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In [1]: import pandas as pd
import matplotlib.pyplot as plt
from statsmodels.tsa.seasonal import seasonal_decompose
from statsmodels.tsa.stattools import adfuller
import plotly.express as px
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
from statsmodels.tsa.statespace.sarimax import SARIMAX
from sklearn.metrics import mean_absolute_error

import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: data = pd.read_csv('Temperatures.csv', encoding='latin-1')
data
```

Out[2]:	Area Code	Area	Months Code	Months	Element Code	Element	Unit	Y1961	Y1962	Y1963	...	Y2010	Y2011	Y2012	Y2013	Y2014
0	2	Afghanistan	7001	January	7271	Temperature change	°C	0.777	0.062	2.744	...	3.601	1.179	-0.583	1.233	1.759
1	2	Afghanistan	7001	January	6078	Standard Deviation	°C	1.950	1.950	1.950	...	1.950	1.950	1.950	1.950	1.950
2	2	Afghanistan	7002	February	7271	Temperature change	°C	-1.743	2.465	3.919	...	1.212	0.321	-3.201	1.494	-3.185
3	2	Afghanistan	7002	February	6078	Standard Deviation	°C	2.597	2.597	2.597	...	2.597	2.597	2.597	2.597	2.597
4	2	Afghanistan	7003	March	7271	Temperature change	°C	0.516	1.336	0.403	...	3.390	0.748	-0.527	2.246	-0.076
...
9651	5873	OECD	7018	Jun□Jul□Aug	6078	Standard Deviation	°C	0.247	0.247	0.247	...	0.247	0.247	0.247	0.247	0.247
9652	5873	OECD	7019	Sep□Oct□Nov	7271	Temperature change	°C	0.036	0.461	0.665	...	0.958	1.106	0.885	1.041	0.999
9653	5873	OECD	7019	Sep□Oct□Nov	6078	Standard Deviation	°C	0.378	0.378	0.378	...	0.378	0.378	0.378	0.378	0.378
9654	5873	OECD	7020	Meteorological year	7271	Temperature change	°C	0.165	-0.009	0.134	...	1.246	0.805	1.274	0.991	0.811
9655	5873	OECD	7020	Meteorological year	6078	Standard Deviation	°C	0.260	0.260	0.260	...	0.260	0.260	0.260	0.260	0.260

9656 rows × 66 columns



```
In [3]: year_columns = [col for col in data.columns if col.startswith('Y')]
temperatures = data[year_columns].mean()

data = temperatures.reset_index()
data.columns = ['Year', 'Average Temperature']
data['Year'] = data['Year'].str[1:]
```

```
data.to_csv("average_temperatures.csv", index=False)  
print(data)
```

	Year	Average Temperature
0	1961	0.402433
1	1962	0.315527
2	1963	0.317393
3	1964	0.269382
4	1965	0.217839
5	1966	0.376419
6	1967	0.263239
7	1968	0.244870
8	1969	0.382172
9	1970	0.365322
10	1971	0.240934
11	1972	0.302553
12	1973	0.427691
13	1974	0.261849
14	1975	0.314653
15	1976	0.221112
16	1977	0.422978
17	1978	0.355488
18	1979	0.442465
19	1980	0.438270
20	1981	0.437693
21	1982	0.404857
22	1983	0.503748
23	1984	0.366971
24	1985	0.365511
25	1986	0.398096
26	1987	0.535514
27	1988	0.546662
28	1989	0.469231
29	1990	0.621797
30	1991	0.499991
31	1992	0.447798
32	1993	0.439094
33	1994	0.611078
34	1995	0.635836
35	1996	0.477239
36	1997	0.617341
37	1998	0.818264
38	1999	0.704445
39	2000	0.674191
40	2001	0.741673
41	2002	0.802509
42	2003	0.769485
43	2004	0.726237

44	2005	0.777465
45	2006	0.791795
46	2007	0.842554
47	2008	0.742614
48	2009	0.814177
49	2010	0.884504
50	2011	0.768488
51	2012	0.788930
52	2013	0.829647
53	2014	0.913872
54	2015	1.018816
55	2016	1.081491
56	2017	1.003342
57	2018	1.010832
58	2019	1.094599

```
In [4]: data.set_index("Year", inplace=True)

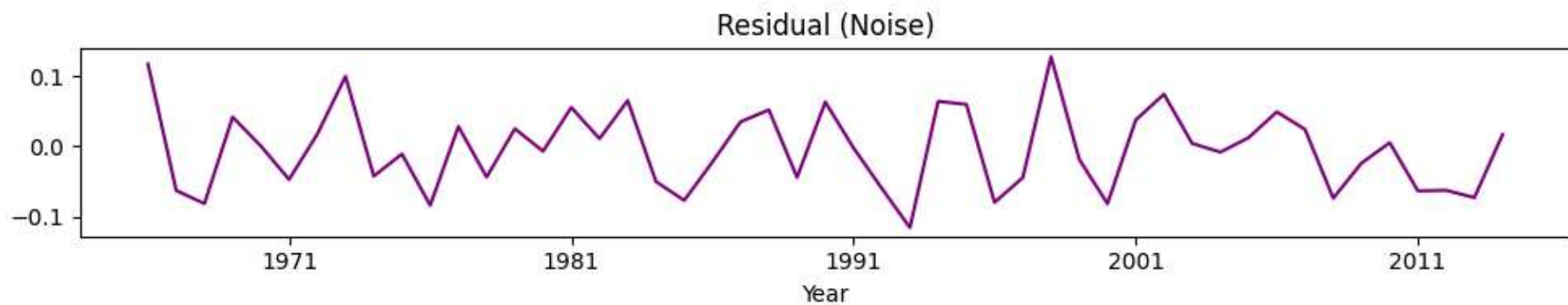
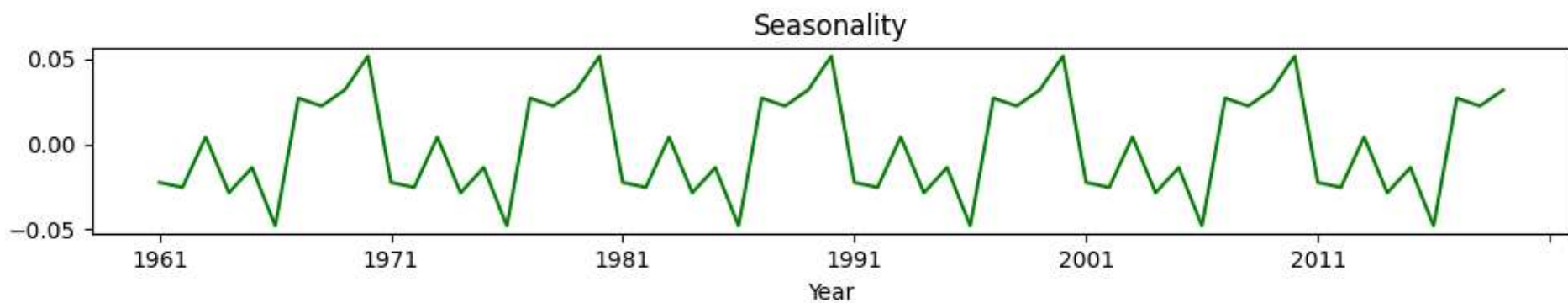
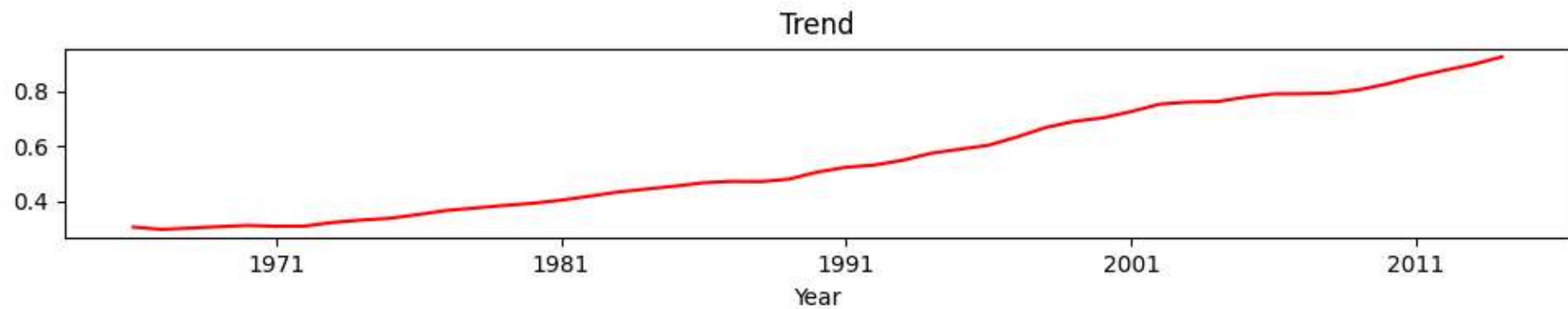
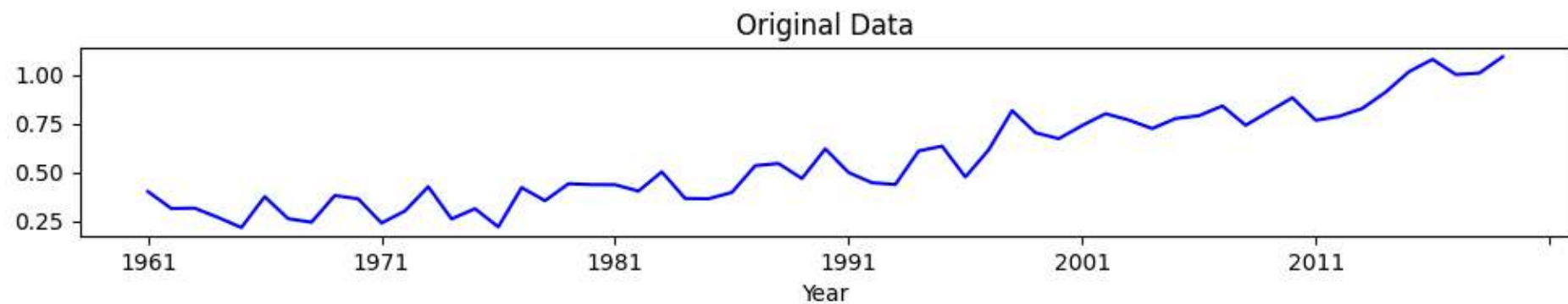
decomposition = seasonal_decompose(data["Average Temperature"], model="additive", period=10)

fig, (ax1, ax2, ax3, ax4) = plt.subplots(4, 1, figsize=(10, 8))

data["Average Temperature"].plot(ax=ax1, title="Original Data", color='blue')

decomposition.trend.plot(ax=ax2, title="Trend", color='red')
decomposition.seasonal.plot(ax=ax3, title="Seasonality", color='green')
decomposition.resid.plot(ax=ax4, title="Residual (Noise)", color='purple')

plt.tight_layout()
plt.show()
```



```
In [5]: result = adfuller(data["Average Temperature"])
print(f"ADF Statistic: {result[0]}")
print(f"p-value: {result[1]}")
```

ADF Statistic: 1.1279797201576414
p-value: 0.9954484841402742

```
In [6]: data.reset_index(inplace=True)
data["Temp_diff"] = data["Average Temperature"].diff()

def plot_temperature(df: pd.DataFrame, y: str, title: str) -> None:
    fig = px.line(df, x='Year', y=y, labels={'Year': 'Year'})
    fig.update_layout(template="simple_white", font=dict(size=18),
                      title_text=title, width=650, title_x=0.5, height=400)
    return fig.show()

plot_temperature(df=data, y='Temp_diff', title='Differenced Temperature Data')
```

```
In [ ]: data["Temp_diff2"] = data["Temp_diff"].diff()
data.dropna(inplace=True)

