

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

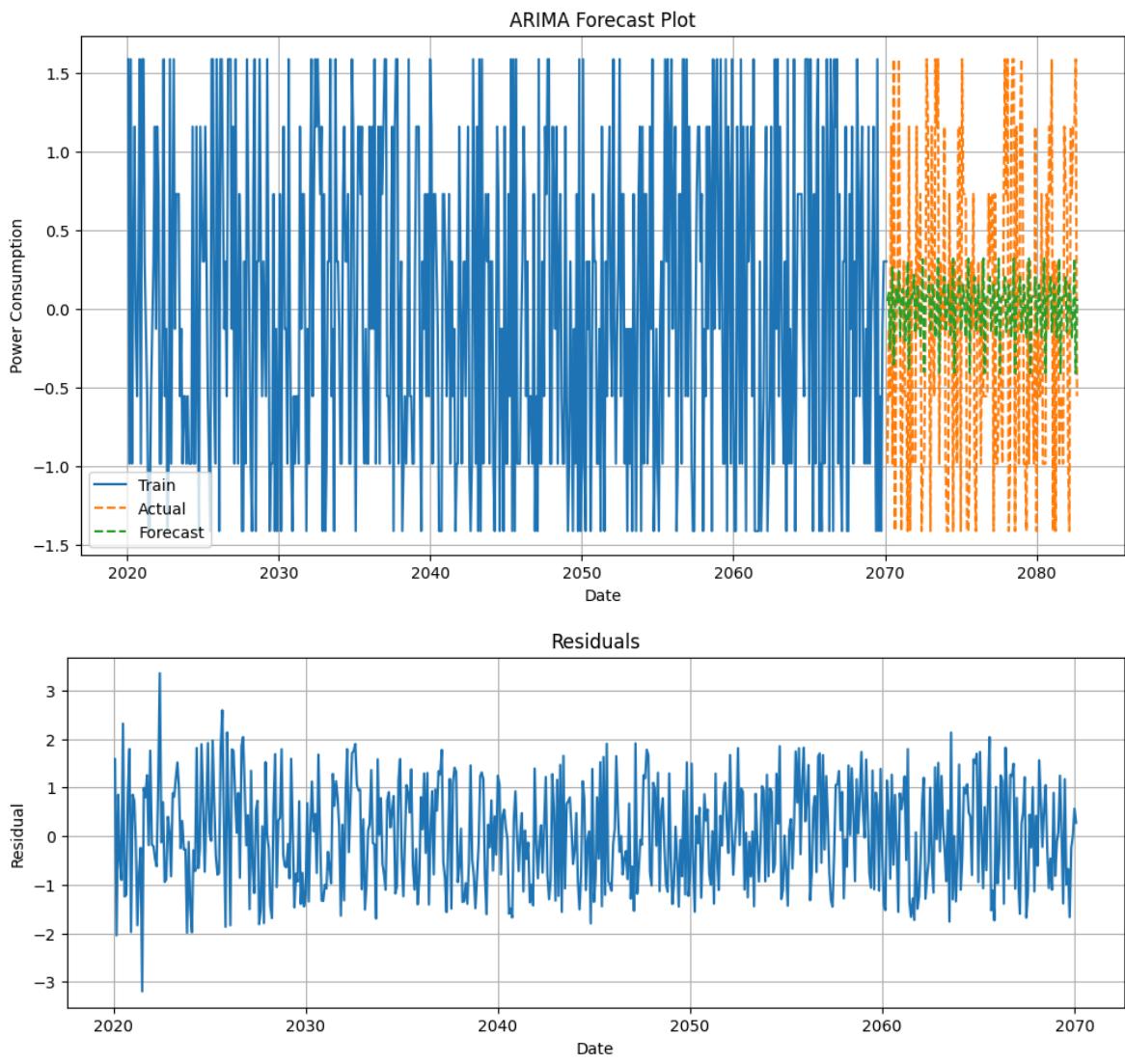
print("\nForecast Accuracy:")
print("MAE : ", mae)
print("RMSE:", rmse)
print("MAPE:", mape, "%")

```

```

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