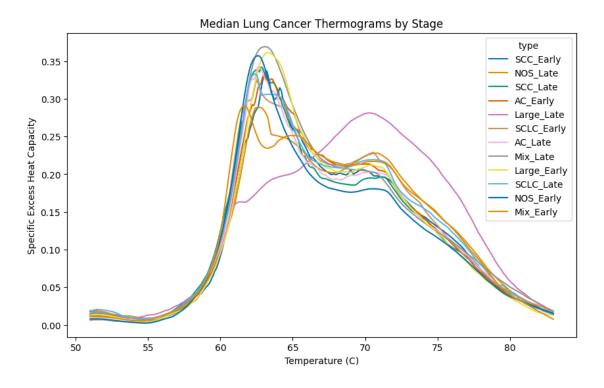
Stage_Results_and_Visualization

November 6, 2023

```
[]: import pandas as pd # standard
     import numpy as np # standard
     import matplotlib.pyplot as plt
     import seaborn as sns
     import thermogram_utilities
     import warnings
     warnings.filterwarnings("ignore")
[]: prev = pd.read_excel("/Users/avery/OneDrive/Documents/GitHub/
      →Clinical_TLB_2023-2024/lung_cancer_tlb.xlsx")
     update = pd.read_excel('/Users/avery/OneDrive/Documents/publication_meta_data.
      ⇔xlsx')
     merged = pd.merge(prev, update, left_on='pub_id', right_on="Patient")
     df = merged.drop(["CancerType", "sample_id", "pub_id", "Patient"], axis = 1)
[]: # get location of cut off values
     lower_column_index = df.columns.get_loc("T51")
     upper_column_index = df.columns.get_loc("T83.1")
     label_column_index = df.columns.get_loc("Diagnosis")
     cancer_column_index = df.columns.get_loc("Stage")
     column_indices = np.arange(lower_column_index, upper_column_index)
     column indices = np.append(column indices, cancer column index)
     column_indices = np.append(column_indices, label_column_index)
     df = df.iloc[:, column_indices]
[]: cancer type = "Control"
     prediction_df = df.loc[df["Diagnosis"] != cancer_type]
     #prediction df["Stage"] = np.where(prediction df["Stage"] == "Early", 0, 1)
     prediction_df["Cancer and Stage"] = prediction_df["Diagnosis"] + "_" +__"
      →prediction_df ["Stage"]
     prediction df = prediction df.drop(["Diagnosis", "Stage"], axis = 1)
```

```
[]: plt.figure(figsize=(10, 6)) # Adjust the figure size if needed sns.lineplot(data=median, x='temperature', y='median', hue='type',□ ⇔palette='colorblind') plt.xlabel('Temperature (C)') plt.ylabel('Specific Excess Heat Capacity') plt.title('Median Lung Cancer Thermograms by Stage')
```

[]: Text(0.5, 1.0, 'Median Lung Cancer Thermograms by Stage')

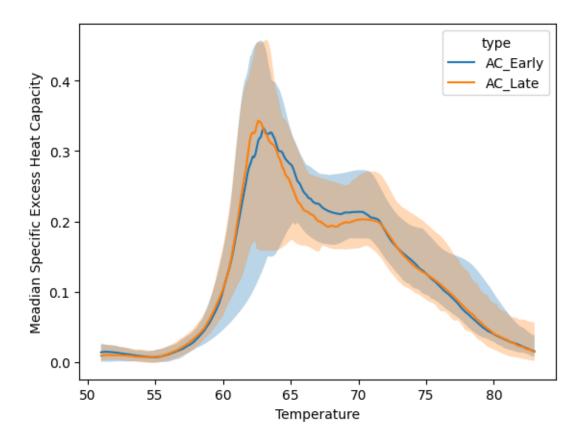


```
# Create a line plot using Seaborn with matching colors
sns.lineplot(data=graph_df, x='temperature', y='median', hue='type')

# Create separate ribbons for each "type" with matching colors
for type_name in graph_df['type'].unique():
    type_data = graph_df[graph_df['type'] == type_name]
    plt.fill_between(type_data["temperature"], type_data["lower_q"],
    type_data["upper_q"], alpha=0.3, label=type_name)

plt.xlabel("Temperature")
plt.ylabel("Meadian Specific Excess Heat Capacity")
```

