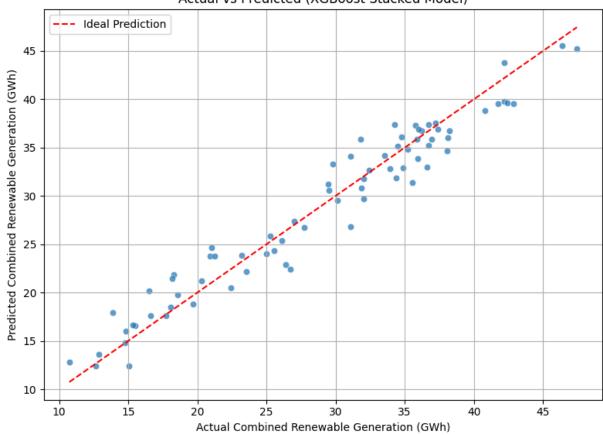
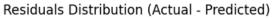
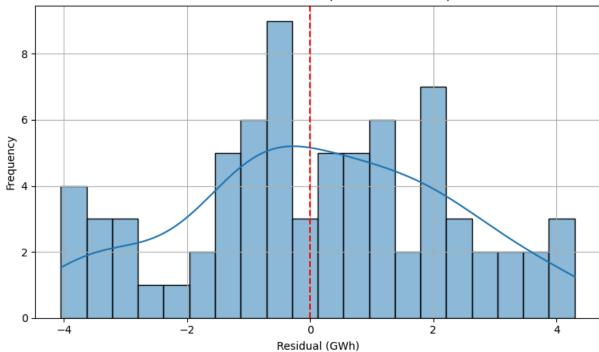
```
In [30]: import numpy as np
          import pandas as pd
          import xqboost as xqb
          from sklearn.model selection import train test split
          from sklearn.metrics import mean squared error, r2 score
          import matplotlib.pyplot as plt
          import seaborn as sns
          import shap
          from tgdm import tgdm
In [26]: # Read data
         df = pd.read excel("./datasets/datasets.xlsx", sheet name="Sheet1")
In [21]: # Features and Targets
          features = [
              'Tempmax_C', 'Tempmin_C', 'windspeedmax', 'windspeedmean',
'solarradiation', 'uvindex', 'cloudcover', 'humidity', 'precip'
          X = df[features]
          y wind = df['Wind GWh']
          y solar = df['(Combined) Solar GWh']
          y combined = df['(Combined) Renewable Generation GWh']
          # Split all data at once to ensure alignment
          X_train, X_test, y_train_wind, y_test_wind, y_train_solar, y_test_solar, y_t
              X, y wind, y solar, y combined, test size=0.2, random state=42
In [22]: # First-level models (Wind & Solar)
          model wind = xgb.XGBRegressor(
              eta=0.31,
              max depth=4,
              n_estimators=200,
              qamma=49,
              reg lambda=1.02,
              random state=42
          model solar = xgb.XGBRegressor(
              eta=0.31,
              max depth=4,
              n estimators=200,
              random state=42
          model wind.fit(X train, y train wind)
          model solar.fit(X train, y train solar)
          # First-level model predictions
          pred wind = model wind.predict(X test)
          pred solar = model solar.predict(X test)
In [23]: # Build stacking features (first-level outputs as inputs)
          stacked X = np.column stack([pred wind, pred solar])
```

```
# Second-level model (meta-regressor)
         meta model = xqb.XGBRegressor(
             eta=0.1352233009446337,
             max depth=2,
             n estimators=150,
             random state=42,
             colsample bytree=0.8532241907732696,
             qamma=10,
             reg lambda=0.8082370013955644,
             subsample=0.6976502088991097,
             tree method='hist',
             device='cuda'
         meta model.fit(stacked X, y test combined) # Note: using y test combined di
         # Predict combined output
         y pred combined = meta model.predict(stacked X)
In [32]: # Evaluation
         rmse stacked = np.sqrt(mean squared error(y test combined, y pred combined))
         r2 stacked = r2 score(y test combined, y pred combined)
         print(f"RMSE (XGB Stacked Combined): {rmse stacked:.3f}")
         print(f"R2 (XGB Stacked Combined): {r2 stacked:.3f}")
         # Scatter plot: True vs Predicted
         plt.figure(figsize=(8, 6))
         sns.scatterplot(x=y test combined, y=y pred combined, alpha=0.7)
         plt.plot([y test combined.min(), y test combined.max()],
                  [y_test_combined.min(), y_test_combined.max()],
                  color='red', linestyle='--', label='Ideal Prediction')
         plt.xlabel("Actual Combined Renewable Generation (GWh)")
         plt.ylabel("Predicted Combined Renewable Generation (GWh)")
         plt.title("Actual vs Predicted (XGBoost Stacked Model)")
         plt.legend()
         plt.grid(True)
         plt.tight_layout()
         plt.show()
         # Residual Plot
         residuals = y test combined - y pred combined
         plt.figure(figsize=(8, 5))
         sns.histplot(residuals, kde=True, bins=20)
         plt.axvline(0, color='red', linestyle='--')
         plt.title("Residuals Distribution (Actual - Predicted)")
         plt.xlabel("Residual (GWh)")
         plt.ylabel("Frequency")
         plt.grid(True)
         plt.tight layout()
         plt.show()
        RMSE (XGB Stacked Combined): 2.134
            (XGB Stacked Combined): 0.947
```



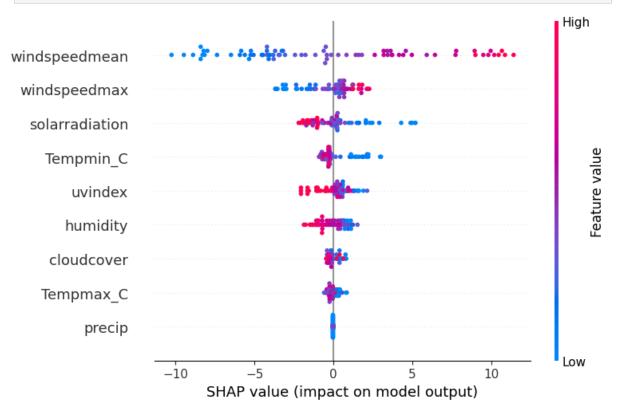






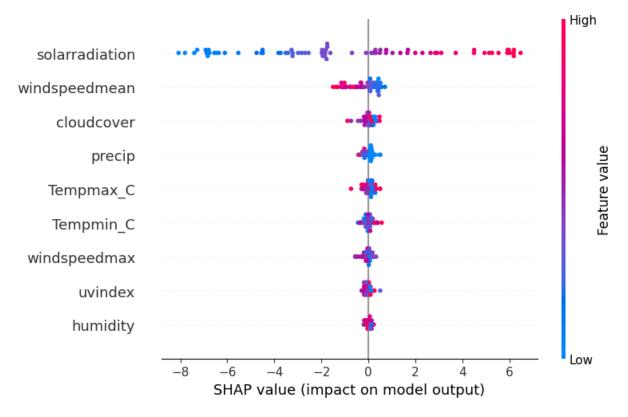
```
In [33]: # Shap Value for Wind power model
    explainer_wind = shap.Explainer(model_wind)
    shap_values_wind = explainer_wind(X_test)
```

```
# SHAP Summary Plot - Wind
shap.summary_plot(shap_values_wind, X_test, show=True)
```



```
In [34]: # Shap Value for Solar power model
explainer_solar = shap.Explainer(model_solar)
shap_values_solar = explainer_solar(X_test)

# SHAP Summary Plot - Solar
shap.summary_plot(shap_values_solar, X_test, show=True)
```



In []:

This notebook was converted with convert.ploomber.io