

```
In [1]: import pandas as pd
        from darts import TimeSeries
```

```
In [2]: df = pd.read_csv("MY3_May_2023_KNN_Imputed.csv")

df
```

```
Out[2]:
```

	Datetime	id	field1	field2	field3	field4	field5	field6	field7	field8
0	2022-07-03 20:30:00	2.200000e+01	8.871990	0.000000	398.496241	3.636364	26.00000	71.000000	45.666667	55.000000
1	2022-07-03 20:45:00	6.700000e+01	3.802281	0.000000	229.323308	0.000000	25.00000	62.000000	44.666667	53.333333
2	2022-07-03 21:00:00	1.120000e+02	12.674271	0.000000	184.210526	0.000000	25.00000	60.000000	43.666667	52.750000
3	2022-07-03 21:15:00	1.560000e+02	16.476553	0.000000	165.413534	0.000000	24.00000	59.666667	43.333333	52.666667
4	2022-07-03 21:30:00	5.307278e+05	321.540266	190.816359	794.429588	371.997558	27.25921	61.186420	18.749892	20.904609
...
28042	2023-04-21 23:00:00	1.061851e+06	397.944200	225.108225	700.960219	489.898990	29.00000	69.000000	27.666667	29.000000
28043	2023-04-21 23:15:00	1.061896e+06	392.070485	225.974026	707.818930	492.424242	29.00000	69.000000	31.666667	36.000000
28044	2023-04-21 23:30:00	1.061941e+06	393.538913	225.974026	716.049383	493.686869	29.00000	70.000000	33.333333	39.333333
28045	2023-04-21 23:45:00	1.061986e+06	392.070485	225.108225	727.023320	488.636364	29.00000	70.000000	35.333333	42.666667
28046	2023-04-22 00:00:00	1.062008e+06	414.096916	220.779221	720.164609	496.212121	29.00000	70.000000	37.000000	44.000000

28047 rows × 10 columns

```
In [3]: df1 = df
df1['Datetime'] = pd.to_datetime(df['Datetime'])

df1 = df1.set_index('Datetime')
#df1
df2=df1.resample('1H').mean()
df2.reset_index(inplace = True)
df2
```

```
Out[3]:
```

	Datetime	id	field1	field2	field3	field4	field5	field6	field7	field8
0	2022-07-03 20:00:00	4.450000e+01	6.337136	0.000000	313.909774	1.818182	25.500000	66.500000	45.166667	54.166667
1	2022-07-03 21:00:00	2.654309e+05	168.057839	95.408180	484.620809	185.998779	25.879605	60.509877	31.124946	36.806471
2	2022-07-03 22:00:00	5.307278e+05	321.540266	190.816359	794.429588	371.997558	27.259210	61.186420	18.749892	20.904609
3	2022-07-03 23:00:00	5.307278e+05	321.540266	190.816359	794.429588	371.997558	27.259210	61.186420	18.749892	20.904609
4	2022-07-04 00:00:00	5.307278e+05	321.540266	190.816359	794.429588	371.997558	27.259210	61.186420	18.749892	20.904609
...
