

Required Libraries

```
In [ ]: import yfinance as yf
import pandas as pd
from datetime import datetime
from arch import arch_model
import matplotlib.pyplot as plt
import numpy as np
import warnings
warnings.filterwarnings("ignore")
```

Question 1

```
In [ ]: start_date = "2022-08-30"
end_date = "2024-09-30"
```

```
In [ ]: exchange_tickers = ["ZAR=X", "AUDUSD=X", "GBPUSD=X"]
exchange_data = {ticker: yf.download(ticker, start=start_date, end=end_date) for ticker in exchange_tickers}
```

```
[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed
```

```
In [ ]: for ticker, data in exchange_data.items():
    print(f"Data for {ticker}")
    print(data.head())
```

Data for ZAR=X

	Open	High	Low	Close	Adj Close	Volume
Date						
2022-08-30	16.809641	16.972799	16.745100	16.809641	16.809641	0
2022-08-31	16.957800	17.087179	16.858629	16.957800	16.957800	0
2022-09-01	17.105209	17.304930	17.088301	17.105209	17.105209	0
2022-09-02	17.251900	17.345699	17.172701	17.251900	17.251900	0
2022-09-05	17.300900	17.418039	17.146500	17.300900	17.300900	0

Data for AUDUSD=X

	Open	High	Low	Close	Adj Close	Volume
Date						
2022-08-30	0.690751	0.695797	0.685990	0.690751	0.690751	0
2022-08-31	0.685740	0.690422	0.684340	0.685740	0.685740	0
2022-09-01	0.682850	0.684360	0.677172	0.682850	0.682850	0
2022-09-02	0.679530	0.685380	0.678109	0.679530	0.679530	0
2022-09-05	0.679030	0.680411	0.677360	0.679030	0.679030	0

Data for GBPUSD=X

	Open	High	Low	Close	Adj Close	Volume
Date						
2022-08-30	1.172127	1.175807	1.162304	1.172251	1.172251	0
2022-08-31	1.166086	1.169454	1.160025	1.165977	1.165977	0
2022-09-01	1.159689	1.161710	1.150324	1.159851	1.159851	0
2022-09-02	1.154734	1.158856	1.153136	1.154894	1.154894	0
2022-09-05	1.147855	1.152074	1.144518	1.147460	1.147460	0

```
In [ ]: def calculate_returns(df):
    df['Returns'] = df['Adj Close'].pct_change().dropna() * 100
    return df['Returns'].dropna()
```

```
In [ ]: def fit_garch_model(returns, model_type='GARCH'):
    if model_type == 'ARCH':
        model = arch_model(returns, vol='Arch', p=1)
    elif model_type == 'GARCH':
        model = arch_model(returns, vol='Garch', p=1, q=1)
    elif model_type == 'EGARCH':
        model = arch_model(returns, vol='EGarch', p=1, q=1)
    elif model_type == 'TGARCH':
        model = arch_model(returns, vol='Garch', p=1, o=1, q=1)
    elif model_type == 'PGARCH':
        model = arch_model(returns, p=1, o=1, q=1, power=1.0)
    else:
        raise ValueError(f"Unsupported model type: {model_type}")
```

```
model_fit = model.fit(disp='off')
return model_fit
```

```
In [ ]: def visualize_results(returns, model_results, title):

    # plot returns
    plt.figure(figsize=(20, 10))
    plt.subplot(2, 1, 1)
    plt.plot(returns, label='Returns', color='blue')
    plt.title(f"Daily Returns for {title}", fontsize=20, fontweight='bold')
    plt.xlabel('Date')
    plt.ylabel('Returns')
    plt.legend()

    # plot volatility
    plt.subplot(2, 1, 2)
    plt.plot(model_results.conditional_volatility, label='Conditional Volatility', color='red')
    plt.title(f"Conditional Volatility for {title}", fontsize=20, fontweight='bold')
    plt.xlabel('Date')
    plt.ylabel('Volatility')
    plt.legend()
    plt.tight_layout()
    plt.show()
```

```
In [ ]: def compare_volatility_forecasting(returns, models):

    realized_volatility = returns**2
    forecast_errors = {}

    for model_name, model_results in models.items():
        forecasted_volatility = model_results.conditional_volatility

        mse = np.mean((realized_volatility - forecasted_volatility**2)**2)

        forecast_errors[model_name] = mse

        print(f"{model_name} - Volatility Forecasting MSE: {mse}")

    best_model = min(forecast_errors, key=forecast_errors.get)
```

```
    return best_model
```

```
In [ ]: def evaluate_and_visualize_models(returns, key):
    print(f"\n*****Processing {key}*****")

    models = {}

    # ARCH
    print("Fitting ARCH(1) model...")
    arch_res = fit_garch_model(returns, model_type='ARCH')
    models['ARCH'] = arch_res
    print(arch_res.summary())
    visualize_results(returns, arch_res, f"{key} ARCH(1)")

    # GARCH
    print("Fitting GARCH(1,1) model...")
    garch_res = fit_garch_model(returns, model_type='GARCH')
    models['GARCH'] = garch_res
    print(garch_res.summary())
    visualize_results(returns, garch_res, f"{key} GARCH(1,1)")

    # EGARCH
    print("Fitting EGARCH(1,1) model...")
    egarch_res = fit_garch_model(returns, model_type='EGARCH')
    models['EGARCH'] = egarch_res
    print(egarch_res.summary())
    visualize_results(returns, egarch_res, f"{key} EGARCH(1,1)")

    # TGARCH
    print("Fitting TGARCH(1,1) model...")
    tgarch_res = fit_garch_model(returns, model_type='TGARCH')
    models['TGARCH'] = tgarch_res
    print(tgarch_res.summary())
    visualize_results(returns, tgarch_res, f"{key} TGARCH(1,1)")

    # PGARCH
    print("Fitting PGARCH(1,1) model...")
    pgarch_res = fit_garch_model(returns, model_type='PGARCH')
    models['PGARCH'] = pgarch_res
    print(pgarch_res.summary())
    visualize_results(returns, pgarch_res, f"{key} PGARCH(1,1)")
```

```
# Compare AIC and BIC
print("\nModel Comparisons:")
print(f"ARCH AIC: {arch_res.aic}, BIC: {arch_res.bic}")
print(f"GARCH AIC: {garch_res.aic}, BIC: {garch_res.bic}")
print(f"EGARCH AIC: {egarch_res.aic}, BIC: {egarch_res.bic}")
print(f"TGARCH AIC: {tgarch_res.aic}, BIC: {tgarch_res.bic}")
print(f"PGARCH AIC: {pgarch_res.aic}, BIC: {pgarch_res.bic}")

best_bic = min(arch_res.bic, garch_res.bic, egarch_res.bic, tgarch_res.bic, pgarch_res.bic)
if best_bic == arch_res.bic:
    print("\nBest model based on BIC: ARCH")
elif best_bic == garch_res.bic:
    print("\nBest model based on BIC: GARCH")
elif best_bic == egarch_res.bic:
    print("\nBest model based on BIC: EGARCH")
elif best_bic == pgarch_res.bic:
    print("\nBest model based on BIC: PGARCH")
else:
    print("\nBest model based on BIC: TGARCH")

# Compare models based on volatility forecasting
print("\nComparing models based on volatility forecasting...")
best_forecast_model = compare_volatility_forecasting(returns, models)
print(f"\nBest model based on volatility forecasting (lowest mse): {best_forecast_model}")
```

In []: `for key, df in exchange_data.items():
 returns = calculate_returns(df)
 evaluate_and_visualize_models(returns, key)`

*****processing ZAR=X*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	ARCH	Log-Likelihood:	-679.737
Distribution:	Normal	AIC:	1365.47
Method:	Maximum Likelihood	BIC:	1378.28
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:39	Df Model:	1

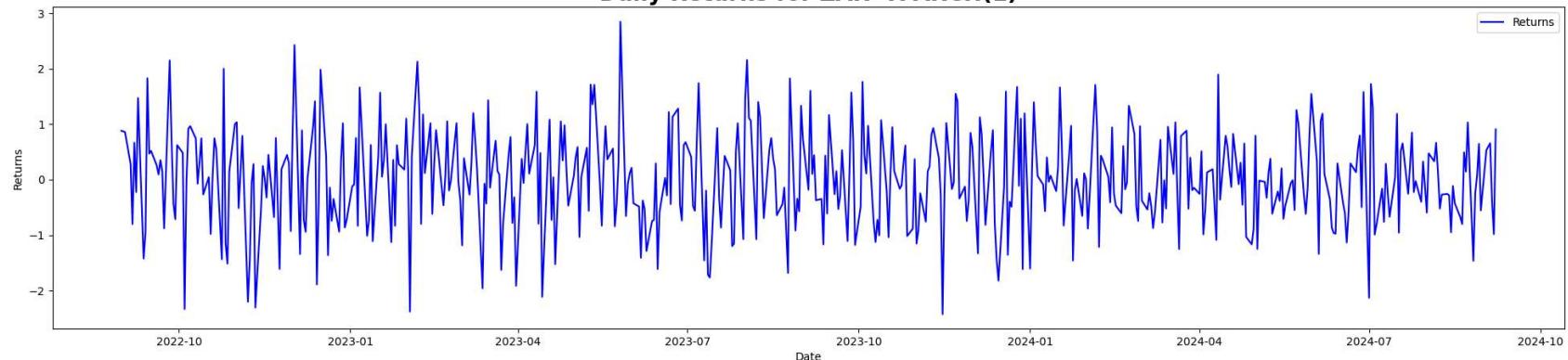
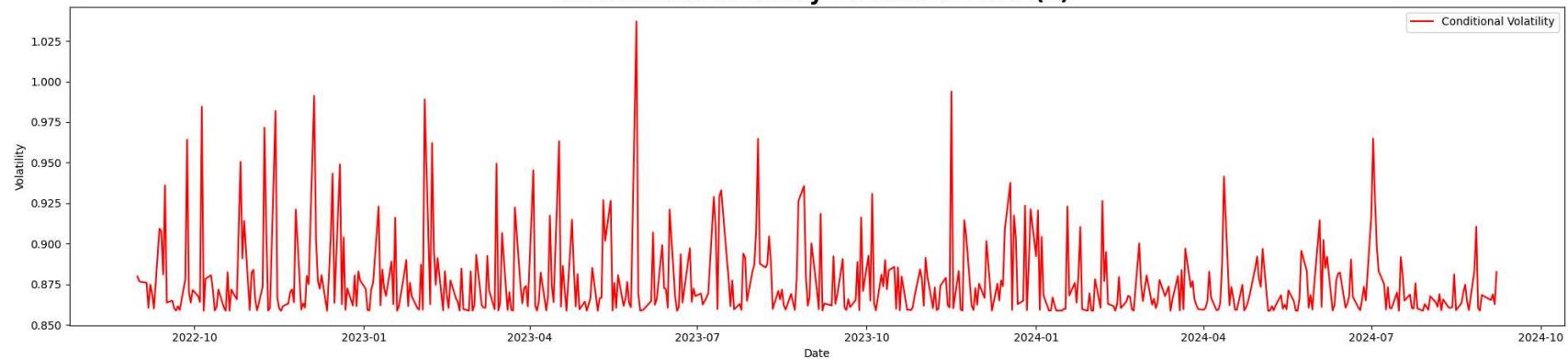
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0134	3.825e-02	0.351	0.726	[-6.157e-02, 8.839e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.7374	6.009e-02	12.273	1.270e-34	[0.620, 0.855]
alpha[1]	0.0421	4.723e-02	0.892	0.373	[-5.046e-02, 0.135]

Covariance estimator: robust

Daily Returns for ZAR=X ARCH(1)**Conditional Volatility for ZAR=X ARCH(1)**

Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-676.275
Distribution:	Normal	AIC:	1360.55
Method:	Maximum Likelihood	BIC:	1377.63
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:39	Df Model:	1

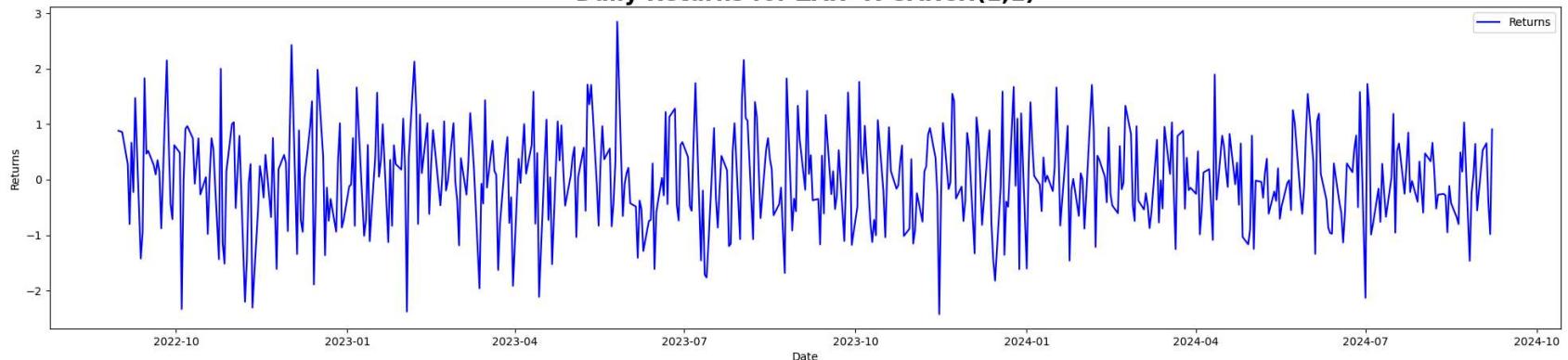
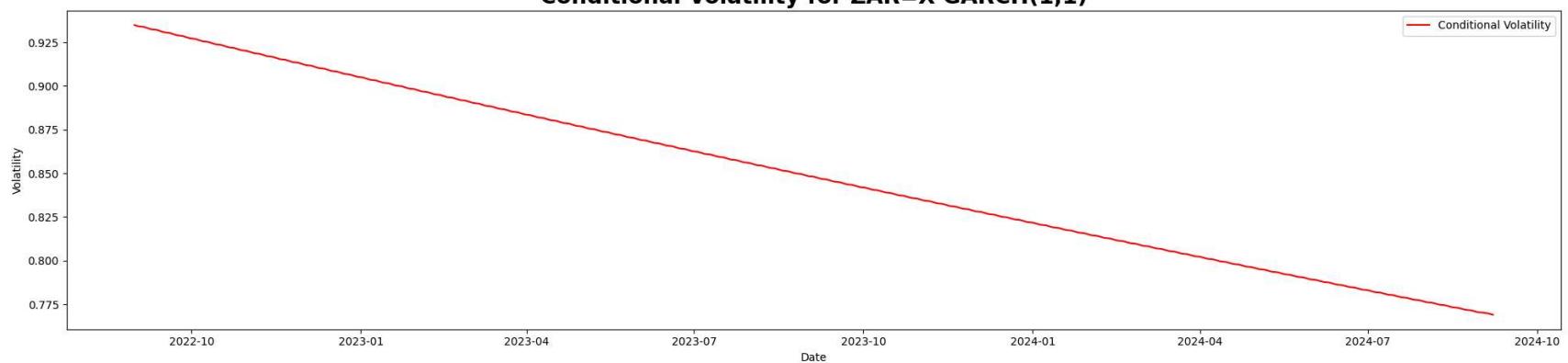
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0111	3.791e-02	0.292	0.770	[-6.322e-02,8.539e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	2.3413e-07	2.277e-03	1.028e-04	1.000	[-4.462e-03,4.462e-03]
alpha[1]	8.0898e-07	3.189e-03	2.537e-04	1.000	[-6.249e-03,6.251e-03]
beta[1]	0.9993	1.678e-03	595.342	0.000	[0.996, 1.003]

Covariance estimator: robust

Daily Returns for ZAR=X GARCH(1,1)**Conditional Volatility for ZAR=X GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

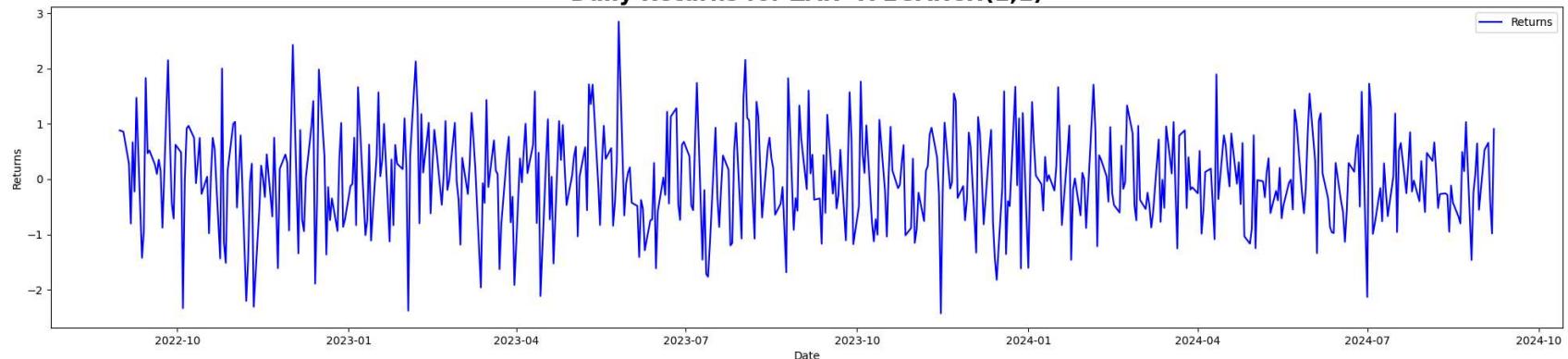
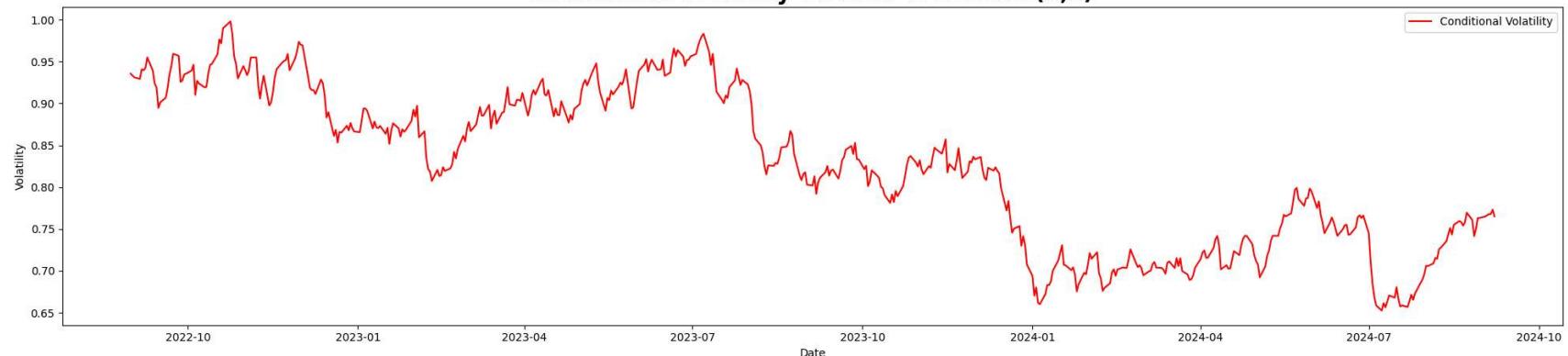
Dep. Variable:	Returns	R-squared:	0.000		
Mean Model:	Constant Mean	Adj. R-squared:	0.000		
Vol Model:	EGARCH	Log-Likelihood:	-670.801		
Distribution:	Normal	AIC:	1349.60		
Method:	Maximum Likelihood	BIC:	1366.68		
		No. Observations:	528		
Date:	Sat, Sep 07 2024	Df Residuals:	527		
Time:	20:05:40	Df Model:	1		
Mean Model					
=====					
	coef	std err	t	P> t	95.0% Conf. Int.
<hr/>					
mu	0.0149	1.210e-08	1.231e+06	0.000	[1.489e-02,1.489e-02]
Volatility Model					
=====					
	coef	std err	t	P> t	95.0% Conf. Int.
<hr/>					
omega	6.6501e-04	3.632e-12	1.831e+08	0.000	[6.650e-04,6.650e-04]
alpha[1]	-0.0466	9.617e-09	-4.849e+06	0.000	[-4.664e-02,-4.664e-02]
beta[1]	0.9979	5.895e-10	1.693e+09	0.000	[0.998, 0.998]
=====					

