Problem 2

September 15, 2025

[8]: pip install pandas scipy seaborn Requirement already satisfied: pandas in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (2.1.4) Requirement already satisfied: scipy in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (1.11.4) Collecting seaborn Using cached seaborn-0.13.2-py3-none-any.whl.metadata (5.4 kB) Requirement already satisfied: numpy<2,>=1.23.2 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from pandas) (1.26.4) Requirement already satisfied: python-dateutil>=2.8.2 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from pandas) (2.8.2) Requirement already satisfied: pytz>=2020.1 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from pandas) (2023.3.post1) Requirement already satisfied: tzdata>=2022.1 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from pandas) (2023.3) Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from seaborn) (3.8.3) Requirement already satisfied: contourpy>=1.0.1 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.2.0) Requirement already satisfied: cycler>=0.10 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (0.12.1) Requirement already satisfied: fonttools>=4.22.0 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (4.50.0) Requirement already satisfied: kiwisolver>=1.3.1 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.4.5) Requirement already satisfied: packaging>=20.0 in /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (24.2)

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Requirement already satisfied: pillow>=8 in
     /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from
     matplotlib!=3.6.1,>=3.4->seaborn) (10.2.0)
     Requirement already satisfied: pyparsing>=2.3.1 in
     /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from
     matplotlib!=3.6.1,>=3.4->seaborn) (3.1.2)
     Requirement already satisfied: six>=1.5 in
     /Users/varagantibasanthkumar/miniconda3/lib/python3.11/site-packages (from
     python-dateutil>=2.8.2->pandas) (1.16.0)
     Using cached seaborn-0.13.2-py3-none-any.whl (294 kB)
     Installing collected packages: seaborn
     Successfully installed seaborn-0.13.2
     Note: you may need to restart the kernel to use updated packages.
 [9]: import numpy as np
      import matplotlib.pyplot as plt
      import pandas as pd
      from scipy import stats
      import warnings
      warnings.filterwarnings('ignore')
[10]: def load sentinel2 data(file path='/Users/varagantibasanthkumar/Desktop/Remote,
       ⇔sensing - IMGS 589/Homework1/sentinel2_rochester.npy'):
          data = np.load(file path)
          print("Data loaded successfully!")
          print("Shape: " + str(data.shape))
          return data
[11]: def get_band_info():
          bands = \{
              0: 'B1', 1: 'B2', 2: 'B3', 3: 'B4', 4: 'B5', 5: 'B6',
              6: 'B7', 7: 'B8', 8: 'B8A', 9: 'B9', 10: 'B11', 11: 'B12'
          }
          return bands
[12]: def find_no_data(data, threshold=0.001):
          mask = (data < threshold) | np.isnan(data)</pre>
          return mask
[13]: def calculate_stats(data):
          print("Calculating band statistics...")
          no_data_mask = find_no_data(data)
          band_names = get_band_info()
          results = {}
          for i in range(12):
```

```
band_name = band_names[i]
    band_data = data[:, :, i]
    # get valid pixels only
    valid_pixels = band_data[~no_data_mask[:, :, i]]
    if len(valid_pixels) == 0:
        continue
    # basic stats
    mean_val = np.mean(valid_pixels)
    std_val = np.std(valid_pixels)
    min_val = np.min(valid_pixels)
    max_val = np.max(valid_pixels)
    # quartiles
    q1 = np.percentile(valid_pixels, 25)
    median = np.median(valid_pixels)
    q3 = np.percentile(valid_pixels, 75)
    # skewness and kurtosis
    skew = stats.skew(valid_pixels)
    kurt = stats.kurtosis(valid_pixels)
    results[band_name] = {
        'mean': mean val,
        'std': std_val,
        'min': min_val,
        'max': max_val,
        'q1': q1,
        'median': median,
        'q3': q3,
        'skewness': skew,
        'kurtosis': kurt,
        'count': len(valid_pixels)
    }
    print(band_name + " done")
return results
```