7008	2023-04-21 20:00:00	1.061380e+06	395.374449	224.242424	685.528121	498.737374	30.000000	68.500000	22.833333	23.916667
7009	2023-04-21 21:00:00	1.061558e+06	396.475771	225.757576	692.386831	496.527778	29.750000	69.000000	25.500000	27.250000
7010	2023-04-21 22:00:00	1.061738e+06	393.538913	227.056277	710.562414	493.686869	29.000000	69.000000	31.000000	34.833333
7011	2023-04-21 23:00:00	1.061918e+06	393.906021	225.541126	712.962963	491.161616	29.000000	69.500000	32.000000	36.750000
7012	2023-04-22 00:00:00	1.062008e+06	414.096916	220.779221	720.164609	496.212121	29.000000	70.000000	37.000000	44.000000

7013 rows × 10 columns

```
In [4]: test1 = 24*7
train1 = len(df2) - test1
df3 = df2.set_index('Datetime')

df_train = df3[:train1]
df_test = df3[train1-24:train1+test1]
```

```
In [5]: targets = ['field1', 'field2', 'field3', 'field4', 'field7', 'field8']
        covariates = ['field5', 'field6']

        X_train_df = df_train[covariates]
        Y_train_df = df_train[targets]

        X_test_df = df_test[covariates]
        Y_test_df = df_test[targets]
```

```
In [6]: X_train = TimeSeries.from_dataframe(X_train_df)
        Y_train = TimeSeries.from_dataframe(Y_train_df)

        X_test = TimeSeries.from_dataframe(X_test_df)
        Y_test = TimeSeries.from_dataframe(Y_test_df)
```

```
In [7]: from darts.models import LightGBMModel
        import warnings
        warnings.filterwarnings("ignore")

        n = len(Y_test)

        # Initialize the LightGBM model
        modelgbm = LightGBMModel(lags=24, lags_past_covariates=24, output_chunk_length=24 )

        # Train the model on the training data
        modelgbm.fit(Y_train, past_covariates=X_train)
```

```
Out[7]: LightGBMModel(lags=24, lags_past_covariates=24, lags_future_covariates=None, output_chunk_length=24, add_encoders=None, likelihood=None, quantiles=None, random_state=None, multi_models=True, use_static_covariates=True, categorical_past_covariates=None, categorical_future_covariates=None, categorical_static_covariates=None)
```

```
In [8]: predgbm1 = modelgbm.predict(n, past_covariates = X_test)

        predictions = TimeSeries.pd_dataframe(predgbm1)
        predictions
```

Out[8]:

component	field1	field2	field3	field4	field7	field8
Datetime						
2023-04-15 01:00:00	236.654303	131.256030	971.778697	274.433483	44.642005	55.872391
2023-04-15 02:00:00	234.272876	131.489841	933.737946	275.134834	44.690564	53.076310
2023-04-15 03:00:00	234.330164	131.079942	875.621723	274.240319	40.307826	44.437782
2023-04-15 04:00:00	235.595233	129.105586	848.216065	271.321883	39.951555	49.712745
2023-04-15 05:00:00	233.550415	128.263549	863.698270	264.853221	41.495975	49.978437
...
2023-04-22 20:00:00	227.294735	128.249004	639.766720	249.824056	22.836582	24.673267
2023-04-22 21:00:00	227.569225	130.060764	699.607650	242.731075	17.601612	23.341107
2023-04-22 22:00:00	231.200047	132.068712	715.596295	256.451034	17.245505	21.593850
2023-04-22 23:00:00	219.432741	132.116203	693.136155	264.042528	18.386934	23.291541
2023-04-23 00:00:00	237.247010	131.696295	761.298249	261.004756	23.053891	23.752423

192 rows × 6 columns

```
In [20]: from darts.metrics import rmse, mae