Covariance estimator: robust

WARNING: The optimizer did not indicate successful convergence. The message was Iteration limit reached.
See convergence_flag.

```
c:\Users\Alli Ajagbe\AppData\Local\Programs\Python\Python311\Lib\site-packages\arch\univariate\base.py:766: ConvergenceWarning: The optimizer returned code 9. The message is:
Iteration limit reached
See scipy.optimize.fmin_slsqp for code meaning.
```

```
warnings.warn(
```

Daily Returns for ZAR=X EGARCH(1,1)**Conditional Volatility for ZAR=X EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-676.178
Distribution:	Normal	AIC:	1362.36
Method:	Maximum Likelihood	BIC:	1383.70
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:40	Df Model:	1

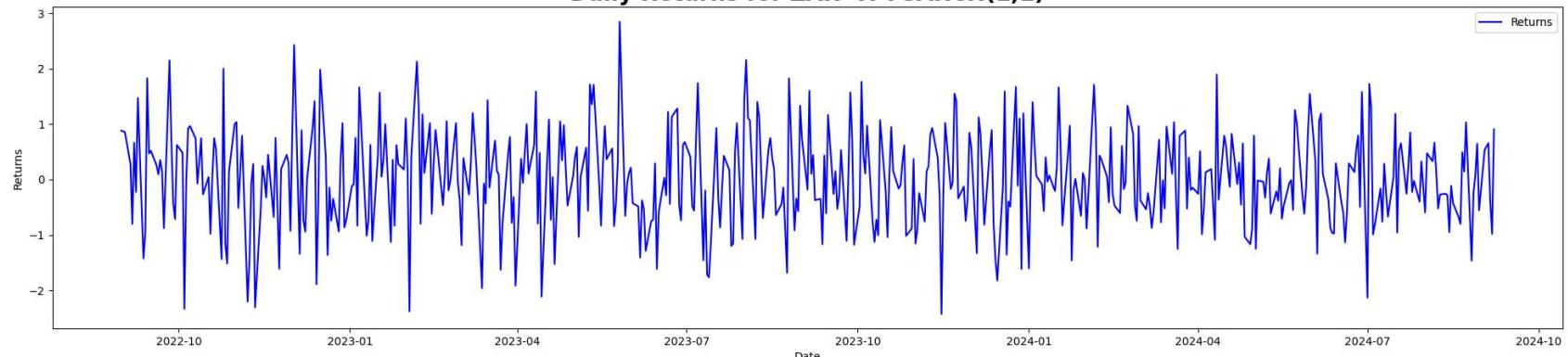
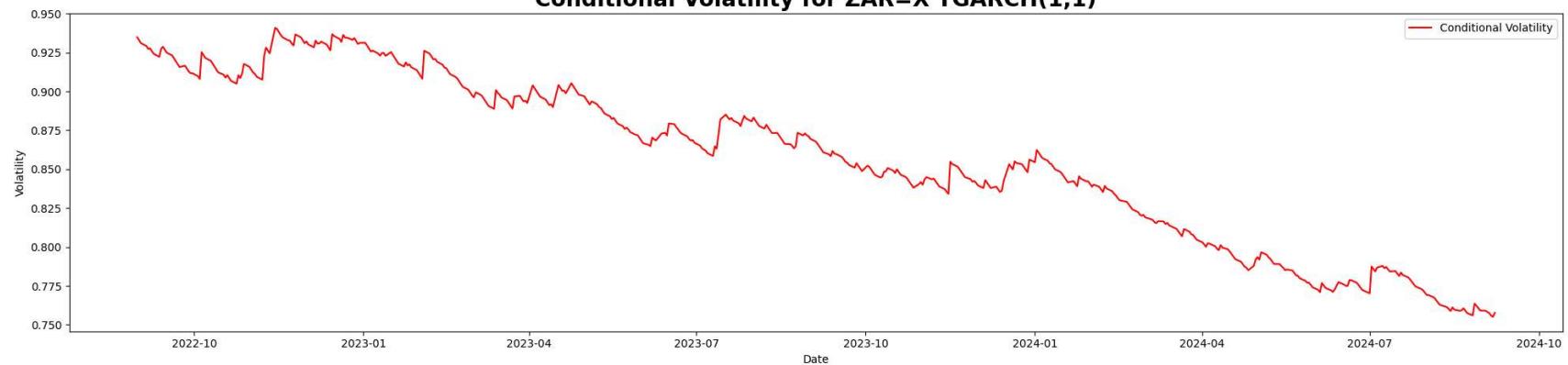
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	6.0180e-03	4.211e-02	0.143	0.886	[-7.652e-02,8.856e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	7.4023e-07	2.767e-03	2.675e-04	1.000	[-5.423e-03,5.424e-03]
alpha[1]	2.9164e-07	1.359e-02	2.147e-05	1.000	[-2.663e-02,2.663e-02]
gamma[1]	6.3935e-03	2.559e-02	0.250	0.803	[-4.376e-02,5.655e-02]
beta[1]	0.9960	4.780e-03	208.361	0.000	[0.987, 1.005]

Covariance estimator: robust

Daily Returns for ZAR=X TGARCH(1,1)**Conditional Volatility for ZAR=X TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-678.950
Distribution:	Normal	AIC:	1367.90
Method:	Maximum Likelihood	BIC:	1389.25
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:41	Df Model:	1

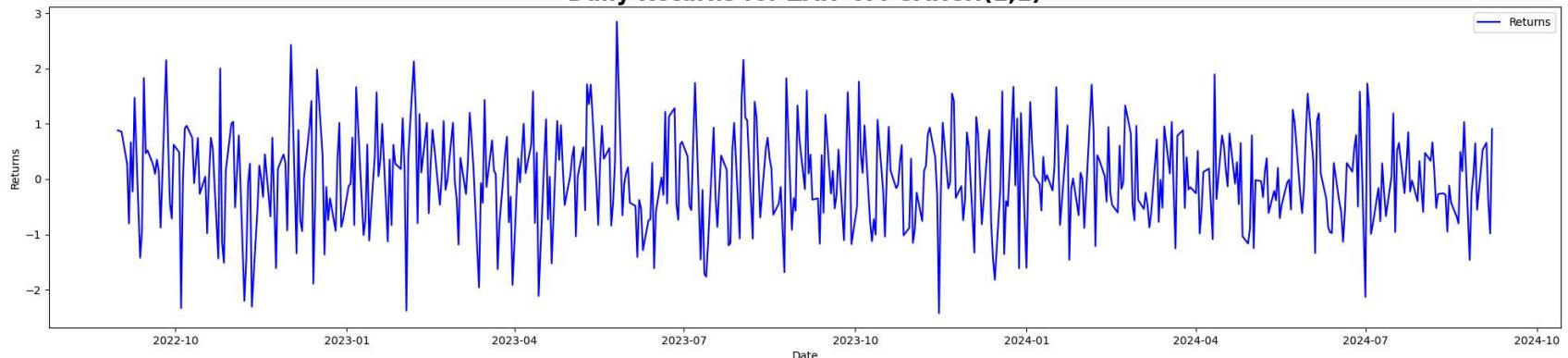
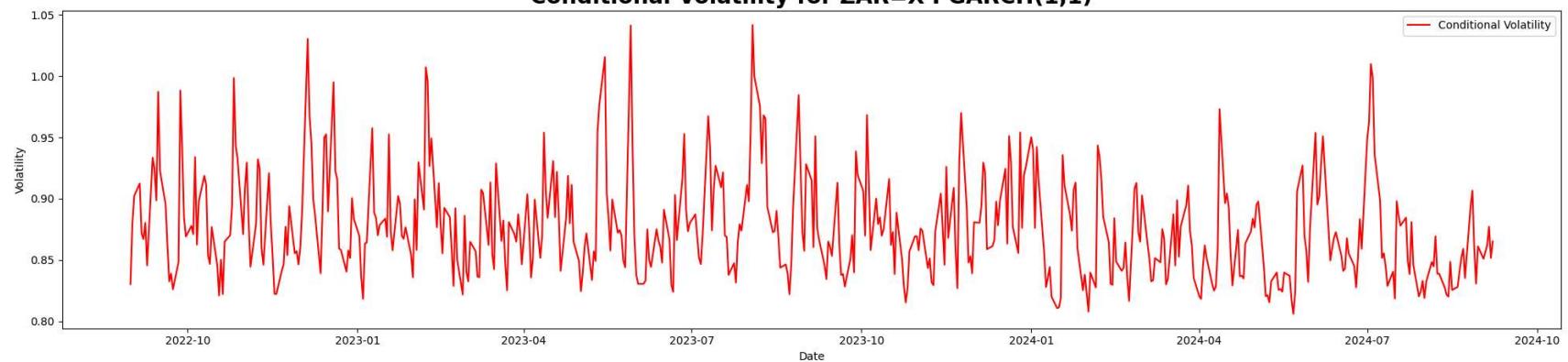
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0138	3.877e-02	0.357	0.721	[-6.216e-02, 8.983e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.4107	0.169	2.426	1.528e-02	[7.883e-02, 0.742]
alpha[1]	0.0789	4.821e-02	1.638	0.101	[-1.553e-02, 0.173]
gamma[1]	-0.0350	5.830e-02	-0.600	0.549	[-0.149, 7.931e-02]
beta[1]	0.4822	0.193	2.494	1.264e-02	[0.103, 0.861]

Covariance estimator: robust

Daily Returns for ZAR=X PGARCH(1,1)**Conditional Volatility for ZAR=X PGARCH(1,1)**

Model Comparisons:

ARCH AIC: 1365.4739414081632, BIC: 1378.281230259282
 GARCH AIC: 1360.5508589318451, BIC: 1377.6272440666703
 EGARCH AIC: 1349.6025407396355, BIC: 1366.6789258744607
 TGARCH AIC: 1362.3556485727302, BIC: 1383.7011299912615
 PGARCH AIC: 1367.9000905762286, BIC: 1389.24557199476

Best model based on BIC: EGARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 1.1501724653014256
 GARCH - Volatility Forecasting MSE: 1.1363881133247706
 EGARCH - Volatility Forecasting MSE: 1.1138408425359607
 TGARCH - Volatility Forecasting MSE: 1.135994053682568
 PGARCH - Volatility Forecasting MSE: 1.1465230503358819

Best model based on volatility forecasting (lowest mse): EGARCH

*****processing AUDUSD=X*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	ARCH	Log-Likelihood:	-542.644
Distribution:	Normal	AIC:	1091.29
Method:	Maximum Likelihood	BIC:	1104.10
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:41	Df Model:	1

Mean Model

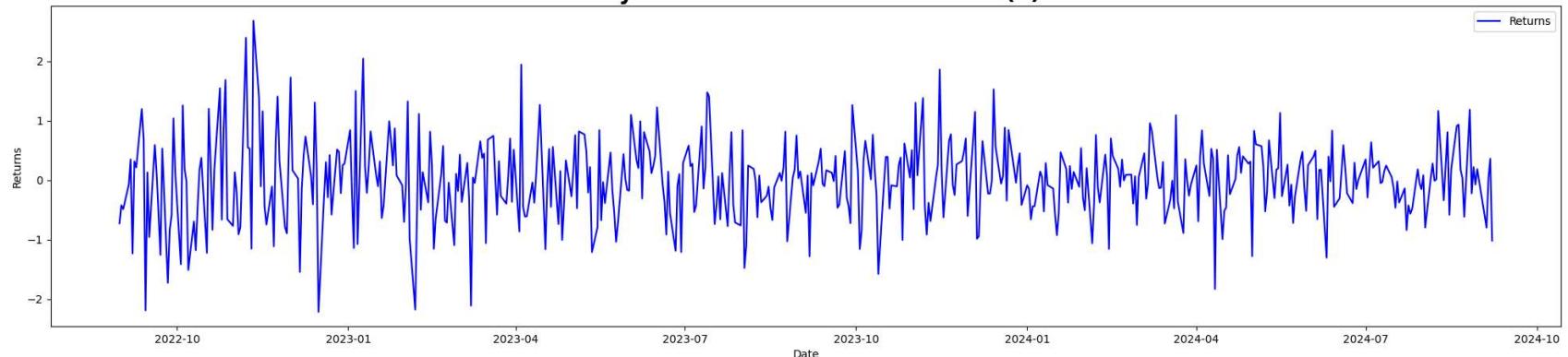
	coef	std err	t	P> t	95.0% Conf. Int.
mu	-3.5260e-03	2.996e-02	-0.118	0.906	[-6.224e-02, 5.519e-02]

Volatility Model

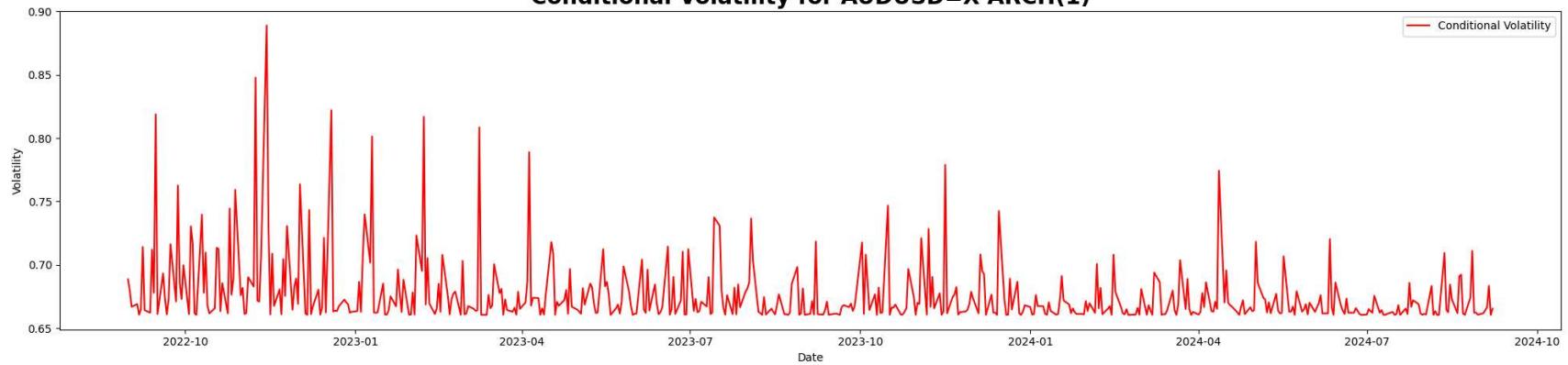
	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.4363	3.882e-02	11.237	2.673e-29	[0.360, 0.512]
alpha[1]	0.0490	6.578e-02	0.744	0.457	[-7.996e-02, 0.178]

Covariance estimator: robust

Daily Returns for AUDUSD=X ARCH(1)



Conditional Volatility for AUDUSD=X ARCH(1)



Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-518.640
Distribution:	Normal	AIC:	1045.28
Method:	Maximum Likelihood	BIC:	1062.36
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:42	Df Model:	1

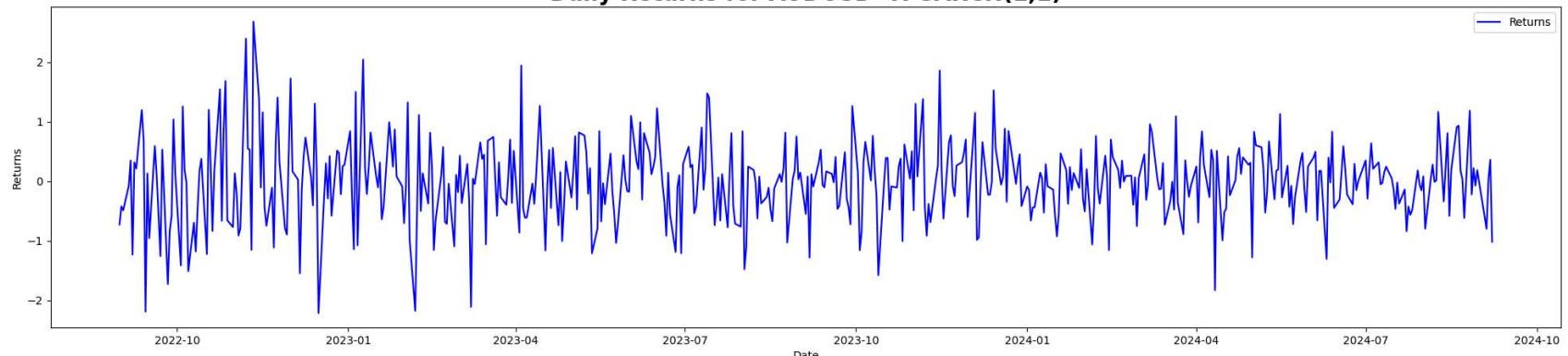
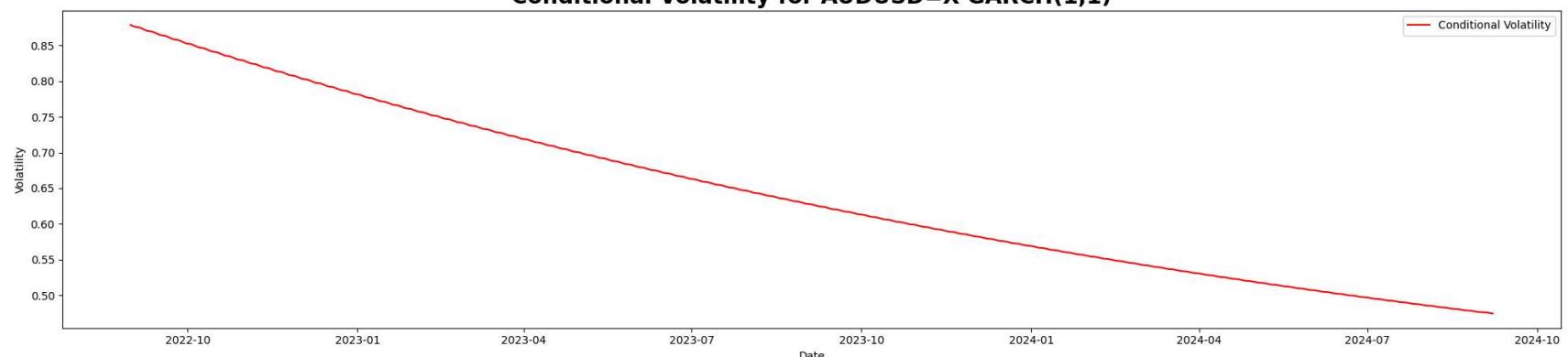
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	6.1500e-04	2.703e-02	2.275e-02	0.982	[-5.237e-02,5.360e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	2.8021e-04	7.842e-04	0.357	0.721	[-1.257e-03,1.817e-03]
alpha[1]	0.0000	3.281e-03	0.000	1.000	[-6.431e-03,6.431e-03]
beta[1]	0.9969	4.602e-03	216.612	0.000	[0.988, 1.006]

Covariance estimator: robust

Daily Returns for AUDUSD=X GARCH(1,1)**Conditional Volatility for AUDUSD=X GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	EGARCH	Log-Likelihood:	-516.859
Distribution:	Normal	AIC:	1041.72
Method:	Maximum Likelihood	BIC:	1058.79
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:43	Df Model:	1

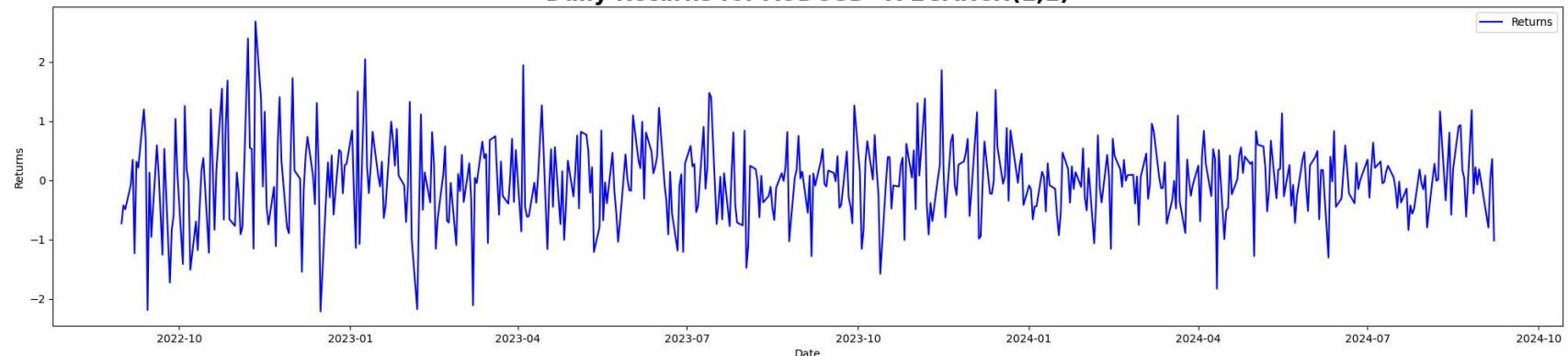
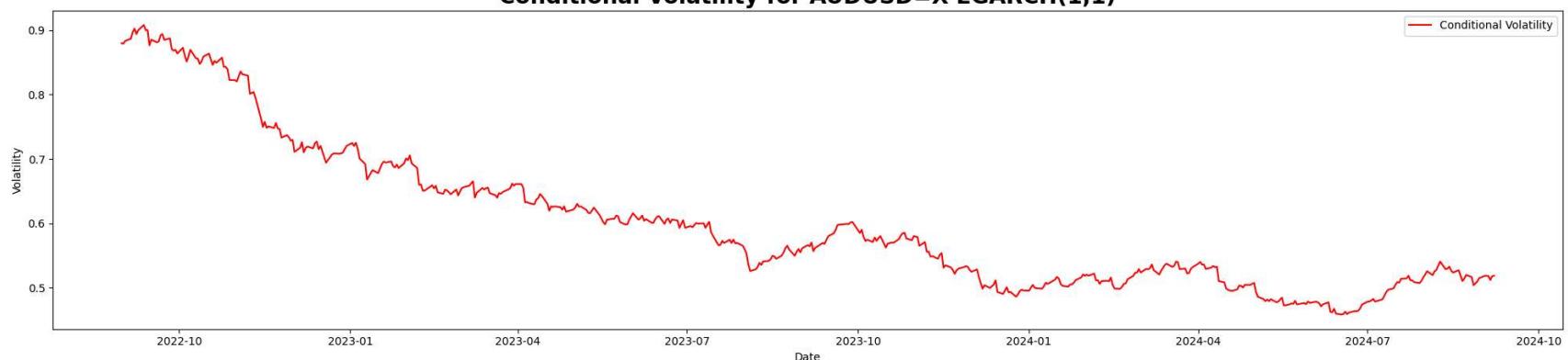
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	-1.7269e-03	8.471e-03	-0.204	0.838	[-1.833e-02,1.488e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	-6.3099e-04	6.665e-10	-9.468e+05	0.000	[-6.310e-04,-6.310e-04]
alpha[1]	-0.0325	1.378e-05	-2359.576	0.000	[-3.253e-02,-3.248e-02]
beta[1]	1.0000	3.385e-11	2.954e+10	0.000	[1.000, 1.000]

Covariance estimator: robust

Daily Returns for AUDUSD=X EGARCH(1,1)**Conditional Volatility for AUDUSD=X EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-518.640
Distribution:	Normal	AIC:	1047.28
Method:	Maximum Likelihood	BIC:	1068.63
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:43	Df Model:	1

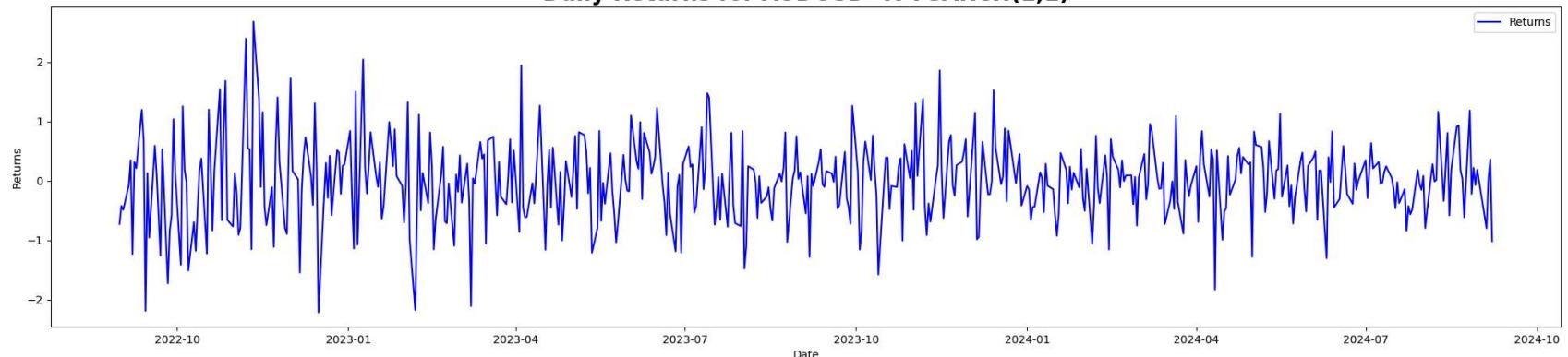
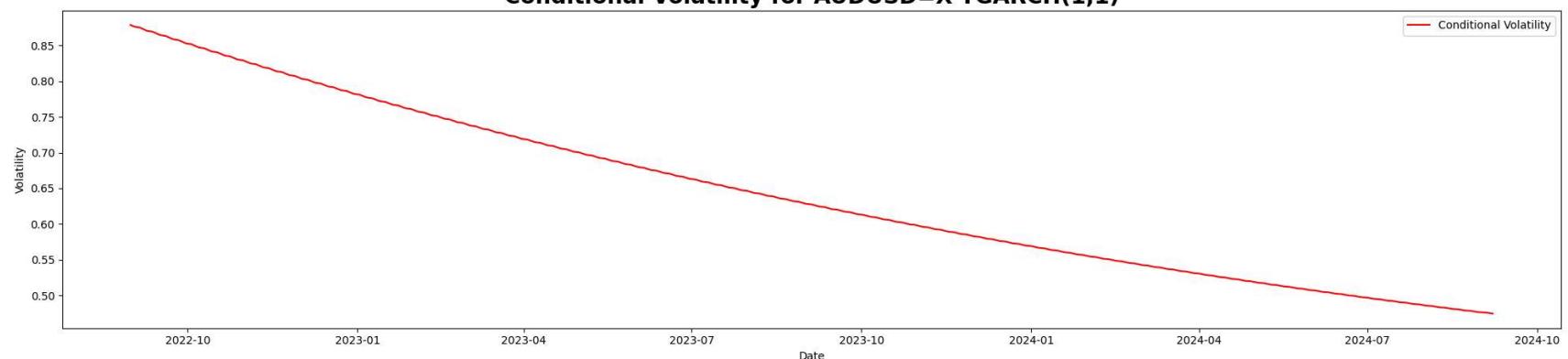
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	6.1404e-04	2.716e-02	2.261e-02	0.982	[-5.262e-02,5.385e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	2.8007e-04	7.843e-04	0.357	0.721	[-1.257e-03,1.817e-03]
alpha[1]	0.0000	6.746e-03	0.000	1.000	[-1.322e-02,1.322e-02]
gamma[1]	-1.4614e-10	1.514e-02	-9.651e-09	1.000	[-2.968e-02,2.968e-02]
beta[1]	0.9969	4.811e-03	207.197	0.000	[0.987, 1.006]

Covariance estimator: robust

Daily Returns for AUDUSD=X TGARCH(1,1)**Conditional Volatility for AUDUSD=X TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-522.817
Distribution:	Normal	AIC:	1055.63
Method:	Maximum Likelihood	BIC:	1076.98
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:44	Df Model:	1

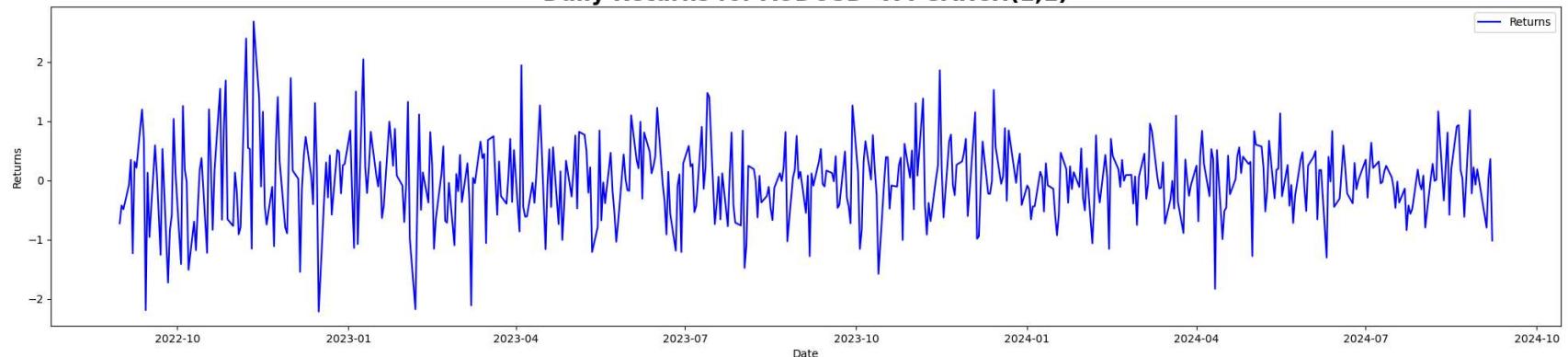
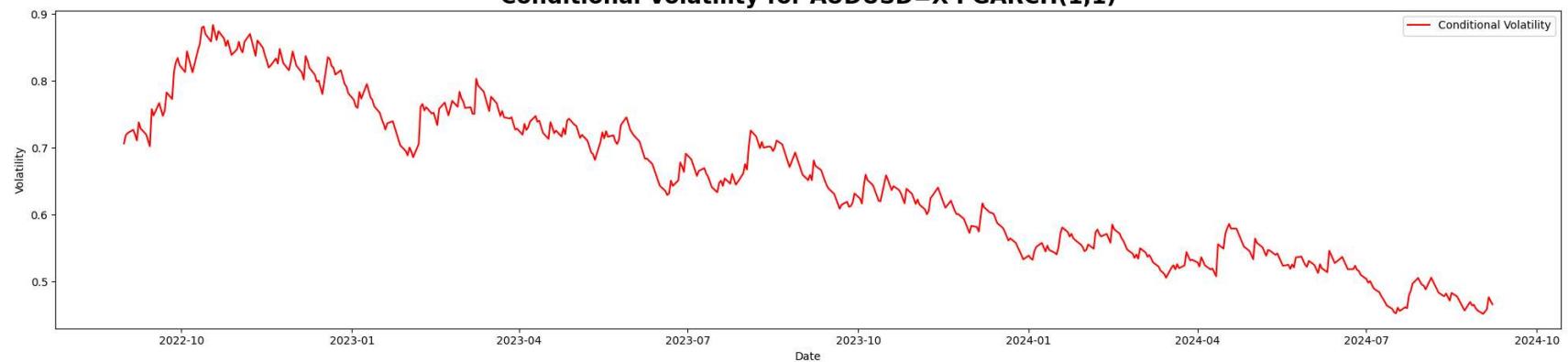
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	-0.0120	2.747e-02	-0.437	0.662	[-6.584e-02,4.183e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.7088e-03	2.605e-03	0.656	0.512	[-3.396e-03,6.814e-03]
alpha[1]	0.0000	2.305e-02	0.000	1.000	[-4.518e-02,4.518e-02]
gamma[1]	0.0295	2.359e-02	1.252	0.211	[-1.671e-02,7.577e-02]
beta[1]	0.9852	1.326e-02	74.291	0.000	[0.959, 1.011]

Covariance estimator: robust

Daily Returns for AUDUSD=X PGARCH(1,1)**Conditional Volatility for AUDUSD=X PGARCH(1,1)**

Model Comparisons:

ARCH AIC: 1091.2878891517414, BIC: 1104.0951780028602
 GARCH AIC: 1045.2798563631154, BIC: 1062.3562414979406
 EGARCH AIC: 1041.7182170275305, BIC: 1058.7946021623557
 TGARCH AIC: 1047.2798566243118, BIC: 1068.6253380428432
 PGARCH AIC: 1055.6334493925367, BIC: 1076.978930811068

Best model based on BIC: EGARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 0.6419641481631231
 GARCH - Volatility Forecasting MSE: 0.6012366262752905
 EGARCH - Volatility Forecasting MSE: 0.6013808781523678
 TGARCH - Volatility Forecasting MSE: 0.6012364888996955
 PGARCH - Volatility Forecasting MSE: 0.6077889091498082

Best model based on volatility forecasting (lowest mse): TGARCH

*****processing GBPUSD=X*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	ARCH	Log-Likelihood:	-461.819
Distribution:	Normal	AIC:	929.639
Method:	Maximum Likelihood	BIC:	942.446
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:44	Df Model:	1

Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	-3.4596e-03	2.401e-02	-0.144	0.885	[-5.053e-02, 4.361e-02]

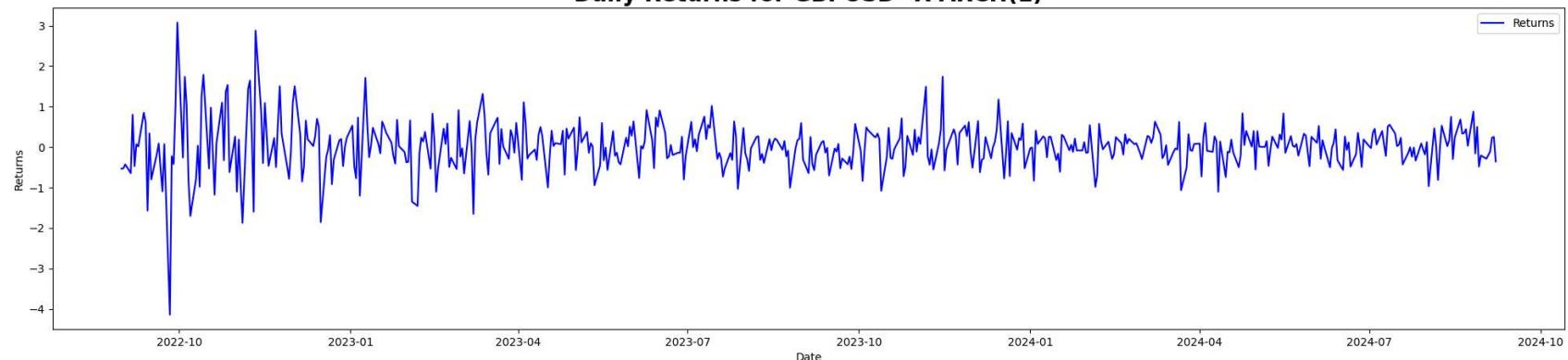
Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.2693	5.535e-02	4.865	1.147e-06	[0.161, 0.378]
alpha[1]	0.2765	0.125	2.217	2.662e-02	[3.207e-02, 0.521]