```
[14]: def explain_stats():
    print("\nWhat these statistics mean:")
    print("Mean: average value")
    print("Std: how spread out the data is")
    print("Min/Max: smallest and largest values")
    print("Q1, Median, Q3: quartiles (25%, 50%, 75%)")
```

```
print("Skewness: if data is symmetric or not")
print("Kurtosis: how \"heavy\" the tails are")
```

```
[15]: def make_table(stats_dict):
          print("\nBand Statistics Table:")
          # make a simple table
          table_data = []
          for band, values in stats_dict.items():
              row = {
                  'Band': band,
                  'Mean': round(values['mean'], 4),
                  'Std': round(values['std'], 4),
                  'Min': round(values['min'], 4),
                  'Max': round(values['max'], 4),
                  'Q1': round(values['q1'], 4),
                  'Median': round(values['median'], 4),
                  'Q3': round(values['q3'], 4),
                  'Skewness': round(values['skewness'], 4),
                  'Kurtosis': round(values['kurtosis'], 4)
              }
              table_data.append(row)
          df = pd.DataFrame(table_data)
          print(df)
          return df
```

```
def standardize_data(data):
    print("\nStandardizing data...")

    no_data_mask = find_no_data(data)
    standardized = np.full_like(data, np.nan)

for i in range(12):
    band_data = data[:, :, i]
    mask = no_data_mask[:, :, i]

# get valid data
    valid_data = band_data[~mask]

if len(valid_data) == 0:
    continue

# calculate mean and std
    mean_val = np.mean(valid_data)
    std_val = np.std(valid_data)
```

```
# z-score formula
              z_scores = (band_data - mean_val) / std_val
              # keep no-data as NaN
              z_scores[mask] = np.nan
              standardized[:, :, i] = z_scores
              print("Band " + str(i+1) + " standardized")
          return standardized
[17]: def explain_standardization():
          print("\nWhat standardization does:")
          print("- Makes mean = 0 and std = 1 for each band")
          print("- Allows comparison between different bands")
          print("- Helps find outliers (values > 3 or < -3)")</pre>
          print("- Common preprocessing step")
[18]: def plot_histograms(original_data, standardized_data):
          print("\nMaking histograms...")
          band_names = get_band_info()
          no_data_mask = find_no_data(original_data)
          fig, axes = plt.subplots(4, 3, figsize=(15, 20))
          fig.suptitle('Band Histograms with Outliers', fontsize=16)
          axes = axes.flatten()
          for i in range(12):
              ax = axes[i]
              band_name = band_names[i]
              # get data
              orig_band = original_data[:, :, i]
              std_band = standardized_data[:, :, i]
              mask = no_data_mask[:, :, i]
              # valid pixels only
              valid_orig = orig_band[~mask]
              valid_std = std_band[~mask]
              if len(valid_orig) == 0:
                  continue
              # plot histogram
```

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ax.hist(valid_orig, bins=50, alpha=0.7, color='blue')
              # find outliers (> 3 std devs)
              outliers = np.abs(valid_std) > 3
              outlier_values = valid_orig[outliers]
              # plot outliers as red dots
              if len(outlier_values) > 0:
                  y_pos = np.random.uniform(0, ax.get_ylim()[1]*0.8,__
       →len(outlier_values))
                  ax.scatter(outlier_values, y_pos, color='red', s=10, alpha=0.7)
              # add title
              mean_val = np.mean(valid_orig)
              std_val = np.std(valid_orig)
              outlier_pct = (len(outlier_values) / len(valid_orig)) * 100
              title = band_name + "\nMean: " + str(round(mean_val, 3)) + ", Std: " + L

str(round(std_val, 3)) + "\nOutliers: " + str(round(outlier_pct, 1)) + "%"

              ax.set_title(title, fontsize=9)
              ax.set_xlabel('Reflectance')
              ax.set_ylabel('Count')
          plt.tight_layout()
          plt.savefig('histograms.png', dpi=150, bbox_inches='tight')
          plt.show()
          print("Histograms saved as histograms.png")
[19]: def analyze_outliers(standardized_data):
          print("\nOutlier Analysis:")
          band_names = get_band_info()
          no_data_mask = find_no_data(standardized_data)
          print("Band\tTotal\t>2 \t>3 \t%>2 \t%>3 ")
          print("-" * 40)
          for i in range(12):
              band_name = band_names[i]
              std_band = standardized_data[:, :, i]
              mask = no_data_mask[:, :, i]
              valid_data = std_band[~mask]
              if len(valid data) == 0:
                  continue
```

```
# count outliers
outliers_2 = np.sum(np.abs(valid_data) > 2)
outliers_3 = np.sum(np.abs(valid_data) > 3)

pct_2 = (outliers_2 / len(valid_data)) * 100
pct_3 = (outliers_3 / len(valid_data)) * 100

print(band_name + "\t" + str(len(valid_data)) + "\t" + str(outliers_2)_

+ "\t" + str(outliers_3) + "\t" + str(round(pct_2, 1)) + "\t" + \undersymbol{\text{t}}" + \undersymbol{\tex
```

```
[20]: def main():
          print("Problem 2: Band Statistics and Standardization")
          print("=" * 50)
          # load data
          data = load_sentinel2_data()
          # part a: calculate statistics
          stats = calculate_stats(data)
          explain_stats()
          table = make_table(stats)
          # part b: standardize data
          standardized = standardize_data(data)
          explain_standardization()
          # plot histograms
          plot_histograms(data, standardized)
          # analyze outliers
          analyze_outliers(standardized)
          print("\nDone with the analysis..!!!!!")
      if __name__ == "__main__":
          main()
```