# Convert actual and predicted values to TimeSeries
actual_series_list = [TimeSeries.from_dataframe(Y_test_df[[target]]) for target in targets]
predicted_series_list = [TimeSeries.from_dataframe(predictions[[target]]) for target in targets]

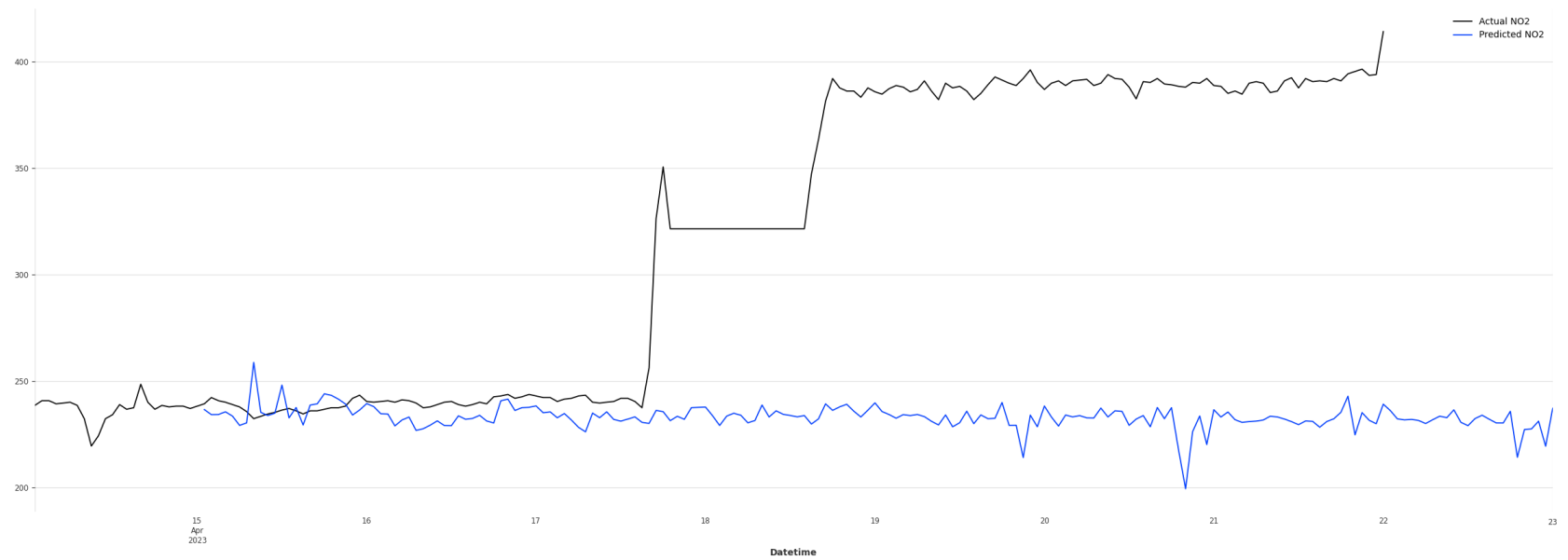
# Calculate RMSE and MAE for each target field
rmse_values = []
mae_values = []

for actual, predicted in zip(actual_series_list, predicted_series_list):
    rmse_value = rmse(actual, predicted)
    mae_value = mae(actual, predicted)
    rmse_values.append(rmse_value)
    mae_values.append(mae_value)

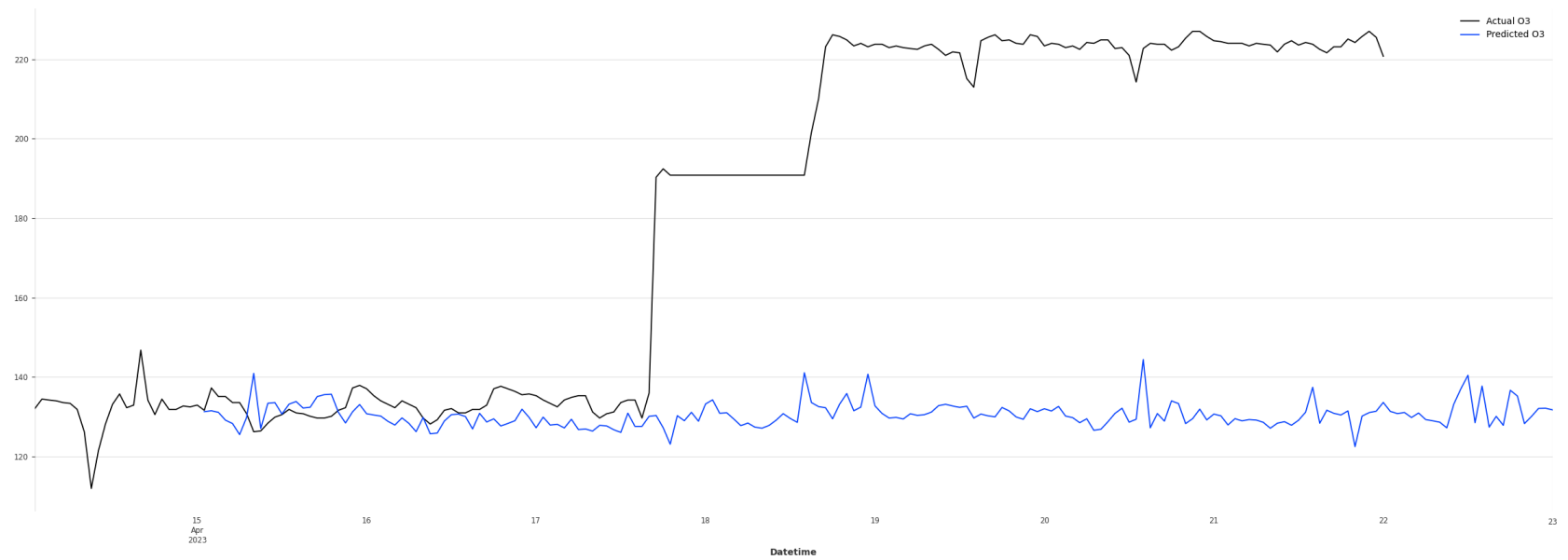
# Create a DataFrame to store the results for LightGBM
results_df = pd.DataFrame({
    'Field': targets,
    'RMSE_LightGBM': rmse_values,
    'MAE_LightGBM': mae_values
})

# Save the LightGBM results to a CSV file
results_df.to_csv('lightgbm_multi_results.csv', index=False)
```

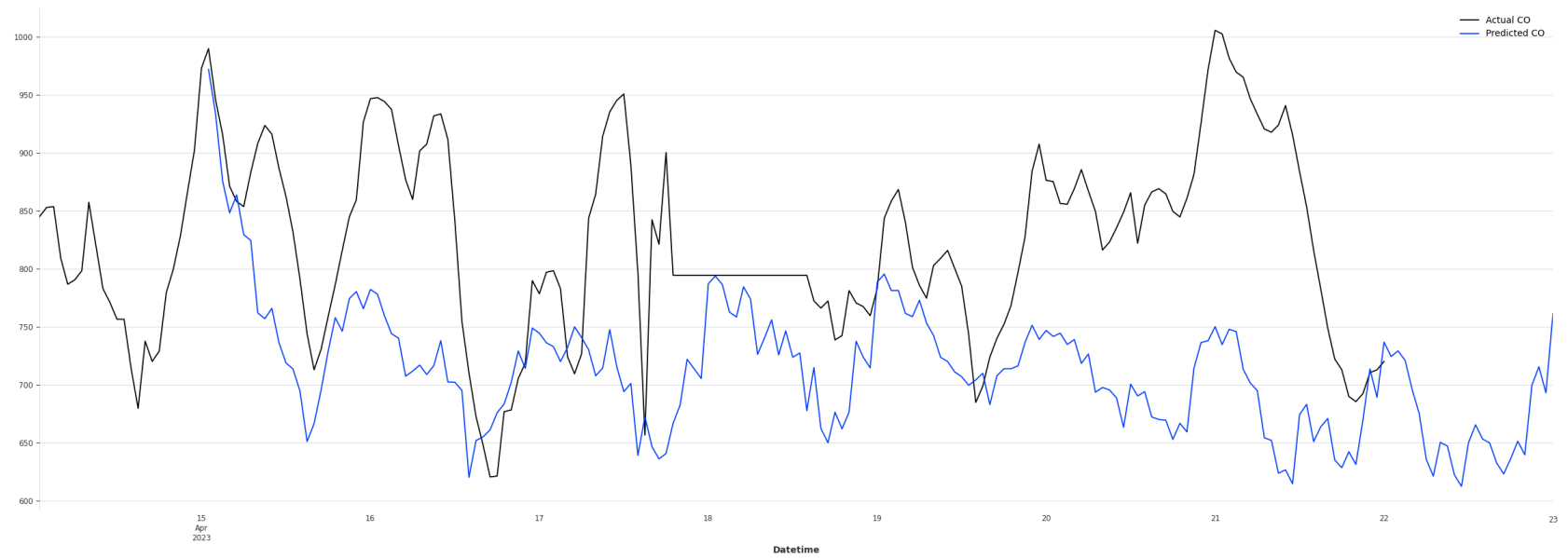
```
In [21]: import matplotlib.pyplot as plt
plt.figure(figsize=(30,10))