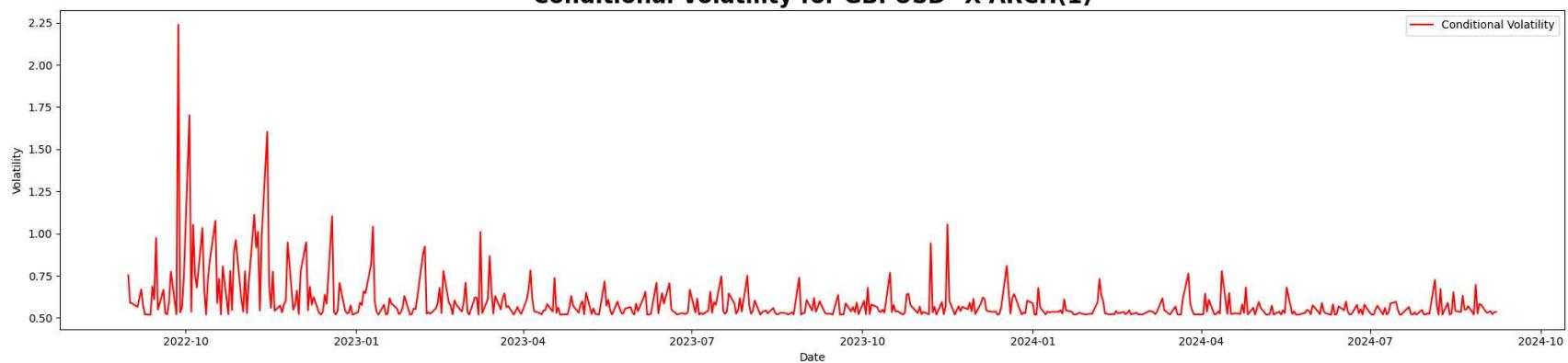
=====

Covariance estimator: robust

Daily Returns for GBPUSD=X ARCH(1)



Conditional Volatility for GBPUSD=X ARCH(1)



Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-396.226
Distribution:	Normal	AIC:	800.452
Method:	Maximum Likelihood	BIC:	817.528
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:45	Df Model:	1

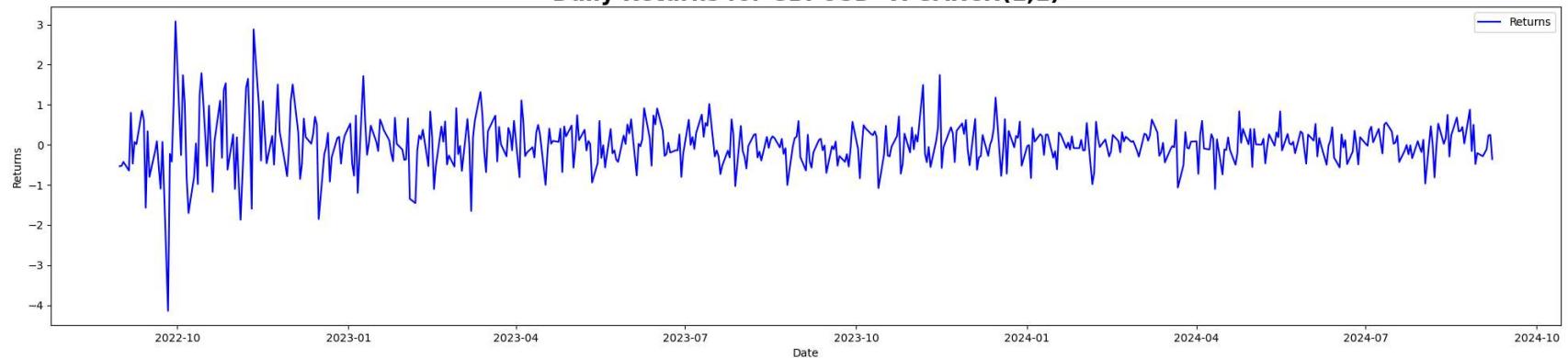
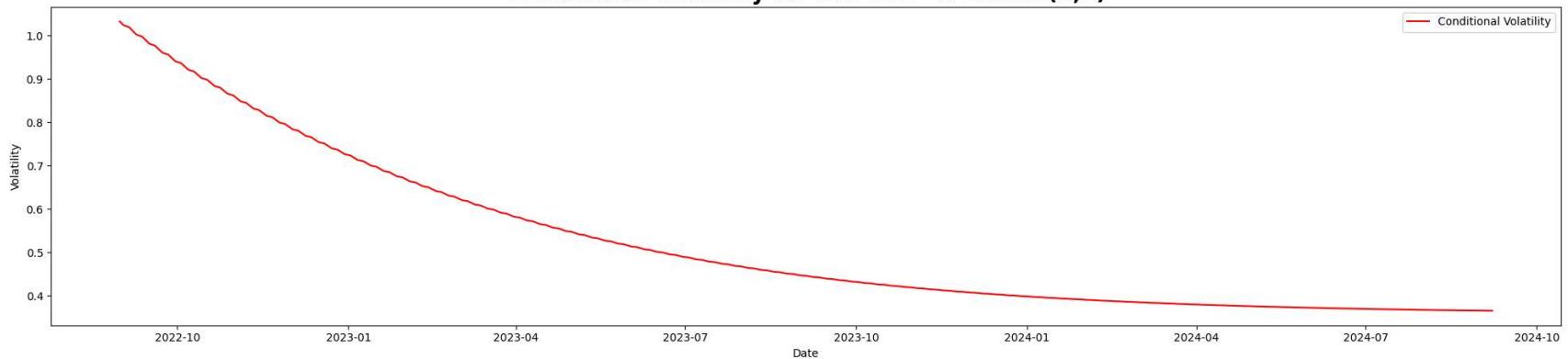
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0213	2.034e-02	1.045	0.296	[-1.860e-02,6.111e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.2528e-03	6.924e-04	1.809	7.041e-02	[-1.043e-04,2.610e-03]
alpha[1]	2.8289e-13	1.029e-02	2.750e-11	1.000	[-2.016e-02,2.016e-02]
beta[1]	0.9902	1.317e-02	75.170	0.000	[0.964, 1.016]

Covariance estimator: robust

Daily Returns for GBPUSD=X GARCH(1,1)**Conditional Volatility for GBPUSD=X GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	EGARCH	Log-Likelihood:	-387.853
Distribution:	Normal	AIC:	783.705
Method:	Maximum Likelihood	BIC:	800.782
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:46	Df Model:	1

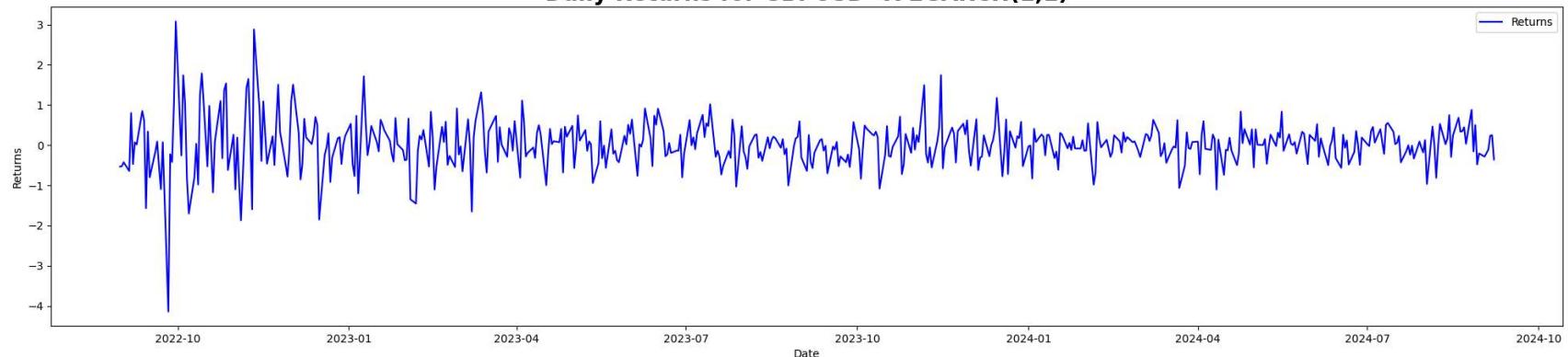
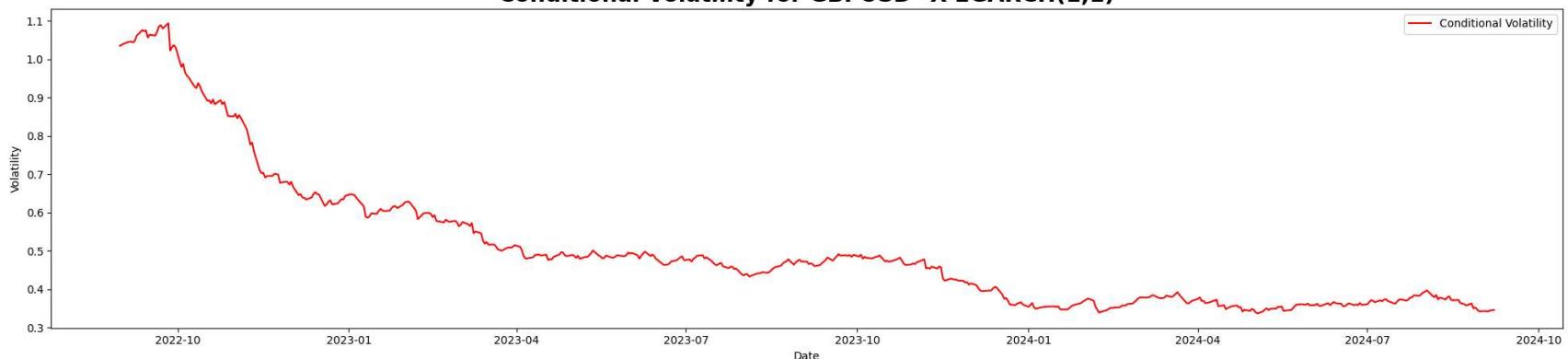
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0124	9.678e-09	1.281e+06	0.000	[1.239e-02,1.239e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	-6.1775e-03	2.369e-12	-2.607e+09	0.000	[-6.177e-03,-6.177e-03]
alpha[1]	-0.0427	5.392e-10	-7.921e+07	0.000	[-4.271e-02,-4.271e-02]
beta[1]	0.9983	8.692e-11	1.149e+10	0.000	[0.998, 0.998]

Covariance estimator: robust

Daily Returns for GBPUSD=X EGARCH(1,1)**Conditional Volatility for GBPUSD=X EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-395.415
Distribution:	Normal	AIC:	800.830
Method:	Maximum Likelihood	BIC:	822.175
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:46	Df Model:	1

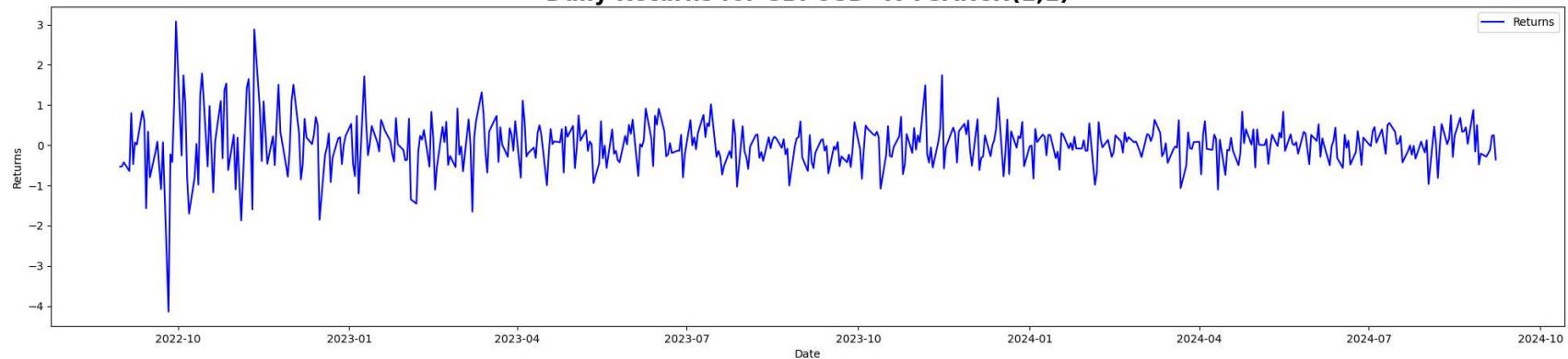
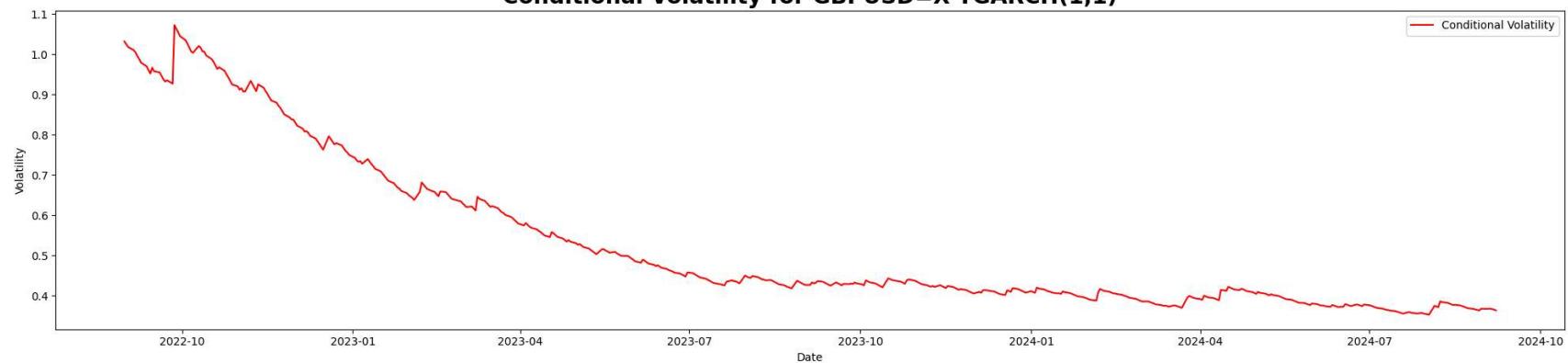
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0164	2.082e-02	0.790	0.430	[-2.437e-02,5.725e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.5388e-03	1.960e-03	0.785	0.432	[-2.303e-03,5.380e-03]
alpha[1]	2.6930e-18	5.383e-02	5.003e-17	1.000	[-0.105, 0.105]
gamma[1]	0.0178	1.613e-02	1.101	0.271	[-1.386e-02,4.938e-02]
beta[1]	0.9799	6.011e-02	16.303	9.451e-60	[0.862, 1.098]

Covariance estimator: robust

Daily Returns for GBPUSD=X TGARCH(1,1)**Conditional Volatility for GBPUSD=X TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-401.060
Distribution:	Normal	AIC:	812.120
Method:	Maximum Likelihood	BIC:	833.465
		No. Observations:	528
Date:	Sat, Sep 07 2024	Df Residuals:	527
Time:	20:05:47	Df Model:	1

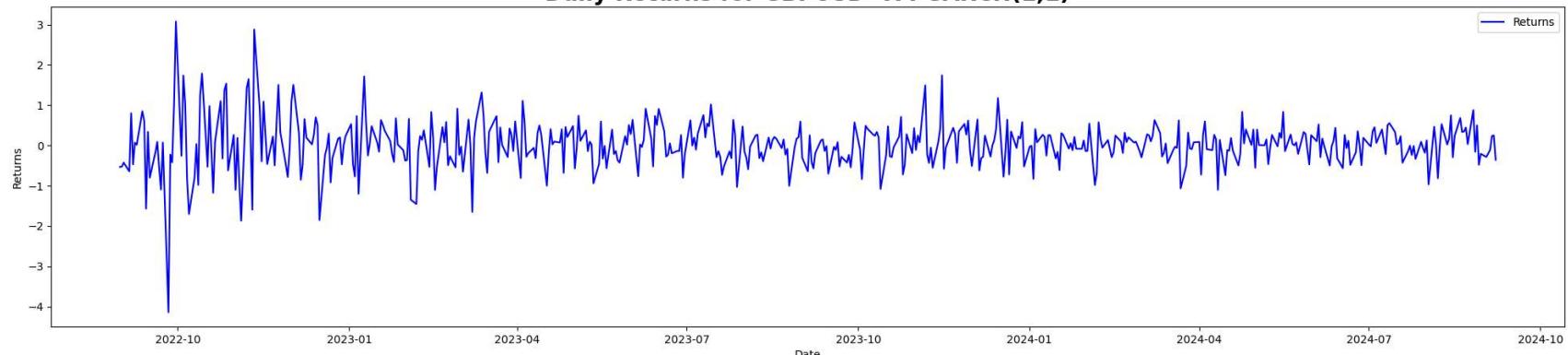
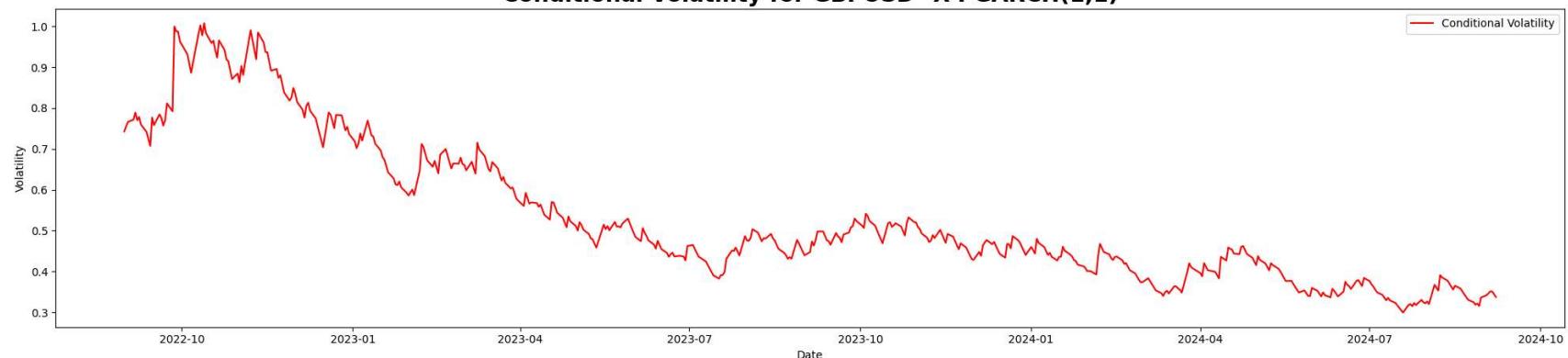
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	9.5323e-03	2.103e-02	0.453	0.650	[-3.168e-02,5.074e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	2.7098e-03	2.587e-03	1.047	0.295	[-2.361e-03,7.780e-03]
alpha[1]	1.2030e-12	3.498e-02	3.439e-11	1.000	[-6.856e-02,6.856e-02]
gamma[1]	0.0545	2.896e-02	1.880	6.005e-02	[-2.304e-03, 0.111]
beta[1]	0.9728	2.386e-02	40.763	0.000	[0.926, 1.020]

Covariance estimator: robust

Daily Returns for GBPUSD=X PGARCH(1,1)**Conditional Volatility for GBPUSD=X PGARCH(1,1)****Model Comparisons:**

ARCH AIC: 929.63885887227, BIC: 942.4461477233888

GARCH AIC: 800.4518750673487, BIC: 817.5282602021738

EGARCH AIC: 783.7052696374003, BIC: 800.7816547722254

TGARCH AIC: 800.8297879626749, BIC: 822.1752693812062

PGARCH AIC: 812.1196711901466, BIC: 833.465152608678

Best model based on BIC: EGARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 1.1437886905815706

GARCH - Volatility Forecasting MSE: 1.0010917673953181

EGARCH - Volatility Forecasting MSE: 0.9842808459673964

TGARCH - Volatility Forecasting MSE: 0.9898297338822423

PGARCH - Volatility Forecasting MSE: 0.9995394824228735

Best model based on volatility forecasting (lowest mse): EGARCH

The results show that for USD/ZAR and GBP/USD, EGARCH was the best model based on both the Bayesian Information Criterion (BIC) and volatility forecasting. This suggests that these currency pairs exhibit strong asymmetric volatility, where negative shocks, such as bad news, have a greater impact on volatility than positive shocks, making EGARCH's ability to capture this leverage effect particularly effective for both model selection and volatility prediction. For AUD/USD, while EGARCH was the best model according to BIC, TGARCH performed better in volatility forecasting. This indicates that while EGARCH captured the overall volatility structure well, TGARCH's handling of threshold effects provided a better forecast of future volatility. This may reflect that in the AUD/USD market, negative shocks have a different but more subtle impact on volatility that TGARCH captures more effectively in short-term forecasting.