Problem 2: Band Statistics and Standardization

```
Data loaded successfully!
Shape: (954, 716, 12)
Calculating band statistics...
B1 done
B2 done
B3 done
B4 done
B5 done
```

B6 done B7 done B8 done B8A done B9 done B11 done B12 done

What these statistics mean:

Mean: average value

Std: how spread out the data is Min/Max: smallest and largest values Q1, Median, Q3: quartiles (25%, 50%, 75%) Skewness: if data is symmetric or not Kurtosis: how "heavy" the tails are

Band Statistics Table:

B1				Max	Q1	Median	Q3	Skewness	`
	0.0887	0.0279	0.0333	0.6021	0.0709	0.0829	0.1007	2.9205	
B2	0.0925	0.0350	0.0386	0.7542	0.0716	0.0853	0.1053	4.4789	
В3	0.1055	0.0344	0.0430	0.7484	0.0858	0.0987	0.1168	4.6922	
B4	0.0943	0.0445	0.0326	0.7728	0.0663	0.0850	0.1093	3.2314	
В5	0.1367	0.0409	0.0346	0.8159	0.1152	0.1310	0.1506	3.1080	
В6	0.2436	0.0610	0.0176	0.7830	0.2131	0.2472	0.2830	-0.7085	
В7	0.2858	0.0773	0.0161	0.7859	0.2444	0.2903	0.3377	-0.7101	
В8	0.2914	0.0803	0.0188	0.9030	0.2475	0.2976	0.3464	-0.7161	
B8A	0.3035	0.0822	0.0157	0.7689	0.2605	0.3098	0.3595	-0.8215	
В9	0.3451	0.0779	0.0653	0.6934	0.3052	0.3446	0.3882	-0.3957	
B11	0.1917	0.0484	0.0218	0.8196	0.1710	0.1889	0.2086	0.5219	
B12	0.1292	0.0486	0.0205	0.9295	0.1020	0.1204	0.1427	1.9499	
	B2 B3 B4 B5 B6 B7 B8 B8A B9	B2 0.0925 B3 0.1055 B4 0.0943 B5 0.1367 B6 0.2436 B7 0.2858 B8 0.2914 B8A 0.3035 B9 0.3451 B11 0.1917	B2 0.0925 0.0350 B3 0.1055 0.0344 B4 0.0943 0.0445 B5 0.1367 0.0409 B6 0.2436 0.0610 B7 0.2858 0.0773 B8 0.2914 0.0803 B8A 0.3035 0.0822 B9 0.3451 0.0779 B11 0.1917 0.0484	B2 0.0925 0.0350 0.0386 B3 0.1055 0.0344 0.0430 B4 0.0943 0.0445 0.0326 B5 0.1367 0.0409 0.0346 B6 0.2436 0.0610 0.0176 B7 0.2858 0.0773 0.0161 B8 0.2914 0.0803 0.0188 B8A 0.3035 0.0822 0.0157 B9 0.3451 0.0779 0.0653 B11 0.1917 0.0484 0.0218	B2 0.0925 0.0350 0.0386 0.7542 B3 0.1055 0.0344 0.0430 0.7484 B4 0.0943 0.0445 0.0326 0.7728 B5 0.1367 0.0409 0.0346 0.8159 B6 0.2436 0.0610 0.0176 0.7830 B7 0.2858 0.0773 0.0161 0.7859 B8 0.2914 