[]: Text(0, 0.5, 'Meadian Specific Excess Heat Capacity')



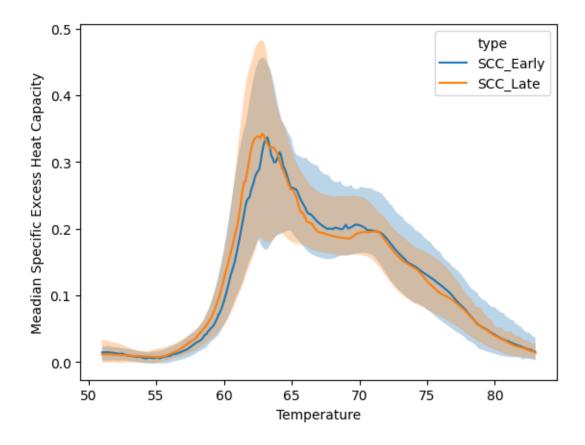
```
adeno_stage['max_features'] = np.where(pd.isna(adeno_stage['max_features']),__

\(\text{\text{"None"}}, \text{ adeno_stage["max_features"]}\)

adeno_stage_results = adeno_stage.groupby(['n_estimators', 'max_depth',__

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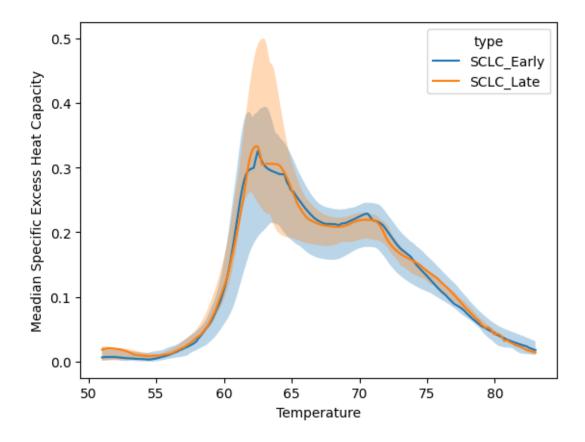
[]: Text(0, 0.5, 'Meadian Specific Excess Heat Capacity')



```
[]:
                                                                        AUC
         n_estimators max_depth max_features
                                              Weighted Accuracy
     19
                 1000
                           23.0
                                         log2
                                                        0.508604
                                                                  0.505465
     22
                 1000
                           None
                                         log2
                                                        0.507866
                                                                  0.505197
     4
                  100
                           None
                                         log2
                                                        0.507601 0.505052
                  100
                           23.0
     1
                                         log2
                                                        0.507088 0.503352
     16
                  500
                           None
                                         log2
                                                        0.506661 0.506488
     7
                  250
                           23.0
                                         log2
                                                        0.505944 0.504251
     10
                  250
                                                        0.505291 0.504999
                           None
                                         log2
     13
                  500
                           23.0
                                         log2
                                                        0.505136 0.506358
```