Y_test_df['field1'].plot(label='Actual NO2')
predictions['field1'].plot(label='Predicted NO2')
plt.legend()
plt.show()
```



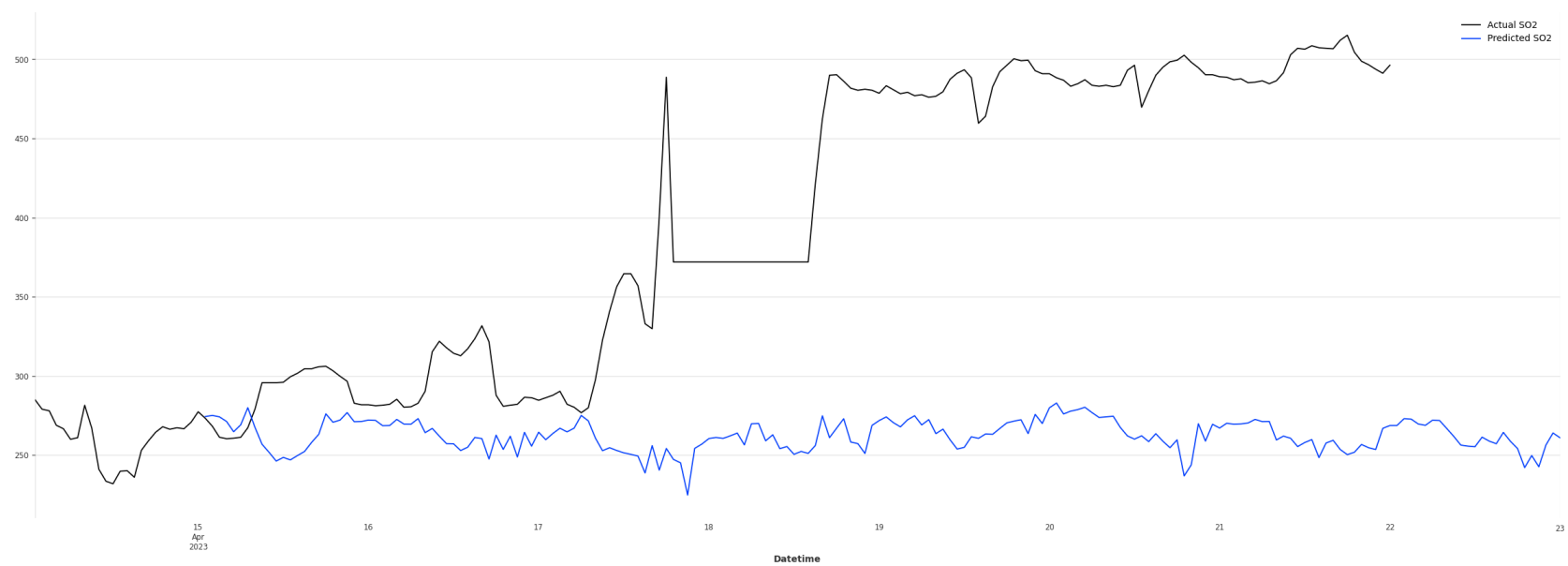
```
In [22]: plt.figure(figsize=(30,10))
Y_test_df['field2'].plot(label='Actual O3')
predictions['field2'].plot(label='Predicted O3')
plt.legend()
plt.show()
```



```
In [23]: plt.figure(figsize=(30,10))
Y_test_df['field3'].plot(label='Actual CO')
predictions['field3'].plot(label='Predicted CO')
plt.legend()
plt.show()
```



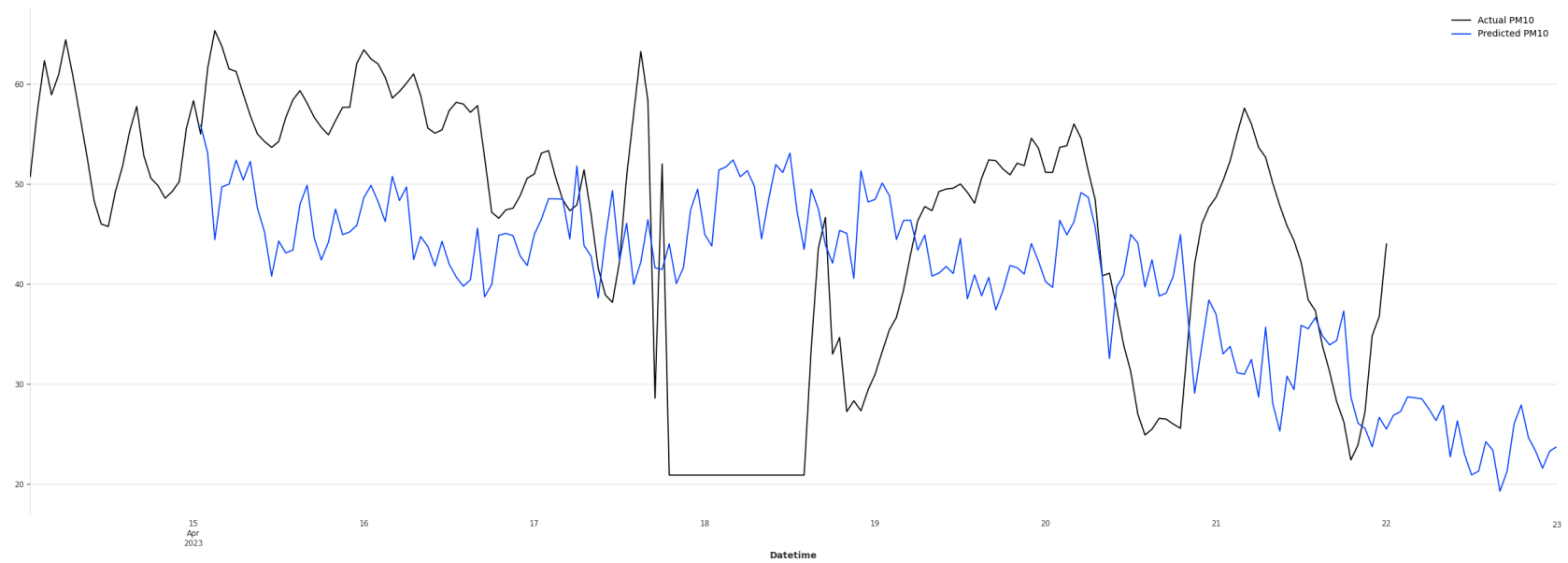

```
In [24]: plt.figure(figsize=(30,10))
Y_test_df['field4'].plot(label='Actual SO2')
predictions['field4'].plot(label='Predicted SO2')
plt.legend()
plt.show()
```



```
In [25]: plt.figure(figsize=(30,10))
Y_test_df['field7'].plot(label='Actual PM2.5')
predictions['field7'].plot(label='Predicted PM2.5')
plt.legend()
plt.show()
```



```
In [26]: plt.figure(figsize=(30,10))
Y_test_df['field8'].plot(label='Actual PM10')
predictions['field8'].plot(label='Predicted PM10')
plt.legend()
plt.show()
```



```
In [ ]:
```