Question 2

```
In [ ]: stock_tickers = ['TCS.NS', 'COALINDIA.NS', 'BAJFINANCE.NS']
stock_data = {ticker: yf.download(ticker, start=start_date, end=end_date) for ticker in stock_tickers}
```

```
[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed
```

```
In [ ]: for ticker, data in stock_data.items():
    print(f"Data for {ticker}")
    print(data.head())
```

Data for TCS.NS

	Open	High	Low	Close	Adj Close	\
Date						
2022-08-30	3155.000000	3226.500000	3142.100098	3211.149902	3063.410156	
2022-09-01	3190.000000	3190.000000	3121.000000	3131.699951	2987.615967	
2022-09-02	3163.000000	3163.000000	3120.300049	3130.399902	2986.375732	
2022-09-05	3123.649902	3147.949951	3112.250000	3133.399902	2989.237793	
2022-09-06	3135.500000	3140.850098	3106.350098	3127.050049	2983.179932	

Volume

Date

2022-08-30	3431525
2022-09-01	3546935
2022-09-02	2052900
2022-09-05	2147912
2022-09-06	1936453

Data for COALINDIA.NS

	Open	High	Low	Close	Adj Close	\
Date						
2022-08-30	231.800003	235.500000	230.850006	234.800003	198.011673	
2022-09-01	234.350006	236.800003	229.550003	230.300003	194.216736	
2022-09-02	231.000000	232.449997	228.199997	229.300003	193.373413	
2022-09-05	230.000000	232.350006	228.500000	231.600006	195.313049	
2022-09-06	232.699997	234.649994	231.000000	232.699997	196.240707	

Volume

Date

2022-08-30	12991363
2022-09-01	9682321
2022-09-02	4599518
2022-09-05	4909470
2022-09-06	5307321

Data for BAJFINANCE.NS

	Open	High	Low	Close	Adj Close	\
Date						
2022-08-30	7000.000000	7335.000000	7000.000000	7306.250000	7238.995117	
2022-09-01	7228.600098	7333.850098	7142.000000	7181.299805	7115.194824	
2022-09-02	7217.100098	7313.000000	7173.450195	7190.350098	7124.162109	
2022-09-05	7190.700195	7254.000000	7158.000000	7196.200195	7129.958496	
2022-09-06	7228.750000	7255.000000	7105.000000	7119.350098	7053.815430	

Volume

Date
2022-08-30 1731381
2022-09-01 1462579
2022-09-02 906370
2022-09-05 707025
2022-09-06 791905

```
In [ ]: for stock_name, df in stock_data.items():
    returns = calculate_returns(df)
    evaluate_and_visualize_models(returns, stock_name)
```

*****processing TCS.NS*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

```
=====
Dep. Variable: Returns R-squared: 0.000
Mean Model: Constant Mean Adj. R-squared: 0.000
Vol Model: ARCH Log-Likelihood: -806.307
Distribution: Normal AIC: 1618.61
Method: Maximum Likelihood BIC: 1631.24
                    No. Observations: 497
Date: Sat, Sep 07 2024 Df Residuals: 496
Time: 20:10:02 Df Model: 1
               Mean Model
=====
```

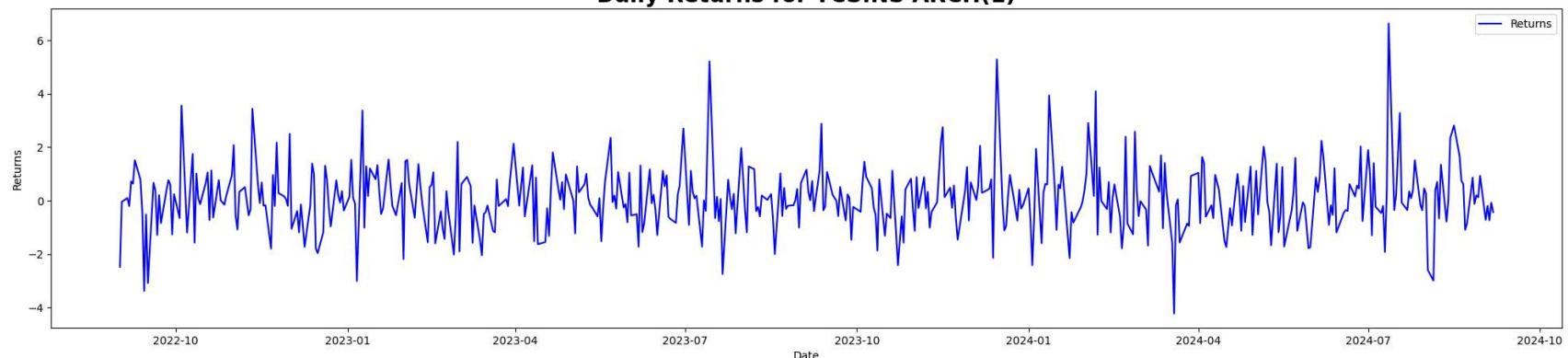
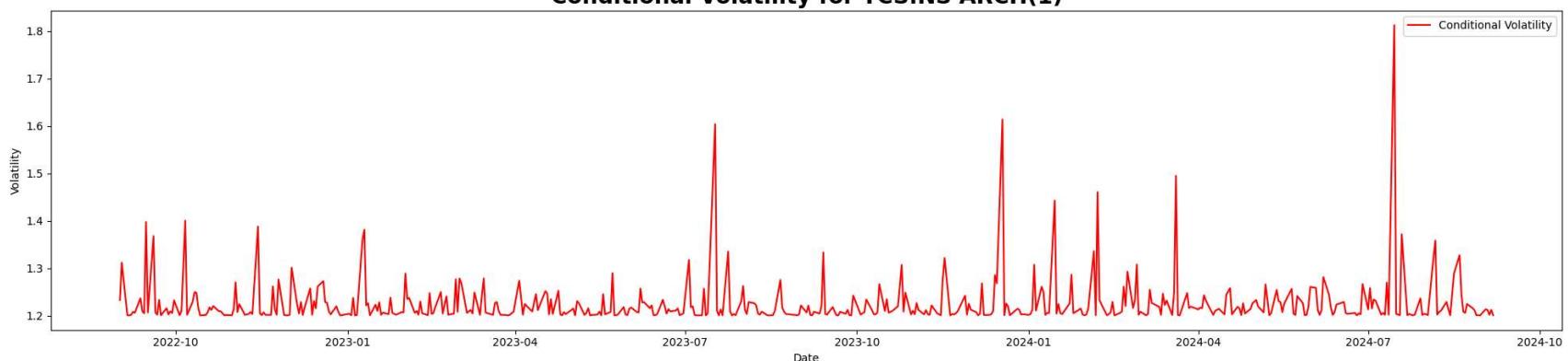
	coef	std err	t	P> t	95.0% Conf. Int.
--	------	---------	---	------	------------------

```
mu 0.0795 5.588e-02 1.423 0.155 [-3.000e-02, 0.189]
      Volatility Model
=====
```

	coef	std err	t	P> t	95.0% Conf. Int.
--	------	---------	---	------	------------------

```
omega 1.4425 0.250 5.775 7.684e-09 [ 0.953, 1.932]
alpha[1] 0.0429 0.162 0.264 0.792 [-0.275, 0.361]
=====
```

Covariance estimator: robust

Daily Returns for TCS.NS ARCH(1)**Conditional Volatility for TCS.NS ARCH(1)**

Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-806.047
Distribution:	Normal	AIC:	1620.09
Method:	Maximum Likelihood	BIC:	1636.93
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:04	Df Model:	1

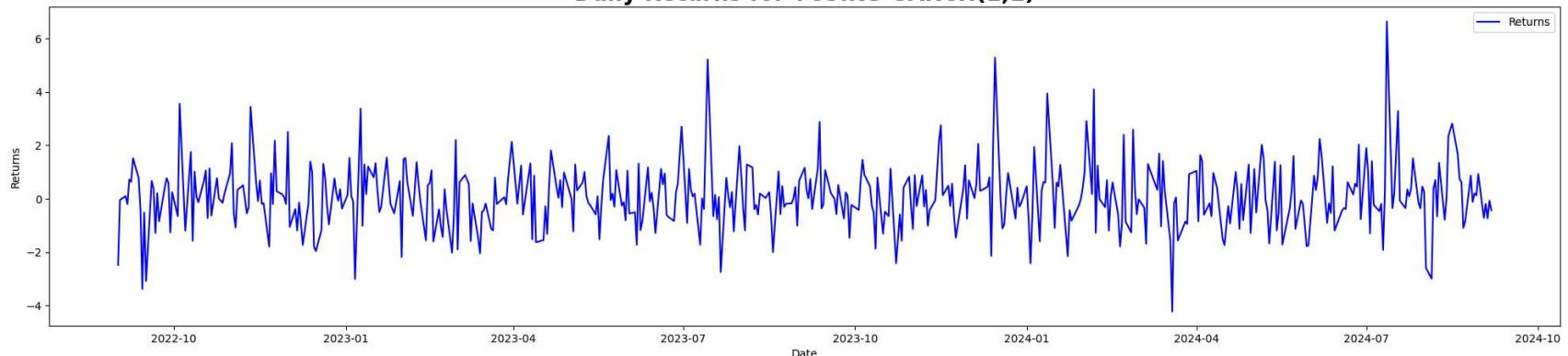
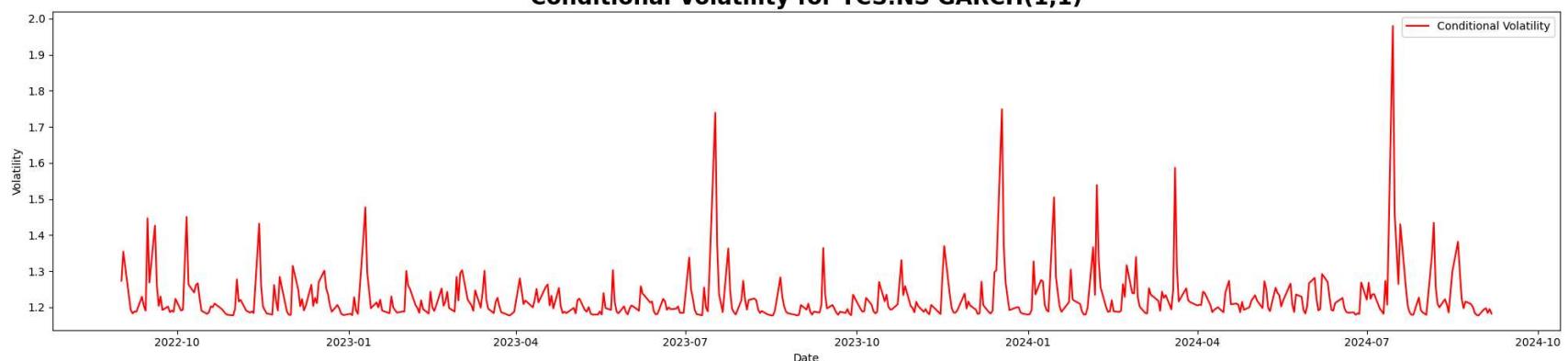
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0814	5.421e-02	1.502	0.133	[-2.484e-02, 0.188]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.9842	0.348	2.831	4.639e-03	[0.303, 1.666]
alpha[1]	0.0586	0.149	0.393	0.695	[-0.234, 0.351]
beta[1]	0.2885	0.259	1.115	0.265	[-0.219, 0.796]

Covariance estimator: robust

Daily Returns for TCS.NS GARCH(1,1)**Conditional Volatility for TCS.NS GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	EGARCH	Log-Likelihood:	-805.792
Distribution:	Normal	AIC:	1619.58
Method:	Maximum Likelihood	BIC:	1636.42
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:05	Df Model:	1

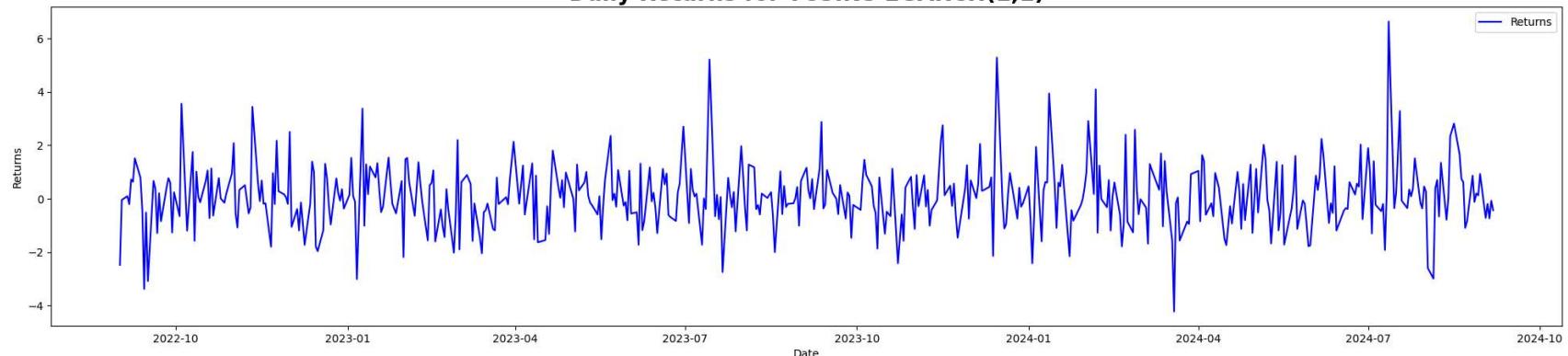
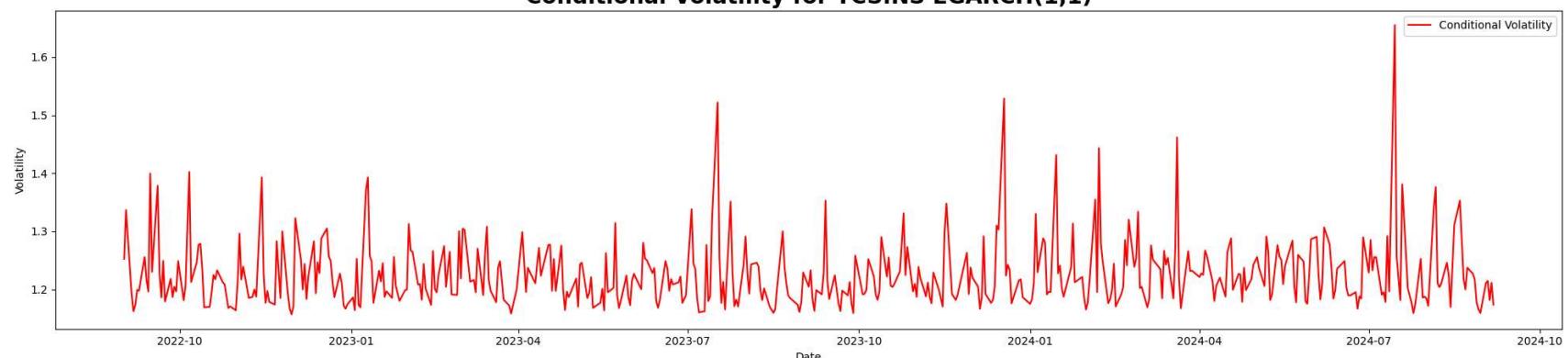
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0756	4.520e-03	16.729	8.061e-63	[6.675e-02, 8.447e-02]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.3340	0.130	2.571	1.013e-02	[7.940e-02, 0.589]
alpha[1]	0.1293	0.211	0.613	0.540	[-0.284, 0.543]
beta[1]	0.1913	0.318	0.602	0.547	[-0.432, 0.814]

Covariance estimator: robust

Daily Returns for TCS.NS EGARCH(1,1)**Conditional Volatility for TCS.NS EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-805.305
Distribution:	Normal	AIC:	1620.61
Method:	Maximum Likelihood	BIC:	1641.65
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:06	Df Model:	1

