	-0.228 0.626	0.054 0.925
Omnibus: Prob(Omnibus): Skew: Kurtosis:		11.9 0.0 0.3 4.1	003 Jarque 887 Prob(S	•		2.023 16.685 0.000238 1.09

# **MLE Results**

MLE Beta: [-0.08738446 0.7752741]

MLE Standard Deviation (Sigma): 1.008813058320225

#### **Comparison of Results:**

OLS Beta: [-0.08738446 0.7752741 ]

OLS Standard Deviation of Errors: 1.008813058320225

Difference in Beta: [0. 0.]

Difference in Standard Deviation (Sigma): 0.0

In a linear regression context, with the assumption of normally distributed errors, **the OLS estimators are also the MLE estimators**. This is why there is no difference between the coefficients (Beta) and the estimated standard deviation of errors (Sigma) when comparing results from the OLS and MLE methods.

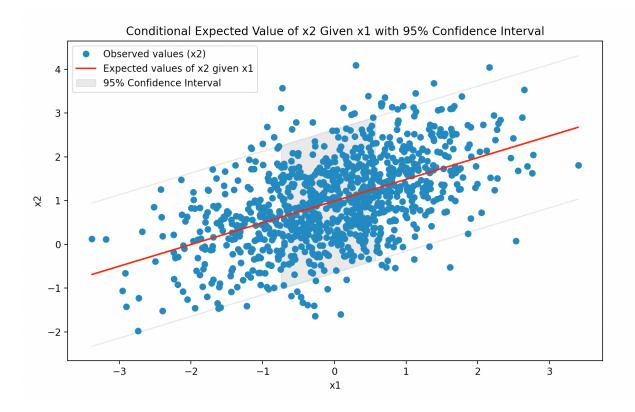
# 2.

### **Comparison of Results**

Adjusted R-squared MLE (Normal): 0.3389632986263845 Adjusted R-squared MLE (T-distribution): 0.3379809763066779

I use goodness of fit statistics to decide which model is the best fit. Although the adjusted R-squared for both models is almost the same, MLE (Normal) would be the best fit in this case since it is slightly greater than MLE (T-distribution).

### 3.



# **Problem 3**

Before comparing the models, we expect that the model with the lowest AIC value is considered the best fit because a lower AIC indicates a model that better balances goodness-of-fit with model complexity.

AR(1) AIC: 1644.6555047688475

AR(2) AIC: 1581.0792659049775

AR(3) AIC: 1436.6598066945892

MA(1) AIC: 1567.4036263707874

MA(2) AIC: 1537.941206380739

MA(3) AIC: 1536.8677087350304

# The best model is: AR(3) with AIC: 1436.6598066945892

Also, by observing the graph, we can find that the autocorrelation value gradually decreases to 0 for the ACF graph. For the PACF graph, the partial autocorrelation value suddenly drops to 0. Therefore, we can infer that the AR(3) model might be the best fit.