0.0803 0.0188 0.9030 B8A 0.3035 0.0822 0.0157 0.7689 B9 0.3451 0.0779 0.0653 0.6934 B11 0.1917 0.0484 0.0218 0.8196	B2 0.0925 0.0350 0.0386 0.7542 0.0716 B3 0.1055 0.0344 0.0430 0.7484 0.0858 B4 0.0943 0.0445 0.0326 0.7728 0.0663 B5 0.1367 0.0409 0.0346 0.8159 0.1152 B6 0.2436 0.0610 0.0176 0.7830 0.2131 B7 0.2858 0.0773 0.0161 0.7859 0.2444 B8 0.2914 0.0803 0.0188 0.9030 0.2475 B8A 0.3035 0.0822 0.0157 0.7689 0.2605 B9 0.3451 0.0779 0.0653 0.6934 0.3052 B11 0.1917 0.0484 0.0218 0.8196 0.1710	B2 0.0925 0.0350 0.0386 0.7542 0.0716 0.0853 B3 0.1055 0.0344 0.0430 0.7484 0.0858 0.0987 B4 0.0943 0.0445 0.0326 0.7728 0.0663 0.0850 B5 0.1367 0.0409 0.0346 0.8159 0.1152 0.1310 B6 0.2436 0.0610 0.0176 0.7830 0.2131 0.2472 B7 0.2858 0.0773 0.0161 0.7859 0.2444 0.2903 B8 0.2914 0.0803 0.0188 0.9030 0.2475 0.2976 B8A 0.3035 0.0822 0.0157 0.7689 0.2605 0.3098 B9 0.3451 0.0779 0.0653 0.6934 0.3052 0.3446 B11 0.1917 0.0484 0.0218 0.8196 0.1710 0.1889	B2 0.0925 0.0350 0.0386 0.7542 0.0716 0.0853 0.1053 B3 0.1055 0.0344 0.0430 0.7484 0.0858 0.0987 0.1168 B4 0.0943 0.0445 0.0326 0.7728 0.0663 0.0850 0.1093 B5 0.1367 0.0409 0.0346 0.8159 0.1152 0.1310 0.1506 B6 0.2436 0.0610 0.0176 0.7830 0.2131 0.2472 0.2830 B7 0.2858 0.0773 0.0161 0.7859 0.2444 0.2903 0.3377 B8 0.2914 0.0803 0.0188 0.9030 0.2475 0.2976 0.3464 B8A 0.3035 0.0822 0.0157 0.7689 0.2605 0.3098 0.3595 B9 0.3451 0.0779 0.0653 0.6934 0.3052 0.3446 0.3882 B11 0.1917 0.0484 0.0218 0.8196 0.1710 0.1889	B2 0.0925 0.0350 0.0386 0.7542 0.0716 0.0853 0.1053 4.4789 B3 0.1055 0.0344 0.0430 0.7484 0.0858 0.0987 0.1168 4.6922 B4 0.0943 0.0445 0.0326 0.7728 0.0663 0.0850 0.1093 3.2314 B5 0.1367 0.0409 0.0346 0.8159 0.1152 0.1310 0.1506 3.1080 B6 0.2436 0.0610 0.0176 0.7830 0.2131 0.2472 0.2830 -0.7085 B7 0.2858 0.0773 0.0161 0.7859 0.2444 0.2903 0.3377 -0.7101 B8 0.2914 0.0803 0.0188 0.9030 0.2475 0.2976 0.3464 -0.7161 B8A 0.3035 0.0822 0.0157 0.7689 0.2605 0.3098 0.3595 -0.8215 B9 0.3451 0.0779 0.0653 0.6934 0.3052 0.3446 0.3882