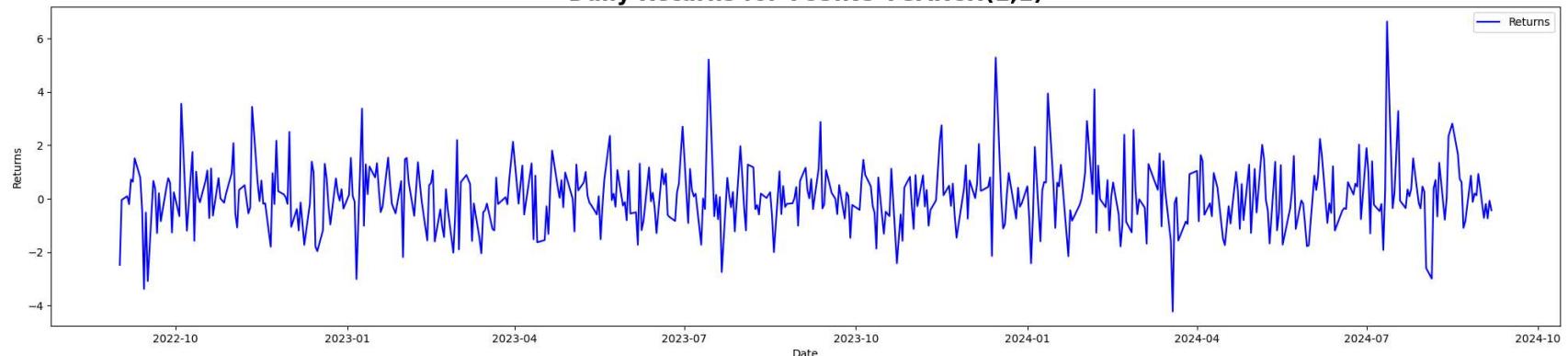
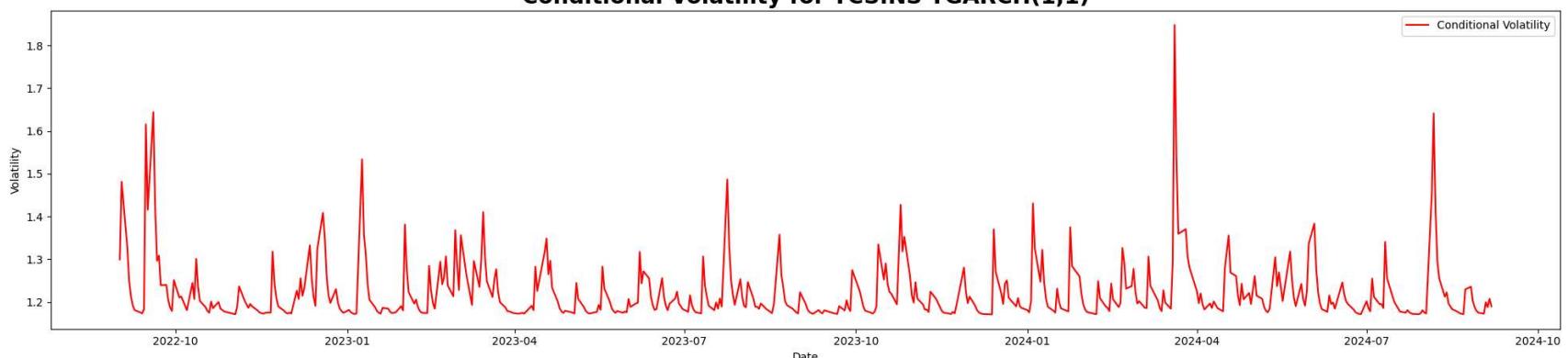
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0796	5.711e-02	1.393	0.163	[-3.235e-02, 0.191]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.7106	2.816	0.252	0.801	[-4.808, 6.229]
alpha[1]	0.0000	1.202	0.000	1.000	[-2.357, 2.357]
gamma[1]	0.1027	0.565	0.182	0.856	[-1.005, 1.211]
beta[1]	0.4819	2.790	0.173	0.863	[-4.986, 5.950]

Covariance estimator: robust

Daily Returns for TCS.NS TGARCH(1,1)**Conditional Volatility for TCS.NS TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-805.693
Distribution:	Normal	AIC:	1621.39
Method:	Maximum Likelihood	BIC:	1642.43
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:06	Df Model:	1

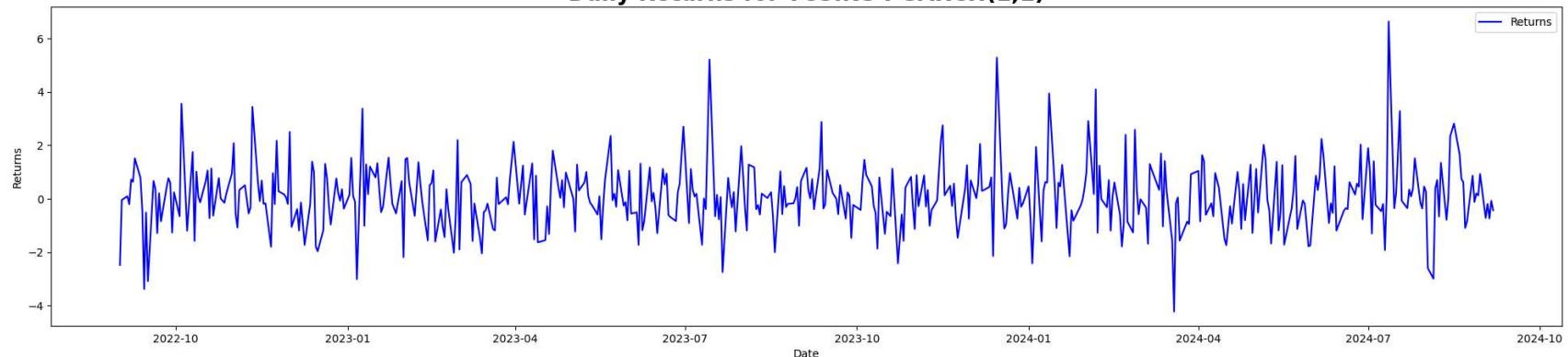
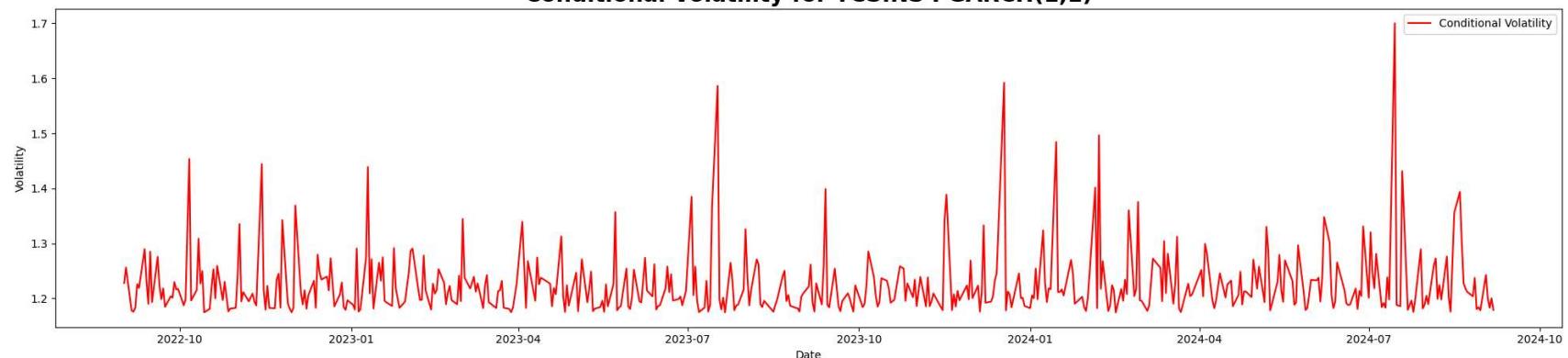
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0772	5.479e-02	1.408	0.159	[-3.022e-02, 0.185]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.1743	0.718	1.637	0.102	[-0.232, 2.581]
alpha[1]	0.0802	0.102	0.784	0.433	[-0.120, 0.281]
gamma[1]	-0.0481	5.002e-02	-0.962	0.336	[-0.146, 4.991e-02]
beta[1]	0.0000	0.569	0.000	1.000	[-1.115, 1.115]

Covariance estimator: robust

Daily Returns for TCS.NS PGARCH(1,1)**Conditional Volatility for TCS.NS PGARCH(1,1)**

Model Comparisons:

ARCH AIC: 1618.614274825084, BIC: 1631.240044903374
 GARCH AIC: 1620.09410527016, BIC: 1636.9284653745465
 EGARCH AIC: 1619.5847850549112, BIC: 1636.4191451592976
 TGARCH AIC: 1620.6102132194715, BIC: 1641.6531633499546
 PGARCH AIC: 1621.3856440470065, BIC: 1642.4285941774897

Best model based on BIC: ARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 11.579840539839495
 GARCH - Volatility Forecasting MSE: 11.594308901283053
 EGARCH - Volatility Forecasting MSE: 11.559982619434779
 TGARCH - Volatility Forecasting MSE: 11.557878604831322
 PGARCH - Volatility Forecasting MSE: 11.559177030413606

Best model based on volatility forecasting (lowest mse): TGARCH

*****processing COALINDIA.NS*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	ARCH	Log-Likelihood:	-980.589
Distribution:	Normal	AIC:	1967.18
Method:	Maximum Likelihood	BIC:	1979.80
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:07	Df Model:	1

Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
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mu	0.1861	7.348e-02	2.533	1.130e-02	[4.212e-02, 0.330]
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Volatility Model

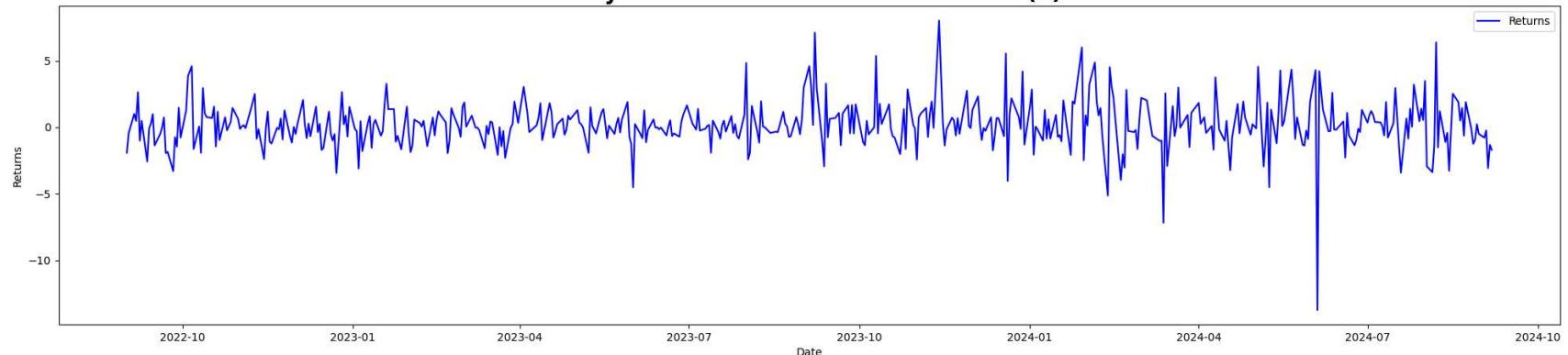
	coef	std err	t	P> t	95.0% Conf. Int.
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omega	2.4183	0.340	7.109	1.173e-12	[1.752, 3.085]
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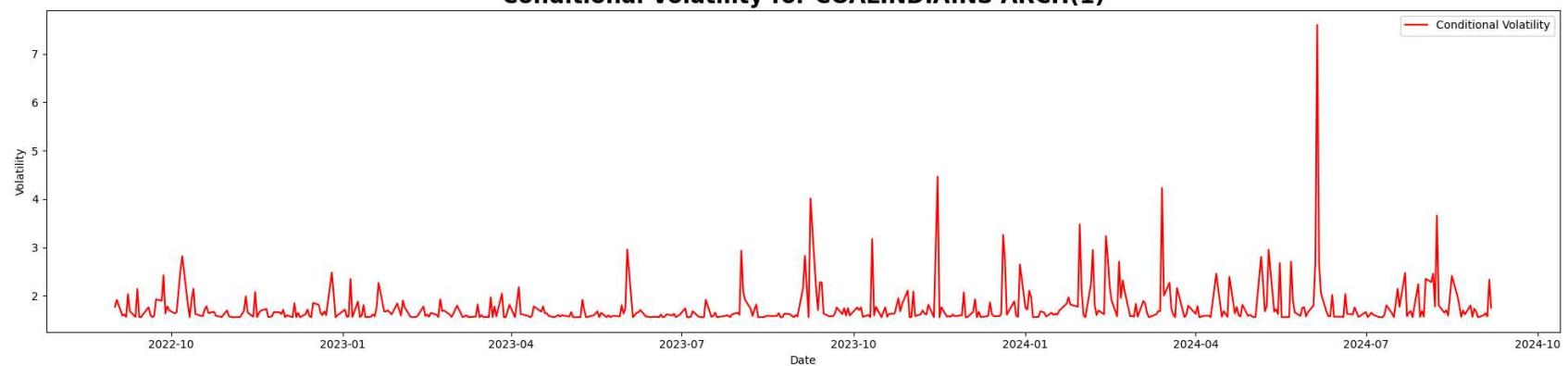
alpha[1]	0.2848	0.157	1.816	6.938e-02	[-2.259e-02, 0.592]
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Covariance estimator: robust

Daily Returns for COALINDIA.NS ARCH(1)



Conditional Volatility for COALINDIA.NS ARCH(1)



Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-977.805
Distribution:	Normal	AIC:	1963.61
Method:	Maximum Likelihood	BIC:	1980.44
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:07	Df Model:	1

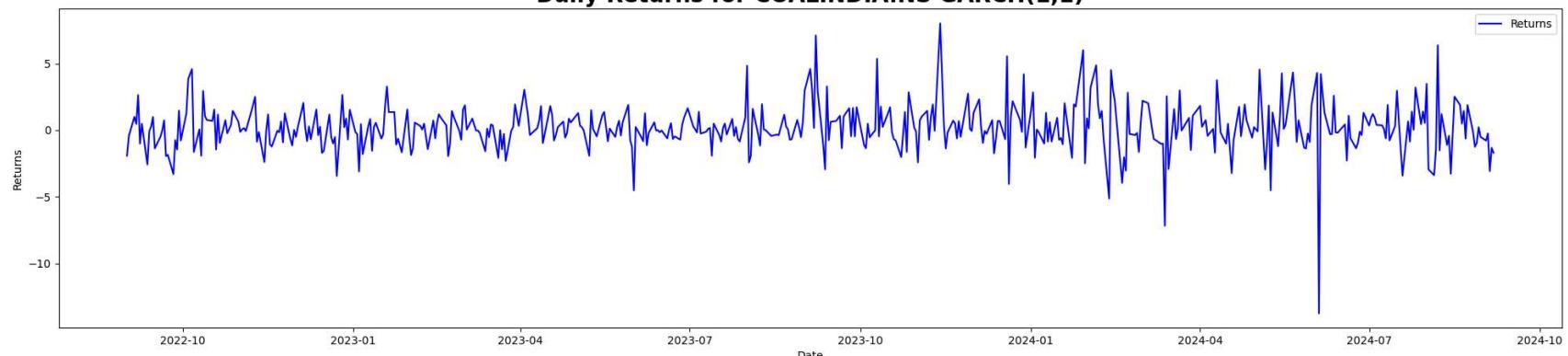
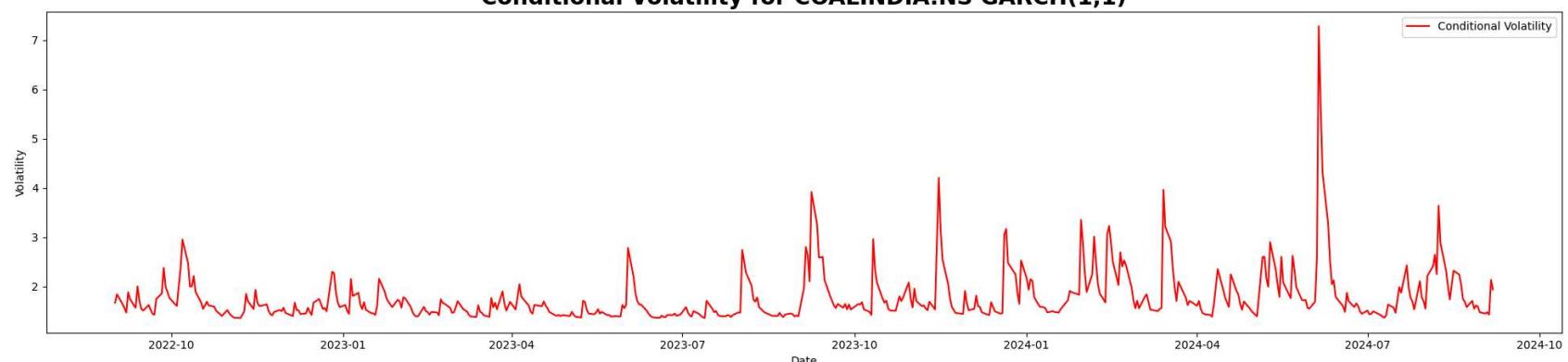
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.1577	7.219e-02	2.184	2.894e-02	[1.620e-02, 0.299]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.8934	0.510	1.751	7.989e-02	[-0.106, 1.893]
alpha[1]	0.2523	0.140	1.804	7.117e-02	[-2.175e-02, 0.526]
beta[1]	0.5070	0.187	2.707	6.782e-03	[0.140, 0.874]

Covariance estimator: robust

Daily Returns for COALINDIA.NS GARCH(1,1)**Conditional Volatility for COALINDIA.NS GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	EGARCH	Log-Likelihood:	-975.827
Distribution:	Normal	AIC:	1959.65
Method:	Maximum Likelihood	BIC:	1976.49
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:08	Df Model:	1