```
14
             500
                      23.0
                                    sqrt
                                                   0.501645 0.501663
20
            1000
                      23.0
                                                   0.501031 0.500505
                                    sqrt
5
             100
                      None
                                    sqrt
                                                   0.500841 0.501724
17
             500
                                                   0.500285 0.500675
                      None
                                    sqrt
23
            1000
                      None
                                                   0.500180 0.502527
                                    sqrt
             250
11
                      None
                                    sqrt
                                                   0.499799 0.499417
8
             250
                      23.0
                                    sqrt
                                                   0.499683 0.500981
2
             100
                      23.0
                                    sqrt
                                                   0.498910 0.502373
9
             250
                                    None
                                                   0.491736 0.490107
                      None
21
            1000
                      None
                                    None
                                                   0.491094 0.488660
                                                   0.490367 0.485551
0
             100
                      23.0
                                    None
15
             500
                      None
                                    None
                                                   0.490193 0.488511
6
             250
                      23.0
                                    None
                                                   0.488881 0.488590
18
            1000
                      23.0
                                    None
                                                   0.488835 0.487184
12
             500
                                                   0.488658 0.487599
                      23.0
                                    None
3
             100
                      None
                                    None
                                                   0.487377 0.488877
```

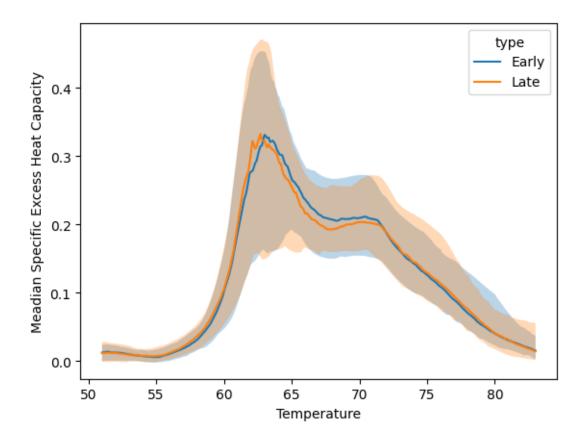
[]: Text(0, 0.5, 'Meadian Specific Excess Heat Capacity')



```
[]:
         n_estimators max_depth max_features Weighted Accuracy
                                                                        AUC
     19
                 1000
                           23.0
                                         log2
                                                        0.508604 0.505465
     22
                 1000
                           None
                                         log2
                                                        0.507866 0.505197
     4
                  100
                           None
                                         log2
                                                        0.507601 0.505052
                           23.0
                  100
                                         log2
                                                        0.507088 0.503352
     16
                  500
                           None
                                         log2
                                                        0.506661 0.506488
     7
                  250
                           23.0
                                         log2
                                                        0.505944 0.504251
     10
                  250
                                                        0.505291 0.504999
                           None
                                         log2
     13
                  500
                           23.0
                                         log2
                                                        0.505136 0.506358
```

```
14
                 500
                          23.0
                                       sqrt
                                                      0.501645 0.501663
    20
                1000
                          23.0
                                                      0.501031 0.500505
                                       sqrt
    5
                 100
                          None
                                       sqrt
                                                      0.500841 0.501724
    17
                                                      0.500285 0.500675
                 500
                          None
                                       sqrt
    23
                1000
                          None
                                                      0.500180 0.502527
                                       sqrt
    11
                 250
                          None
                                       sqrt
                                                      0.499799 0.499417
    8
                 250
                          23.0
                                                      0.499683 0.500981
                                       sqrt
    2
                 100
                          23.0
                                       sqrt
                                                      0.498910 0.502373
    9
                 250
                                                      0.491736 0.490107
                          None
                                       None
    21
                 1000
                          None
                                       None
                                                      0.491094 0.488660
                                                      0.490367 0.485551
                          23.0
    0
                 100
                                       None
    15
                 500
                          None
                                       None
                                                      0.490193 0.488511
    6
                 250
                          23.0
                                       None
                                                      0.488881 0.488590
    18
                1000
                          23.0
                                       None
                                                      0.488835 0.487184
    12
                 500
                          23.0
                                       None
                                                      0.488658 0.487599
    3
                 100
                          None
                                       None
                                                      0.487377 0.488877
[]: cancer type = "Control"
    prediction_df = df.loc[df["Diagnosis"] != cancer_type]
     #prediction_df["Stage"] = np.where(prediction_df["Stage"] == "Early", 0, 1)
     \#prediction\_df["Cancer and Stage"] = prediction\_df["Diagnosis"] + "_ " +_ \pm \]
      →prediction_df["Stage"]
    prediction_df = prediction_df.drop(["Diagnosis"], axis = 1)
    df_long = pd.melt(prediction_df, id_vars=['Stage'], var_name='temp',__
      →value name='dsp' )
    median = thermogram utilities.median_curve(df_long, 'Stage', 'temp', 'dsp')
    median['temperature'] = median['temperature'].str.replace('T', '').astype(float)
[]: graph_df = median
     # Create a line plot using Seaborn with matching colors
    sns.lineplot(data=graph_df, x='temperature', y='median', hue='type')
     # Create separate ribbons for each "type" with matching colors
    for type name in graph df['type'].unique():
        type_data = graph_df[graph_df['type'] == type_name]
        plt.fill_between(type_data["temperature"], type_data["lower_q"],__
      plt.xlabel("Temperature")
    plt.ylabel("Meadian Specific Excess Heat Capacity")
```

[]: Text(0, 0.5, 'Meadian Specific Excess Heat Capacity')