Kurtosis

- 0 23.7974
- 1 43.1080
- 2 46.2931
- 3 23.4076
- 4 27.3514
- 5 3.4061
- 6 1.7201
- 7 1.4222
- 8 1.6743
- 9 2.2100
- 10 6.4363
- 11 8.0381

Standardizing data...

Band 1 standardized

Band 2 standardized

```
Band 3 standardized
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Band 4 standardized

Band 5 standardized

Band 6 standardized

Band 7 standardized

Band 8 standardized

Band 9 standardized

Band 10 standardized

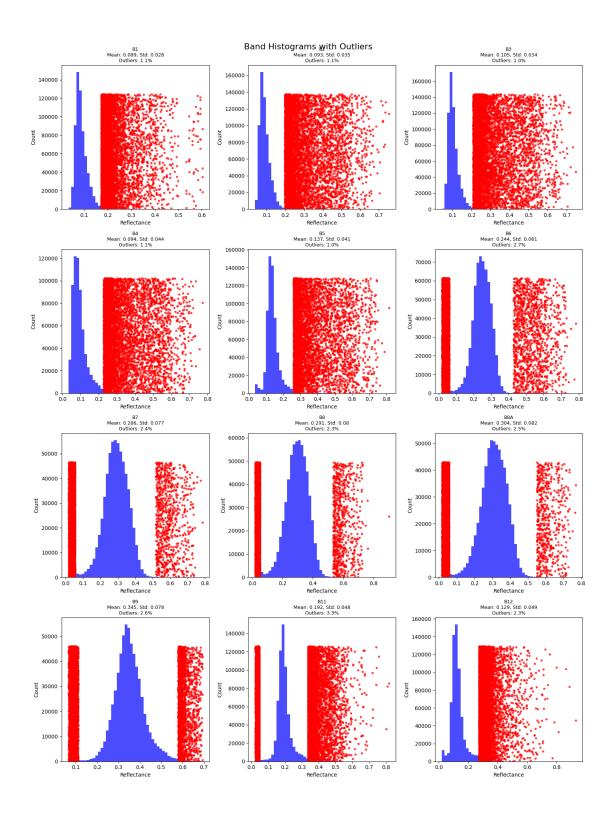
Band 11 standardized

Band 12 standardized

What standardization does:

- Makes mean = 0 and std = 1 for each band
- Allows comparison between different bands
- Helps find outliers (values > 3 or < -3)
- Common preprocessing step

Making histograms...



Histograms saved as histograms.png

Outlier Analysis:

Band	Total	>2	>3	%>2	%>3
B1	250040	20697	7092	8.3	2.8
B2	247924	15450	6636	6.2	2.7
В3	243315	16604	6573	6.8	2.7
B4	243780	25517	6903	10.5	2.8
B5	258693	21809	6261	8.4	2.4
B6	331840	4027	1341	1.2	0.4
B7	330408	4653	841	1.4	0.3
B8	335393	4392	756	1.3	0.2
B8A	336217	4034	611	1.2	0.2
В9	312530	17551	2080	5.6	0.7
B11	288172	24177	6869	8.4	2.4
B12	242053	33001	14252	13.6	5.9

Done!

[]:[