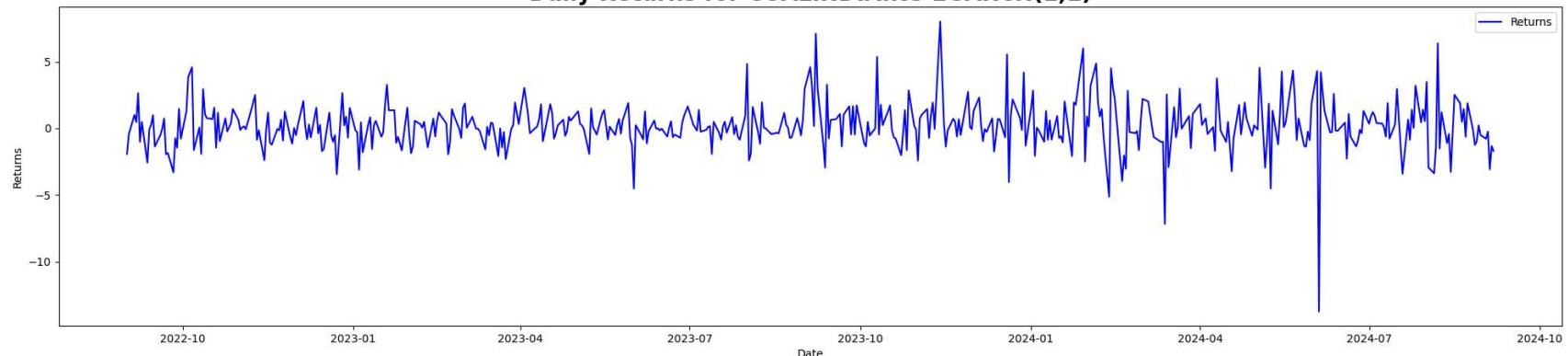
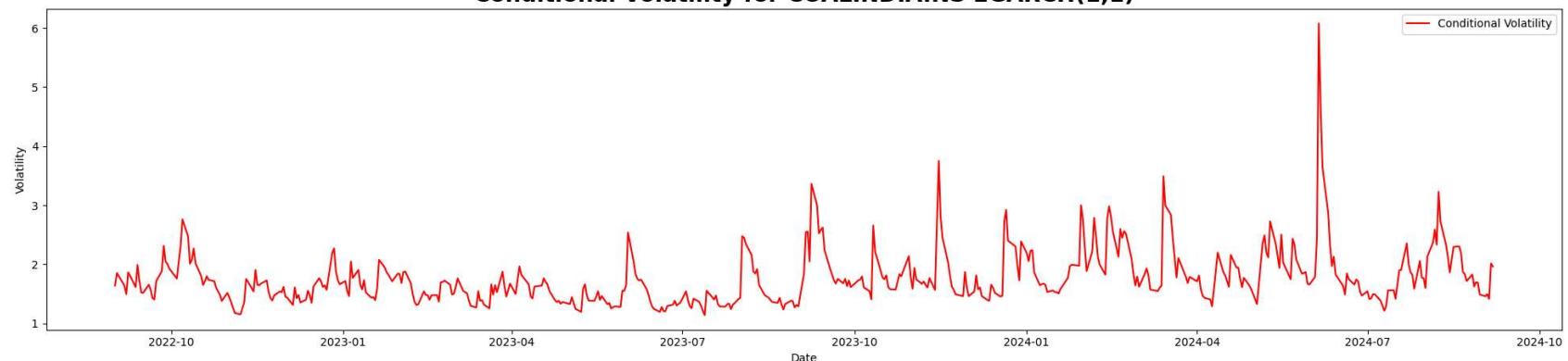
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.1578	6.968e-02	2.265	2.351e-02	[2.126e-02, 0.294]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.2705	0.135	2.000	4.547e-02	[5.450e-03, 0.536]
alpha[1]	0.3948	0.160	2.469	1.355e-02	[8.138e-02, 0.708]
beta[1]	0.7876	0.101	7.779	7.319e-15	[0.589, 0.986]

Covariance estimator: robust

Daily Returns for COALINDIA.NS EGARCH(1,1)**Conditional Volatility for COALINDIA.NS EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-973.653
Distribution:	Normal	AIC:	1957.31
Method:	Maximum Likelihood	BIC:	1978.35
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:08	Df Model:	1

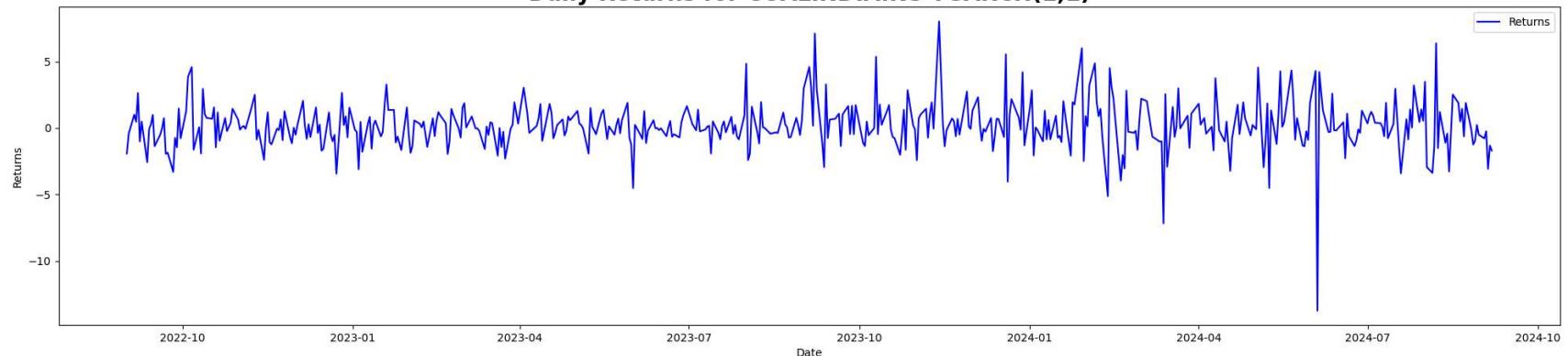
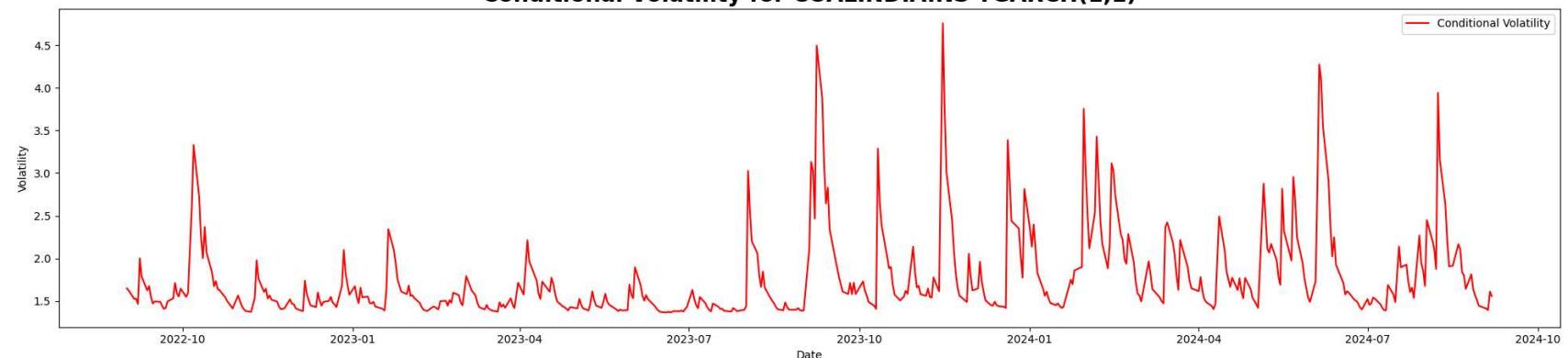
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.1966	7.083e-02	2.776	5.510e-03	[5.777e-02, 0.335]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.7662	0.373	2.055	3.989e-02	[3.539e-02, 1.497]
alpha[1]	0.3326	0.187	1.775	7.583e-02	[-3.457e-02, 0.700]
gamma[1]	-0.2668	0.175	-1.528	0.126	[-0.609, 7.542e-02]
beta[1]	0.5805	0.155	3.738	1.858e-04	[0.276, 0.885]

Covariance estimator: robust

Daily Returns for COALINDIA.NS TGARCH(1,1)**Conditional Volatility for COALINDIA.NS TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-972.694
Distribution:	Normal	AIC:	1955.39
Method:	Maximum Likelihood	BIC:	1976.43
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:09	Df Model:	1

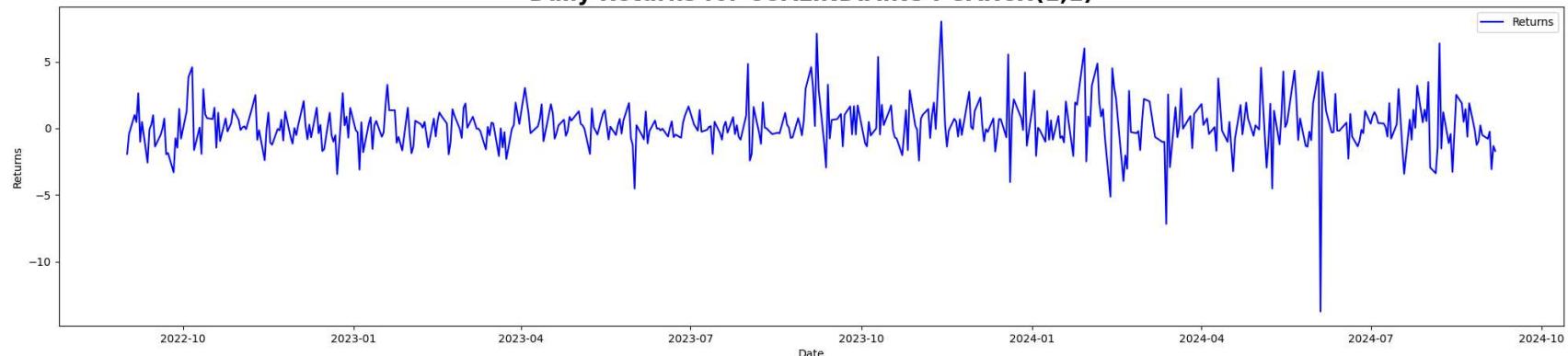
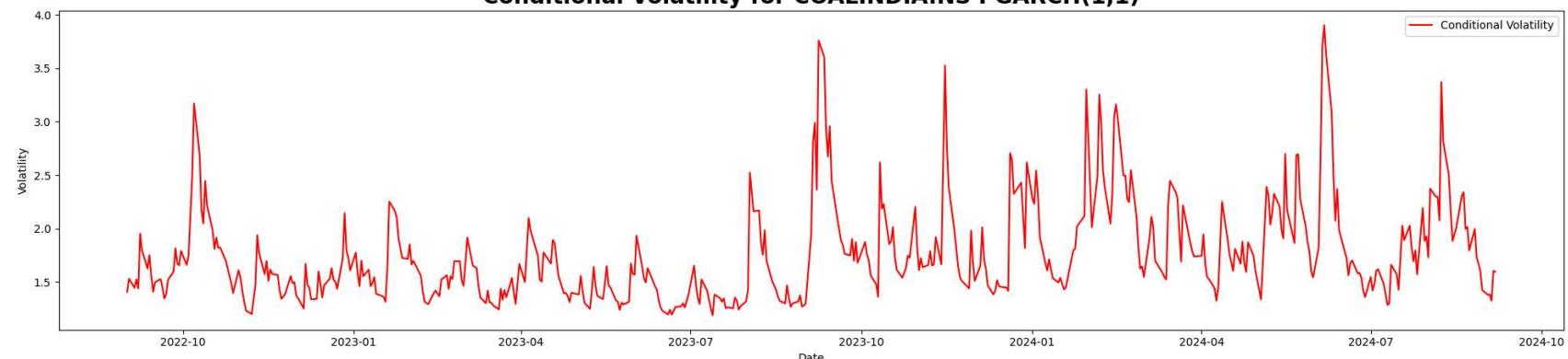
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.2099	7.676e-02	2.735	6.246e-03	[5.946e-02, 0.360]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.3375	0.190	1.779	7.531e-02	[-3.442e-02, 0.709]
alpha[1]	0.2638	0.128	2.068	3.860e-02	[1.384e-02, 0.514]
gamma[1]	-0.1509	0.111	-1.361	0.174	[-0.368, 6.645e-02]
beta[1]	0.6767	0.151	4.485	7.302e-06	[0.381, 0.972]

Covariance estimator: robust

Daily Returns for COALINDIA.NS PGARCH(1,1)**Conditional Volatility for COALINDIA.NS PGARCH(1,1)**

Model Comparisons:

ARCH AIC: 1967.177122648522, BIC: 1979.802892726812
 GARCH AIC: 1963.609092941013, BIC: 1980.4434530453993
 EGARCH AIC: 1959.6536192708732, BIC: 1976.4879793752596
 TGARCH AIC: 1957.3050204839146, BIC: 1978.3479706143978
 PGARCH AIC: 1955.3878971832896, BIC: 1976.4308473137728

Best model based on BIC: PGARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 112.95719562081153
 GARCH - Volatility Forecasting MSE: 114.66886125352477
 EGARCH - Volatility Forecasting MSE: 111.24486967124801
 TGARCH - Volatility Forecasting MSE: 110.93134570677911
 PGARCH - Volatility Forecasting MSE: 109.8635533889916

Best model based on volatility forecasting (lowest mse): PGARCH

*****processing BAJFINANCE.NS*****

Fitting ARCH(1) model...

Constant Mean - ARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	ARCH	Log-Likelihood:	-917.964
Distribution:	Normal	AIC:	1841.93
Method:	Maximum Likelihood	BIC:	1854.55
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:09	Df Model:	1

Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
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mu	6.5384e-03	6.820e-02	9.587e-02	0.924	[-0.127, 0.140]
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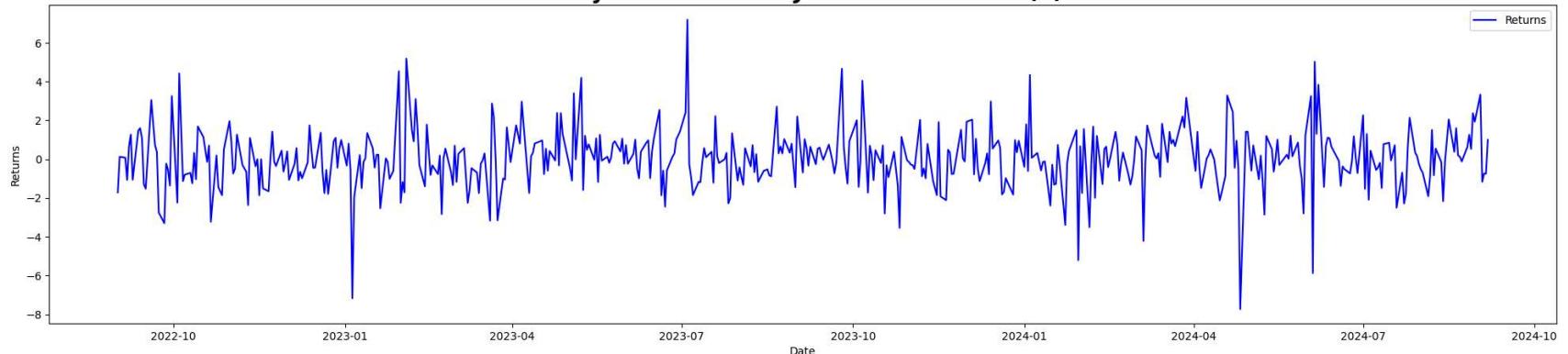
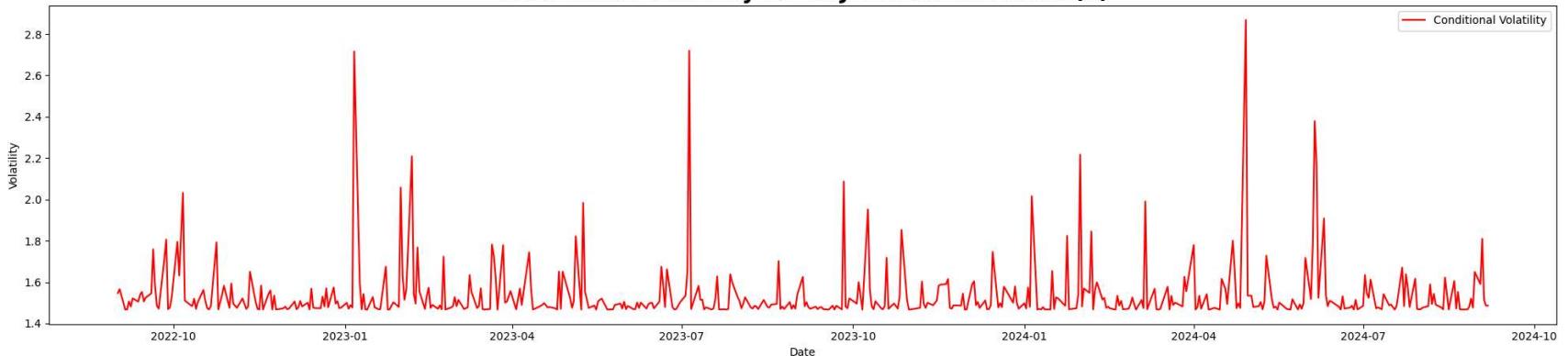
Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
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omega	2.1521	0.292	7.360	1.833e-13	[1.579, 2.725]
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alpha[1]	0.1016	9.375e-02	1.083	0.279	[-8.220e-02, 0.285]
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Covariance estimator: robust

Daily Returns for BAJFINANCE.NS ARCH(1)**Conditional Volatility for BAJFINANCE.NS ARCH(1)**

Fitting GARCH(1,1) model...