result_diff2 = adfuller(data["Temp_diff2"])
print(f"ADF Statistic: {result_diff2[0]}")
print(f"p-value: {result_diff2[1]}")
```

ADF Statistic: -4.759518603359893
p-Value: 6.496008212398227e-05

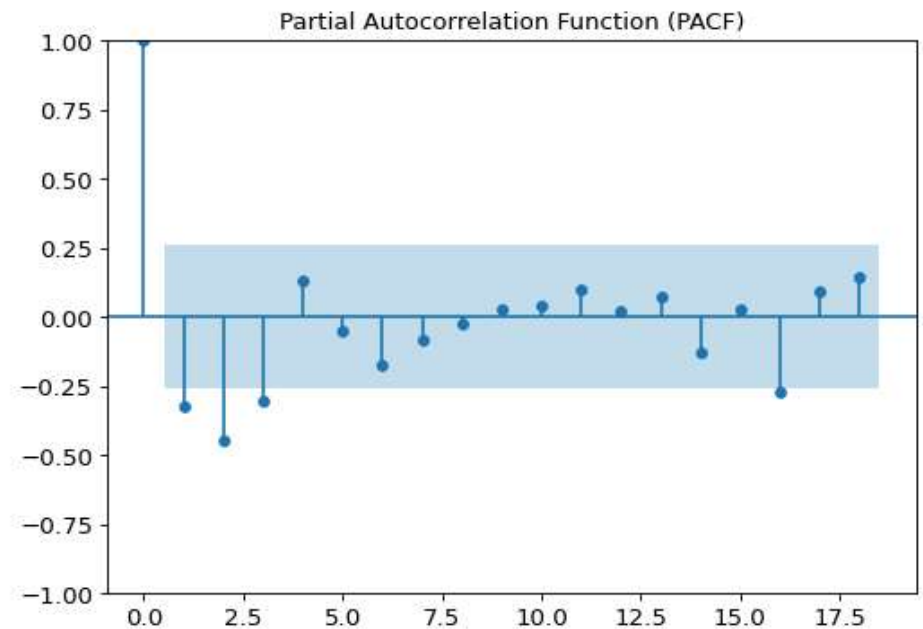
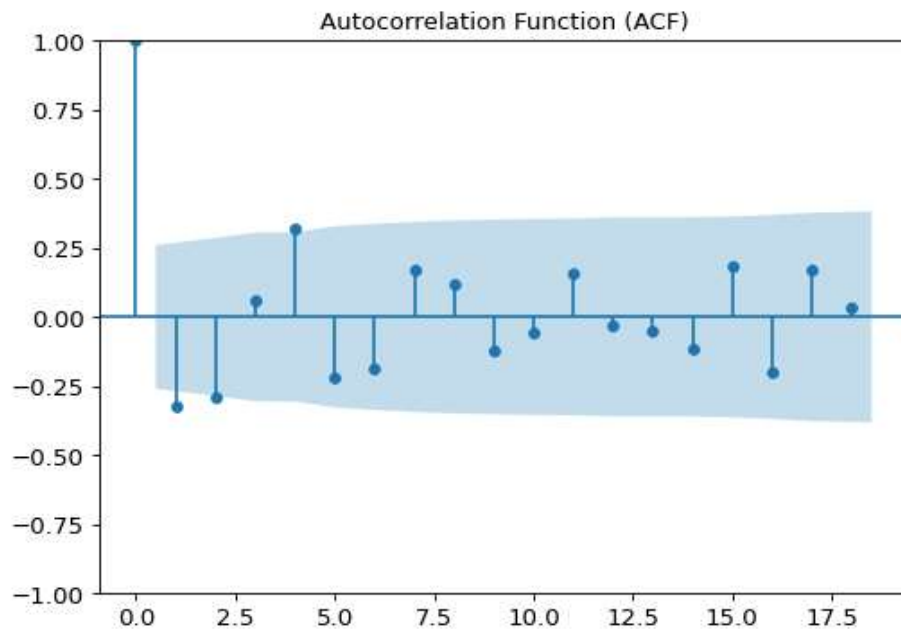
```
In [8]: fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(16, 5), dpi=80)

plot_acf(data['Temp_diff'].dropna(), ax=ax1)
plot_pacf(data['Temp_diff'].dropna(), ax=ax2, method='ywm')

ax1.set_title("Autocorrelation Function (ACF)")
ax2.set_title("Partial Autocorrelation Function (PACF)")

ax1.tick_params(axis='both', labelsize=12)
ax2.tick_params(axis='both', labelsize=12)

plt.show()
```



```
In [9]: data.set_index("Year", inplace=True)
data.index = data.index.astype(int)

sarima = SARIMAX(data['Average Temperature'],
                  order=(1,0,1),
                  seasonal_order=(1,0,1,10))
model_fit = sarima.fit()