```
[]:
         n_estimators max_depth max_features
                                               Weighted Accuracy
                                                                        AUC
                  250
                            74.0
                                                        0.544999
                                                                   0.561948
     8
                                         sqrt
                  500
     15
                           None
                                         None
                                                        0.544723 0.560832
     23
                 1000
                           None
                                         sqrt
                                                        0.544414 0.562680
     20
                 1000
                           74.0
                                         sqrt
                                                        0.544325 0.563296
     16
                  500
                           None
                                         log2
                                                        0.543799 0.562731
     12
                  500
                           74.0
                                         None
                                                        0.543614 0.559947
     6
                  250
                           74.0
                                         None
                                                        0.543485
                                                                   0.559607
     18
                 1000
                           74.0
                                         None
                                                        0.543387 0.560408
```

```
14
             500
                      74.0
                                    sqrt
                                                   0.543377 0.562651
21
            1000
                      None
                                    None
                                                   0.543317 0.560060
3
             100
                      None
                                    None
                                                   0.543149
                                                             0.558518
17
             500
                      None
                                                   0.543094 0.562576
                                    sqrt
19
            1000
                      74.0
                                    log2
                                                   0.542974 0.563247
22
            1000
                      None
                                                   0.542952 0.562719
                                    log2
             250
                                                   0.542803 0.561582
10
                      None
                                    log2
9
             250
                      None
                                   None
                                                   0.542734 0.559709
0
             100
                      74.0
                                                   0.542715 0.560208
                                   None
7
             250
                      74.0
                                    log2
                                                   0.542478 0.562791
13
             500
                      74.0
                                    log2
                                                   0.542420 0.562156
1
             100
                      74.0
                                    log2
                                                   0.542262 0.560518
4
             100
                      None
                                    log2
                                                   0.541701 0.560100
2
             100
                      74.0
                                                   0.541341 0.560550
                                    sqrt
5
             100
                      None
                                                   0.541248
                                                             0.560732
                                    sqrt
11
             250
                      None
                                    sqrt
                                                   0.541178 0.561758
```

```
[]: results = pd.concat([adeno_stage_results.head(1), scc_stage_results.head(1), usclc_stage_results.head(1), all_stage_results.head(1)], ignore_index=True)

#results = results.drop("Type", axis=1)

results.insert(0, "Type", ["AC", "SCC", "SCLC", "All"])

results
```

```
[]:
        Type n_estimators max_depth max_features Weighted Accuracy
                                                                            AUC
     0
          AC
                      1000
                                None
                                             sqrt
                                                             0.550638
                                                                       0.584997
     1
         SCC
                      1000
                                23.0
                                             log2
                                                             0.508604
                                                                       0.505465
     2 SCLC
                      1000
                                23.0
                                             log2
                                                             0.508604
                                                                       0.505465
                       250
                                74.0
     3
         All
                                             sqrt
                                                             0.544999
                                                                       0.561948
```