Constant Mean - GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	-917.736
Distribution:	Normal	AIC:	1843.47
Method:	Maximum Likelihood	BIC:	1860.31
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:10	Df Model:	1

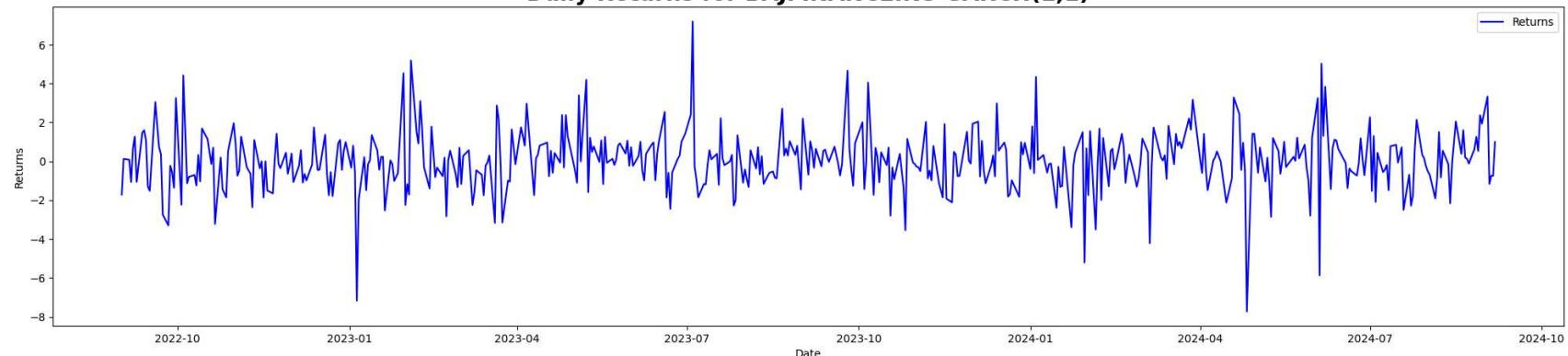
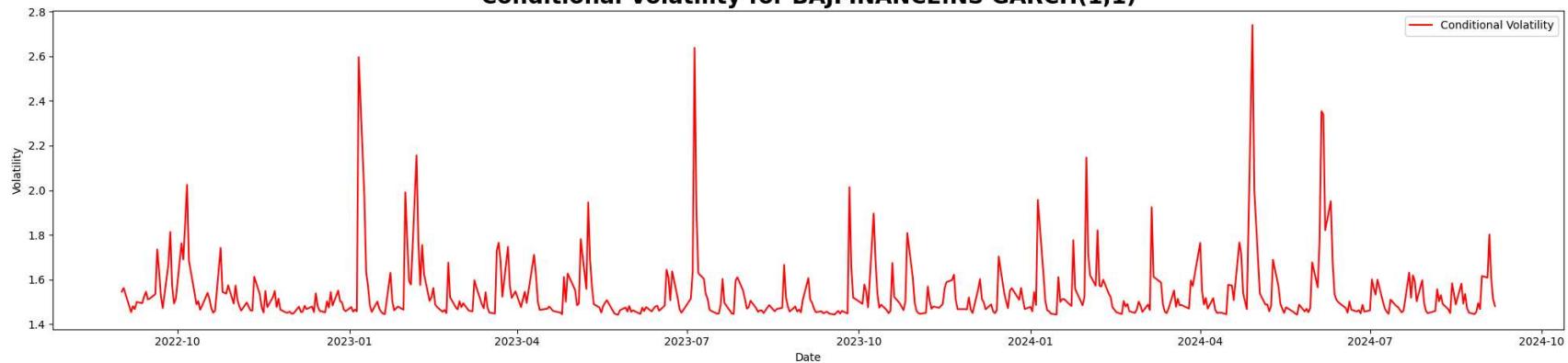
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
<hr/>					
mu	-3.2361e-03	6.847e-02	-4.726e-02	0.962	[-0.137, 0.131]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
<hr/>					
omega	1.4130	1.435	0.984	0.325	[-1.400, 4.226]
alpha[1]	0.0907	0.102	0.892	0.372	[-0.108, 0.290]
beta[1]	0.3190	0.649	0.491	0.623	[-0.954, 1.592]
<hr/>					

Covariance estimator: robust

Daily Returns for BAJFINANCE.NS GARCH(1,1)**Conditional Volatility for BAJFINANCE.NS GARCH(1,1)**

Fitting EGARCH(1,1) model...

Constant Mean - EGARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	EGARCH	Log-Likelihood:	-916.959
Distribution:	Normal	AIC:	1841.92
Method:	Maximum Likelihood	BIC:	1858.75
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:11	Df Model:	1

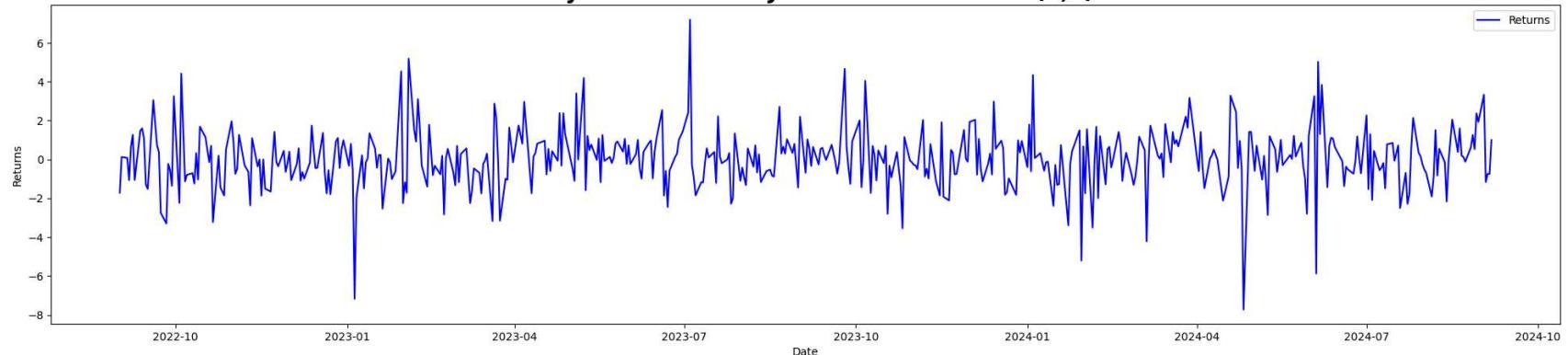
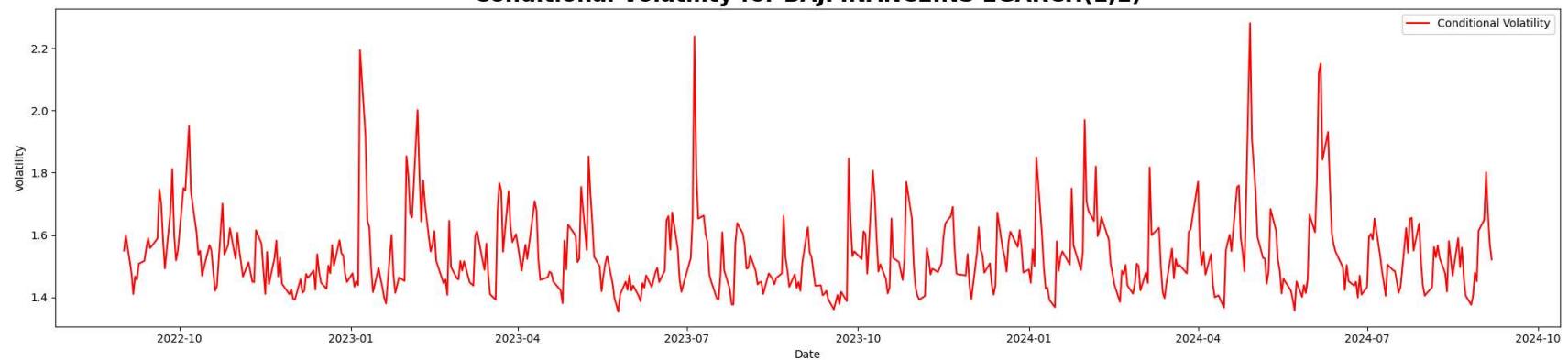
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0151	7.114e-02	0.213	0.831	[-0.124, 0.155]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.3765	0.242	1.558	0.119	[-9.705e-02, 0.850]
alpha[1]	0.1850	0.113	1.636	0.102	[-3.658e-02, 0.407]
beta[1]	0.5758	0.277	2.075	3.795e-02	[3.201e-02, 1.119]

Covariance estimator: robust

Daily Returns for BAJFINANCE.NS EGARCH(1,1)**Conditional Volatility for BAJFINANCE.NS EGARCH(1,1)**

Fitting TGARCH(1,1) model...

Constant Mean - GJR-GARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GJR-GARCH	Log-Likelihood:	-916.589
Distribution:	Normal	AIC:	1843.18
Method:	Maximum Likelihood	BIC:	1864.22
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:11	Df Model:	1

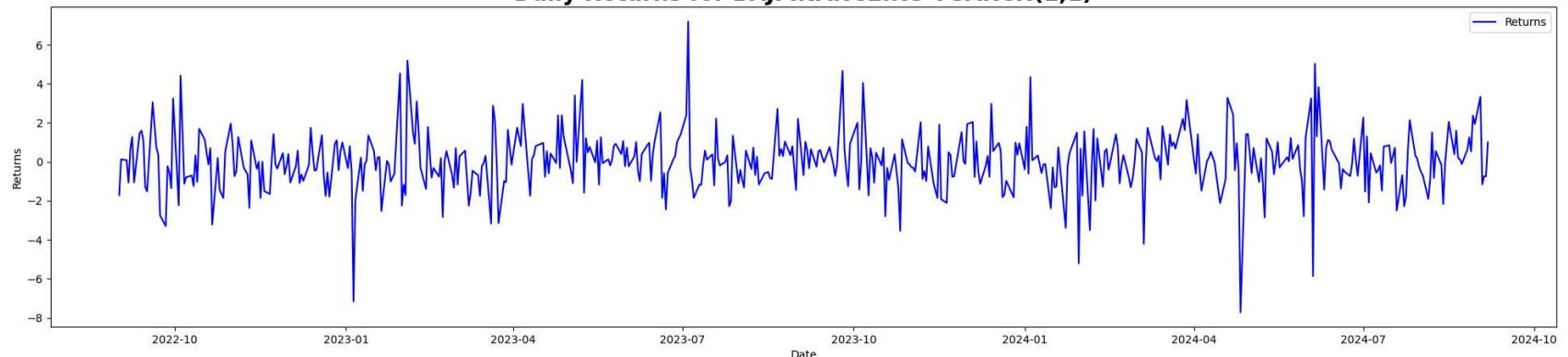
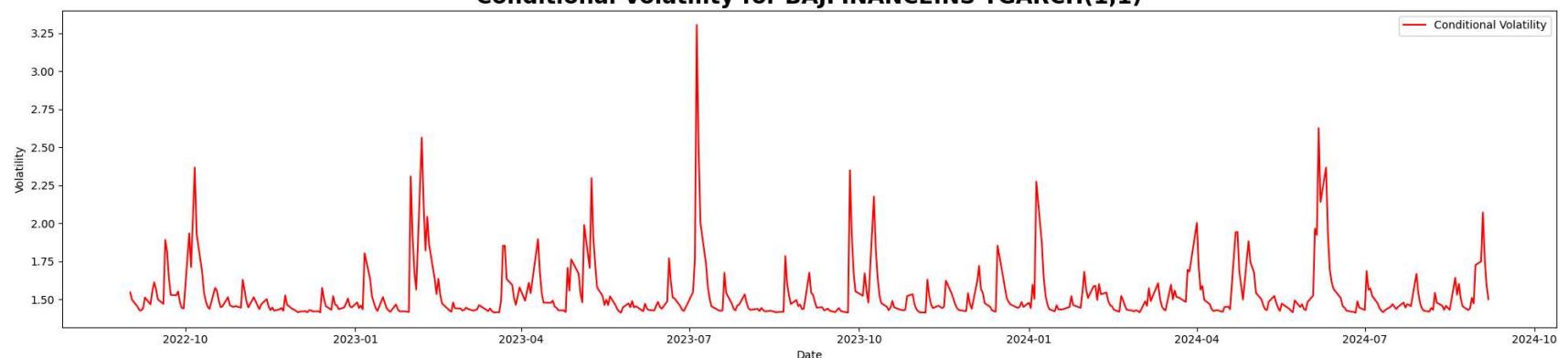
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	2.8055e-03	6.775e-02	4.141e-02	0.967	[-0.130, 0.136]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.0436	0.548	1.905	5.683e-02	[-3.031e-02, 2.118]
alpha[1]	0.1625	0.132	1.228	0.220	[-9.696e-02, 0.422]
gamma[1]	-0.1385	0.149	-0.933	0.351	[-0.430, 0.153]
beta[1]	0.4736	0.263	1.803	7.142e-02	[-4.129e-02, 0.989]

Covariance estimator: robust

Daily Returns for BAJFINANCE.NS TGARCH(1,1)**Conditional Volatility for BAJFINANCE.NS TGARCH(1,1)**

Fitting PGARCH(1,1) model...

Constant Mean - TARCH/ZARCH Model Results

Dep. Variable:	Returns	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	TARCH/ZARCH	Log-Likelihood:	-916.060
Distribution:	Normal	AIC:	1842.12
Method:	Maximum Likelihood	BIC:	1863.16
		No. Observations:	497
Date:	Sat, Sep 07 2024	Df Residuals:	496
Time:	20:10:12	Df Model:	1

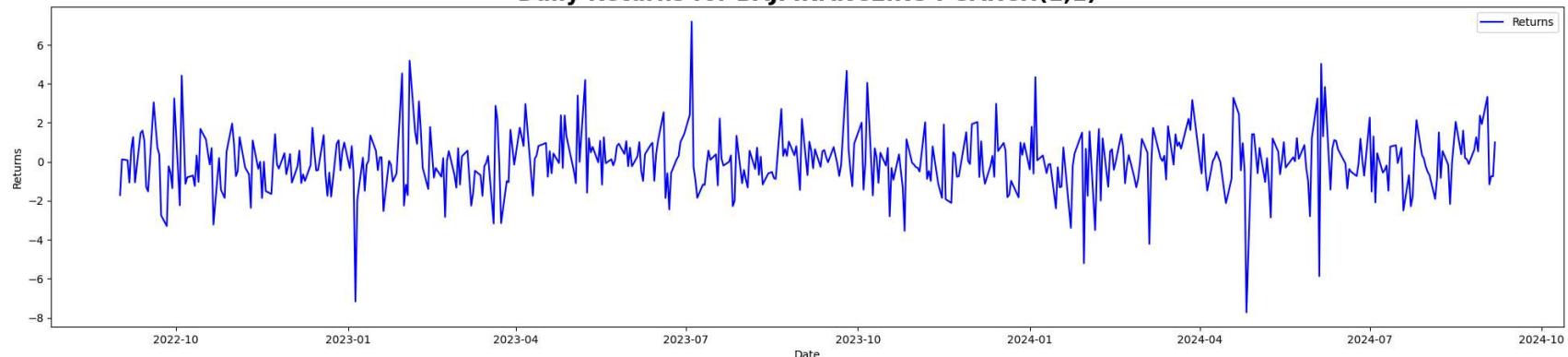
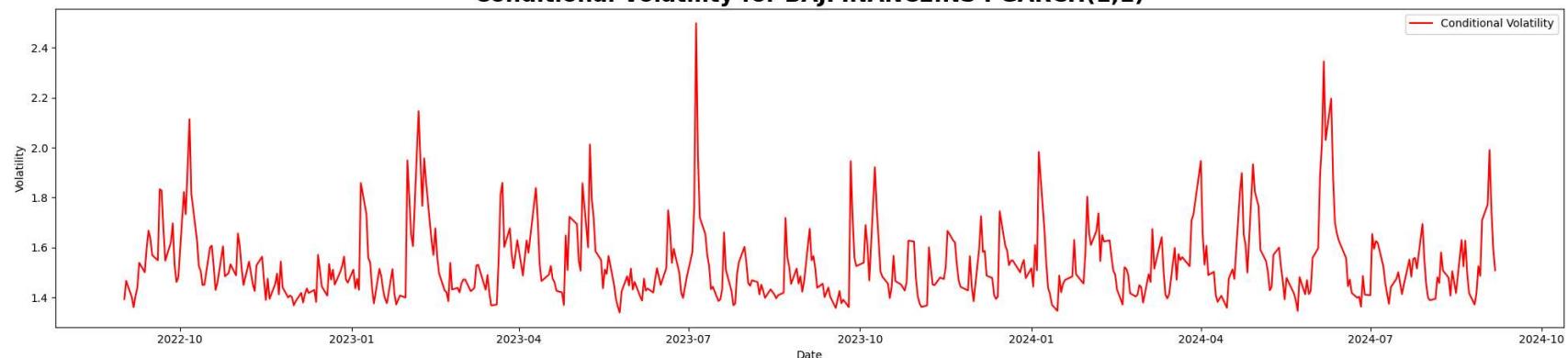
Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	0.0248	7.243e-02	0.342	0.732	[-0.117, 0.167]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.5995	0.295	2.034	4.195e-02	[2.182e-02, 1.177]
alpha[1]	0.1323	8.629e-02	1.533	0.125	[-3.680e-02, 0.301]
gamma[1]	-0.0643	9.608e-02	-0.670	0.503	[-0.253, 0.124]
beta[1]	0.5379	0.223	2.410	1.596e-02	[0.100, 0.975]

Covariance estimator: robust

Daily Returns for BAJFINANCE.NS PGARCH(1,1)**Conditional Volatility for BAJFINANCE.NS PGARCH(1,1)****Model Comparisons:**

ARCH AIC: 1841.9278396797808, BIC: 1854.5536097580707

GARCH AIC: 1843.4723447869778, BIC: 1860.3067048913642

EGARCH AIC: 1841.917238364929, BIC: 1858.7515984693155

TGARCH AIC: 1843.1787132736674, BIC: 1864.2216634041506

PGARCH AIC: 1842.120045353808, BIC: 1863.1629954842913

Best model based on BIC: ARCH

Comparing models based on volatility forecasting...

ARCH - Volatility Forecasting MSE: 32.0599508425907

GARCH - Volatility Forecasting MSE: 32.02554627106747

EGARCH - Volatility Forecasting MSE: 31.92403591329105

TGARCH - Volatility Forecasting MSE: 32.222930254099964

PGARCH - Volatility Forecasting MSE: 31.89102989564405

Best model based on volatility forecasting (lowest mse): PGARCH

The results indicate that different models perform best depending on the evaluation metric (BIC or volatility forecast) for each stock. For Bajaj Finance, while the ARCH model had the lowest BIC, indicating it fits the data well, PGARCH provided the best volatility forecast, suggesting it captures more nuances in volatility dynamics over time. In the case of Coal India, PGARCH was the best model both in terms of BIC and volatility forecasting, meaning it not only fits the data well but also excels at predicting future volatility. Finally, for TCS, the ARCH model had the lowest BIC, suggesting it offers a simpler, better-fitting model, but TGARCH produced better volatility forecasts, likely because TCS exhibits asymmetric volatility patterns that TGARCH is better equipped to capture.