predictions = model_fit.predict()
```

```
In [10]: future_years = 15

last_year = data.index.max()
forecast_index = list(range(last_year + 1, last_year + future_years + 1))
forecast_values = model_fit.forecast(steps=future_years)
```

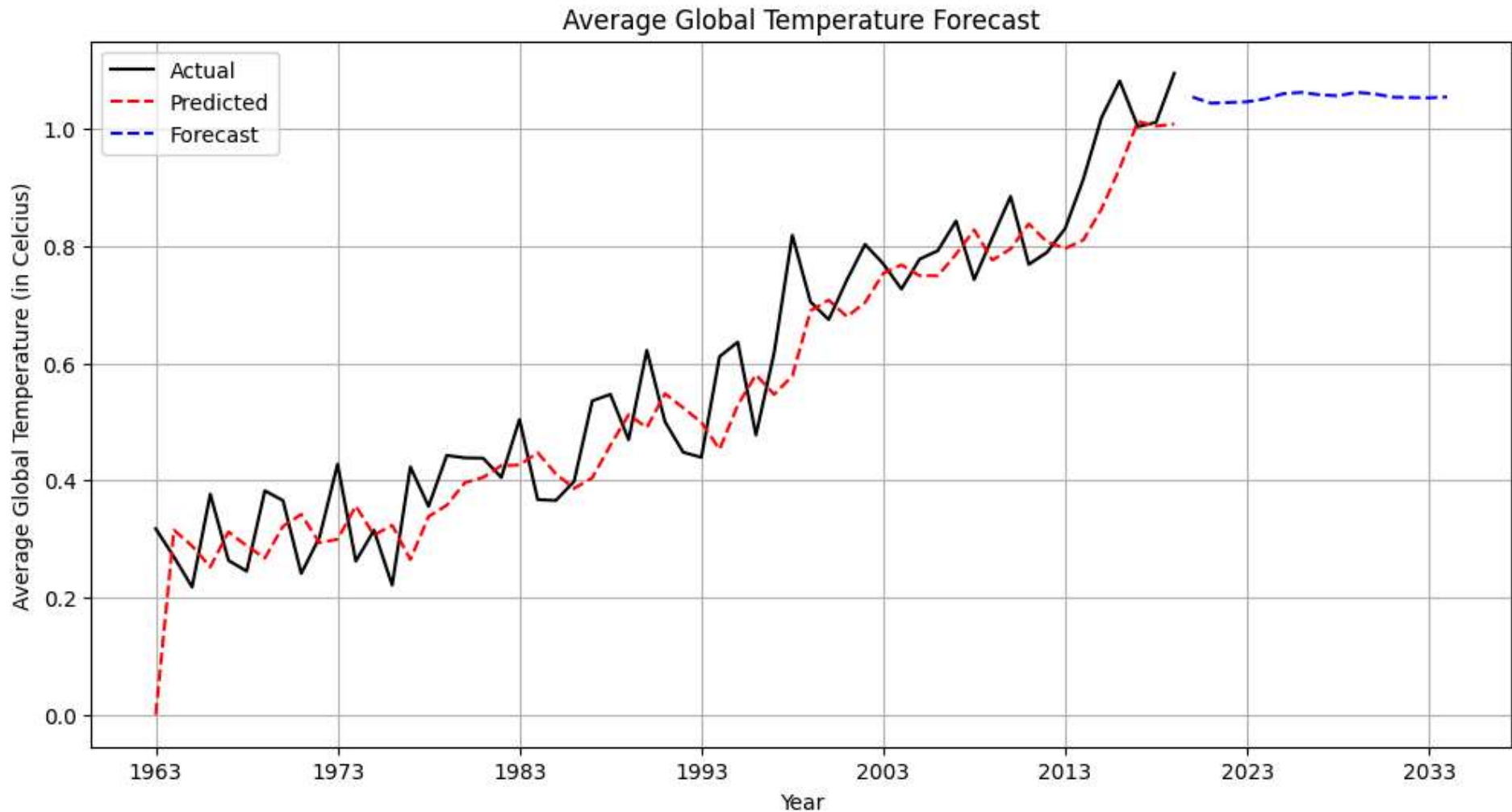
```
In [11]: plt.figure(figsize=(12,6))
plt.plot(data.index, data['Average Temperature'], label="Actual", color="black")
plt.plot(data.index, predictions, label="Predicted", color="red", linestyle="dashed")
plt.plot(forecast_index, forecast_values, label="Forecast", color="blue", linestyle="dashed")

plt.xticks(ticks=range(data.index.min(), last_year + future_years + 1, 10))

plt.xlabel("Year")
plt.ylabel("Average Global Temperature (in Celcius)")
```



```
plt.legend()
plt.title("Average Global Temperature Forecast")
plt.grid(True)
plt.show()
```



```
In [12]: actual = data["Average Temperature"]
predicted = model_fit.predict()

mae = mean_absolute_error(actual, predicted)
print(f"Mean Absolute Error (MAE): {mae:.4f}")
```

Mean Absolute Error (MAE): 0.0758