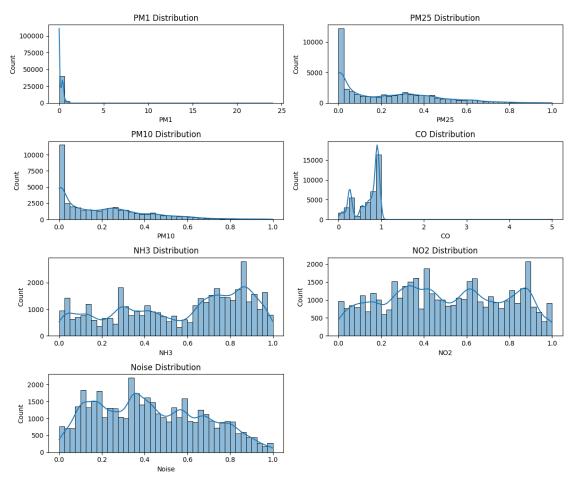
eda

June 2, 2025

```
[55]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     import plotly.express as px
     import plotly.graph_objects as go
     %matplotlib inline
[56]: data_path = r"data\DigitalExposome Dataset.csv"
     df = pd.read_csv(data_path)
[57]: df.columns
[57]: Index(['IBI', 'HR', 'NO2', 'Noise', 'NH3', 'PM10', 'CO', 'PM25', 'Label',
            'PM1', 'EDA', 'BVP'],
           dtype='object')
[58]: df.head()
[58]:
        IBI
                   HR NO2
                                                              CO
                                                                      PM25 \
                              Noise
                                          NH3
                                                  PM10
     0 0.0 0.377574
                                     0.003018 0.003091
                                                        0.871758 0.000000
                      0.0 0.511358
     1 0.0 0.196398
                      0.0 0.490903
                                     0.003018 0.003091
                                                        0.876848
                                                                  0.003091
     2 0.0 0.454163
                      0.0 0.470449
                                     0.006036 0.006181 0.881939 0.006181
     3 0.0 0.322451
                      0.0 0.449995
                                     0.009055 0.009272
                                                        0.887030 0.009272
     4 0.0 0.237595 0.0 0.429540 0.012073 0.012362 0.892121 0.012362
                    PM1 EDA BVP
        Label
            5 0.000000 0.0 0.0
     0
     1
            5 0.001854 0.0 0.0
     2
            5 0.003709 0.0 0.0
     3
            5 0.005563 0.0 0.0
            5 0.007417 0.0 0.0
[59]: df.shape
[59]: (42436, 12)
```

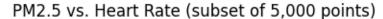
```
[60]: df.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 42436 entries, 0 to 42435
     Data columns (total 12 columns):
          Column Non-Null Count Dtvpe
      0
          IBI
                  42436 non-null float64
                  42436 non-null float64
      1
          HR.
      2
          NO2
                  42436 non-null float64
      3
          Noise
                  42436 non-null float64
      4
          NH3
                  42436 non-null float64
      5
          PM10
                  42436 non-null float64
      6
                  42436 non-null float64
      7
          PM25
                  42436 non-null float64
                  42436 non-null int64
      8
          Label
      9
          PM1
                  42436 non-null float64
      10 EDA
                  42436 non-null float64
      11 BVP
                  42436 non-null float64
     dtypes: float64(11), int64(1)
     memory usage: 3.9 MB
[61]: df.describe().T
[61]:
              count
                         mean
                                    std
                                         min
                                                   25%
                                                             50%
                                                                       75%
                                                                             max
                     0.178688 0.230127
      IBI
             42436.0
                                              0.000000 0.016760
                                                                  0.309984
                                                                             1.0
                                         0.0
     HR
            42436.0
                    0.530687 0.266531
                                         0.0
                                              0.309505
                                                        0.491213
                                                                  0.776388
                                                                             1.0
     NO2
            42436.0
                     0.502239 0.274914
                                         0.0
                                              0.285412
                                                        0.489703
                                                                  0.736983
                                                                             1.0
      Noise 42436.0
                    0.427541 0.247601
                                         0.0
                                              0.214008 0.403671
                                                                  0.618592
                                                                             1.0
     NH3
             42436.0 0.564276 0.294515
                                         0.0
                                              0.305285
                                                        0.644756
                                                                  0.826943
                                                                             1.0
     PM10
             42436.0 0.208474 0.210981
                                        0.0
                                              0.017624 0.158454
                                                                  0.326889
                                                                             1.0
      CO
             42436.0 0.679874 0.308111 0.0
                                              0.458047
                                                        0.805352
                                                                  0.909832
                                                                             5.0
     PM25
             42436.0 0.224901 0.223242 0.0 0.012362 0.176150
                                                                  0.376813
                                                                             1.0
                                                                  5.000000
     Label 42436.0 3.326185 1.599863
                                         1.0
                                              2.000000 4.000000
                                                                             5.0
     PM1
             42436.0 0.230773 0.255253
                                         0.0
                                              0.006310 0.176316
                                                                  0.407268
                                                                            24.0
      EDA
             42436.0 0.260318 0.221725
                                         0.0
                                              0.060686 0.216064
                                                                  0.421109
                                                                             1.0
     BVP
            42436.0 0.482902 0.175569
                                         0.0 0.449963 0.512280
                                                                  0.579756
                                                                             1.0
[62]: missing = df.isna().sum()
      missing[missing > 0].sort_values(ascending=False)
[62]: Series([], dtype: int64)
 []: env cols = ["PM1", "PM25", "PM10", "CO", "NH3", "NO2", "Noise"]
      plt.figure(figsize=(10, 8))
      for i, col in enumerate(env_cols, 1):
         plt.subplot(4, 2, i)
```

