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In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from statsmodels.tsa.arima.model import ARIMA
from statsmodels.stats.diagnostic import acorr_ljungbox
from sklearn.metrics import mean_absolute_error, mean_squared_error

# 1. Load dataset
# Replace with your actual file and column name
df = pd.read_excel(r"C:\Users\HP\OneDrive\Desktop\NLP\Data\ML470_S3_Diabetes_Dat

# Assume column name is 'Consumption'
series = df['HbA1c']
series.index = pd.date_range(start='2020-01-01', periods=len(series), freq='M')

train_size = int(len(series) * 0.8)
train, test = series[:train_size], series[train_size:]

model = ARIMA(train, order=(2, 1, 2), seasonal_order=(1, 1, 1, 12), enforce_stat
model_fit = model.fit()

print(model_fit.summary())

# 4. Forecast
forecast = model_fit.forecast(steps=len(test))
forecast = pd.Series(forecast, index=test.index)

# 5. Plot Forecast
plt.figure(figsize=(12,6))
plt.plot(train, label='Train')
plt.plot(test, label='Actual', linestyle='--')
plt.plot(forecast, label='Forecast', linestyle='--')
plt.title("ARIMA Forecast Plot")
plt.xlabel("Date")
plt.ylabel("Power Consumption")
plt.legend()
plt.grid()
plt.show()

# 6. Residual Analysis
residuals = model_fit.resid

plt.figure(figsize=(12,4))
plt.plot(residuals)
plt.title("Residuals")
plt.xlabel("Date")
plt.ylabel("Residual")
plt.grid()
plt.show()

# 7. Ljung-Box Test
ljung_box = acorr_ljungbox(residuals, lags=[10], return_df=True)
print("\nLjung-Box Test:")
print(ljung_box)

# 8. Accuracy Metrics
mae = mean_absolute_error(test, forecast)

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rmse = np.sqrt(mean_squared_error(test, forecast))
mape = np.mean(np.abs((test - forecast) / test)) * 100

print("\nForecast Accuracy:")
print("MAE :", mae)
print("RMSE:", rmse)
print("MAPE:", mape, "%")

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C:\Users\HP\AppData\Local\Temp\ipykernel_10440\4007929928.py:15: FutureWarning: 'M' is deprecated and will be removed in a future version, please use 'ME' instead.

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series.index = pd.date_range(start='2020-01-01', periods=len(series), freq='M')
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SARIMAX Results

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Dep. Variable: HbA1c No. Observations:

601

Model: ARIMA(2, 1, 2)x(1, 1, [1], 12) Log Likelihood

-841.486

Date: Fri, 30 Jan 2026 AIC

1696.971

Time: 13:09:09 BIC

1727.427

Sample: 01-31-2020 HQIC

1708.851

- 01-31-2070

Covariance Type: opg

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	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.9966	0.052	-19.053	0.000	-1.099	-0.894
ar.L2	-0.0777	0.045	-1.737	0.082	-0.165	0.010
ma.L1	-0.0462	0.039	-1.188	0.235	-0.122	0.030
ma.L2	-0.9615	0.038	-25.631	0.000	-1.035	-0.888
ar.S.L12	-4.23e-06	0.021	-0.000	1.000	-0.041	0.041
ma.S.L12	-1.0494	0.026	-40.390	0.000	-1.100	-0.999
sigma2	0.9456	0.090	10.490	0.000	0.769	1.122

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Ljung-Box (L1) (Q): 0.00 Jarque-Bera (JB): 31.28

Prob(Q): 0.97 Prob(JB): 0.00

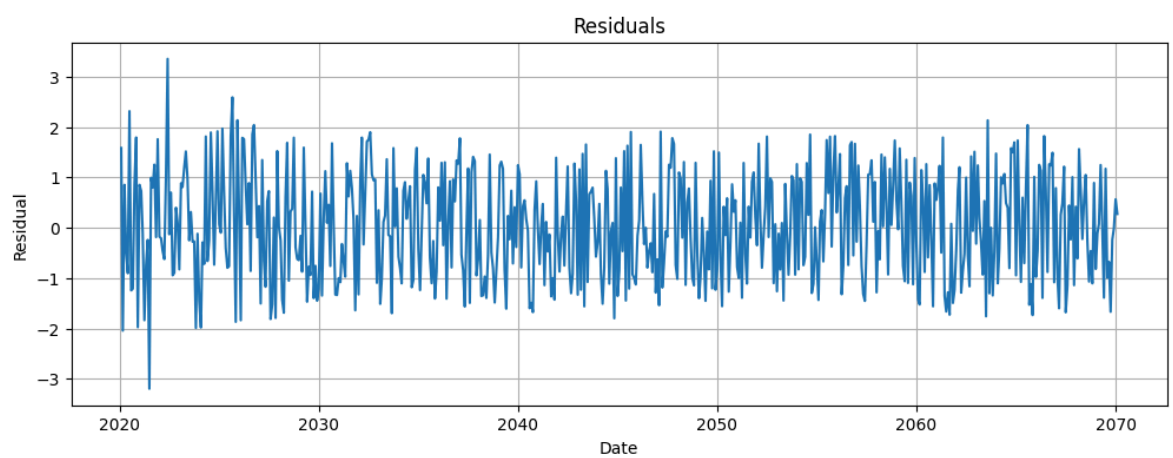
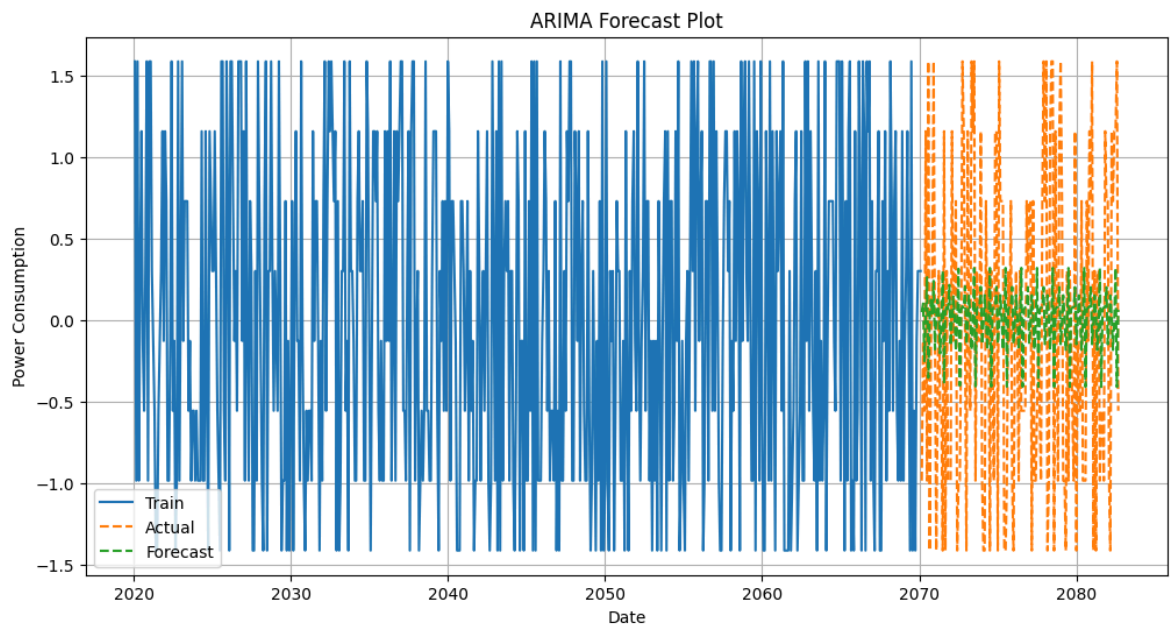
Heteroskedasticity (H): 1.03 Skew: 0.11

Prob(H) (two-sided): 0.82 Kurtosis: 1.87

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Warnings:

[1] Covariance matrix calculated using the outer product of gradients (complex-step).



Ljung-Box Test:

	lb_stat	lb_pvalue
10	4.578971	0.917474

Forecast Accuracy:

MAE	: 0.8669819466344655
RMSE	: 0.9921834162899332
MAPE	: 107.54848366520405 %