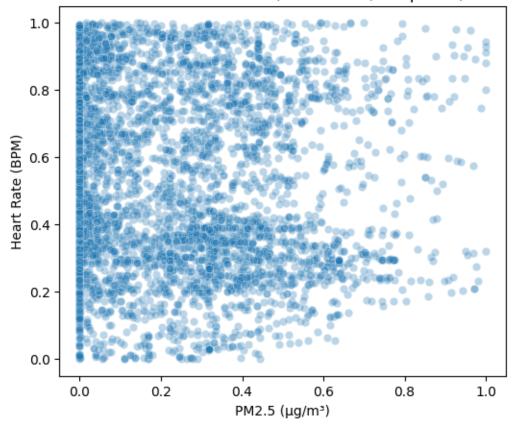
```
sns.histplot(df[col], kde=True, bins=40)
  plt.title(f"{col} Distribution")
plt.tight_layout()
plt.show()
```



```
fig.show()
[66]: def noise_category(x):
          if x < 0.3:
              return "Low"
          elif x < 0.6:
              return "Medium"
          else:
              return "High"
      df["Noise_cat"] = df["Noise"].apply(noise_category)
      fig = px.box(
          df,
          x="Noise_cat",
          y="EDA",
          title="EDA by Noise Category",
          width=800,
          height=600
      fig.show()
[67]: corr =
      ⇔df[["PM1","PM25","PM10","CO","NH3","NO2","Noise","HR","EDA","BVP","IBI"]].
      ⇔corr()
      fig = px.imshow(
          corr,
          text_auto=".2f",
          title="Correlation Matrix (All Env + Physio Variables)",
          width=800,
          height=600
      fig.show()
[68]: spikes = df[df["PM1"] > 5.0]
      fig = go.Figure()
      fig.add_trace(go.Scatter(x=spikes.index, y=spikes["PM1"], mode="markers",
       →name="PM1"))
      fig.add_trace(go.Scatter(x=spikes.index, y=spikes["HR"], mode="lines", name="HR__
      ⇔(closeup)"))
      fig.update_layout(
          title="PM1 Spikes & HR Overlaps",
          xaxis_title="Time",
          yaxis_title="Values",
          width=800,
          height=600
```

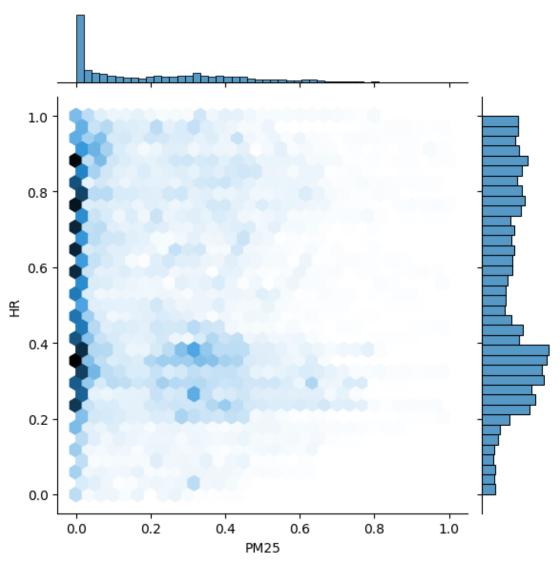
```
fig.show()
[69]: df["PM25_bin"] = pd.qcut(df["PM25"], q=4, labels=False)
      fig = px.box(
          df,
          x="PM25_bin",
          y="EDA",
          title="EDA by PM2.5 Quartile",
          width=800,
          height=600
      fig.show()
[70]: plt.figure(figsize=(6, 5))
      sns.scatterplot(x="PM25", y="HR", data=df.sample(5000), alpha=0.3)
      plt.title("PM2.5 vs. Heart Rate (subset of 5,000 points)")
      plt.xlabel("PM2.5 (µg/m³)")
      plt.ylabel("Heart Rate (BPM)")
      plt.show()
```



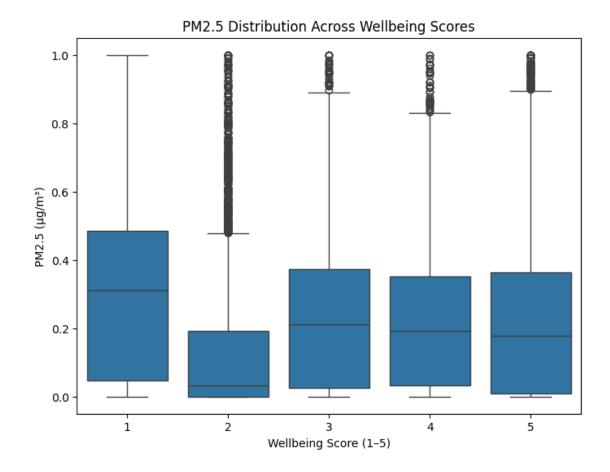


```
[71]: sns.jointplot(x="PM25", y="HR", data=df, kind="hex", gridsize=30)
plt.suptitle("Hexbin: PM2.5 vs. HR", y=1.02)
plt.show()
```

Hexbin: PM2.5 vs. HR



```
[72]: plt.figure(figsize=(8, 6))
sns.boxplot(x="Label", y="PM25", data=df)
plt.title("PM2.5 Distribution Across Wellbeing Scores")
plt.xlabel("Wellbeing Score (1-5)")
plt.ylabel("PM2.5 (µg/m³)")
plt.show()
```



[]: