Analysis of the Results

July 2, 2025

Wed Jul 2nd 2025, Jakob Balkovec

This serves as a follow up to the report from O_O2_algorithm_for_patch_exteaction.ipynb. Here I took a look at the results from the algorithm.

```
[]: import os
import json
import pandas as pd
import matplotlib.pyplot as plt
```

0.1 Parsing the Results

0.1.1 Config

```
[]: DIR = "../data/algorithm_output"

FILES = [
    "area.json",
    "area_weights_1.json",
    "area_weights_2.json",
    "area_weights_3.json",
    "area_weights_DBSCAN_1.json",
    "area_weights_DBSCAN_2.json",
    "area_weights_DBSCAN_3.json"
]
```

0.1.2 Reading the Data

```
[]: def load_json_logs(json_dir, filenames):
    all_data = {}
    for fname in filenames:
        full_path = os.path.join(json_dir, fname)
        try:
        with open(full_path, "r") as f:
            data = json.load(f)
            if isinstance(data, list):
                all_data[fname.replace(".json", "")] = data
        else:
            all_data[fname.replace(".json", "")] = [data]
```

0.1.3 Compression

```
[]: def build_comparison_dataframe(all_data):
         rows = []
         for method, entries in all_data.items():
             for entry in entries:
                 candidates = entry.get("candidates", {})
                 picked = entry.get("picked")
                 if isinstance(picked, dict):
                     picked_label = list(picked.keys())[0]
                     picked_score = picked[picked_label]
                 else:
                     picked_label = picked
                     picked_score = candidates.get(picked_label, None)
                 rows.append({
                     "method": method,
                     "picked_label": picked_label,
                     "picked_score": picked_score,
                     "num_candidates": len(candidates),
                     "total score": sum(candidates.values())
                 })
         return pd.DataFrame(rows)
     # call
     comparison_df = build_comparison_dataframe(all_data)
     comparison df["patch_id"] = comparison_df.groupby("method").cumcount()
```

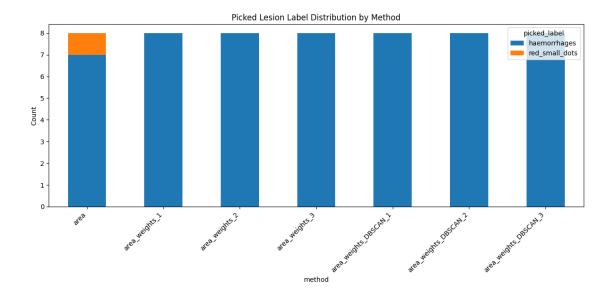
0.1.4 Plot Functions

```
[]: def plot_label_distribution(df):
    label_dist = df.groupby(["method", "picked_label"]).size().
    unstack(fill_value=0)
    label_dist.plot(kind="bar", stacked=True, figsize=(12, 6))
    plt.title("Picked Lesion Label Distribution by Method")
    plt.ylabel("Count")
    plt.xticks(rotation=45, ha="right")
    plt.tight_layout()
```

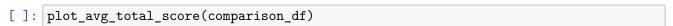
```
plt.show()
def plot_avg_total_score(df):
    df.groupby("method")["total_score"].mean().sort_values().plot(kind="bar",_
 \rightarrowfigsize=(10, 4))
    plt.title("Average Total Candidate Score per Patch")
    plt.ylabel("Avg Total Score")
    plt.xticks(rotation=45, ha="right")
    plt.tight_layout()
    plt.show()
def plot_label_diversity(df):
    diversity = df.groupby("method")["picked_label"].nunique().sort_values()
    diversity.plot(kind="bar", figsize=(10, 4))
    plt.title("Number of Unique Labels Picked per Method")
    plt.ylabel("Unique Picked Labels")
    plt.xticks(rotation=45, ha="right")
    plt.tight_layout()
    plt.show()
def plot_score_distribution_boxplot(df):
    df = df.copy()
    df["picked_pct"] = df["picked_score"] / df["total_score"]
    plt.figure(figsize=(12, 6))
    df.boxplot(column="picked_pct", by="method", grid=False)
    plt.title("Distribution of Picked Score % per Method")
    plt.suptitle("")
    plt.ylabel("Picked Score / Total Score")
    plt.xticks(rotation=45, ha="right")
    plt.tight_layout()
    plt.show()
```

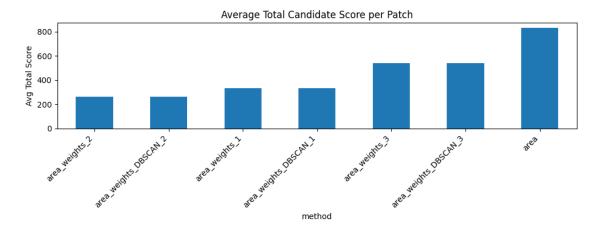
0.2 Results

```
[]: plot_label_distribution(comparison_df)
```



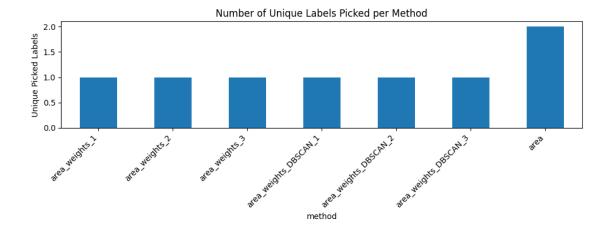
The Area based approach was the only one that picked red_small_dots over hemorrhages.





This visualizes how weights and DBSCAN influence the final score. MAP #2 dominates here since it's risk based. I think this is worth discussing on Monday.

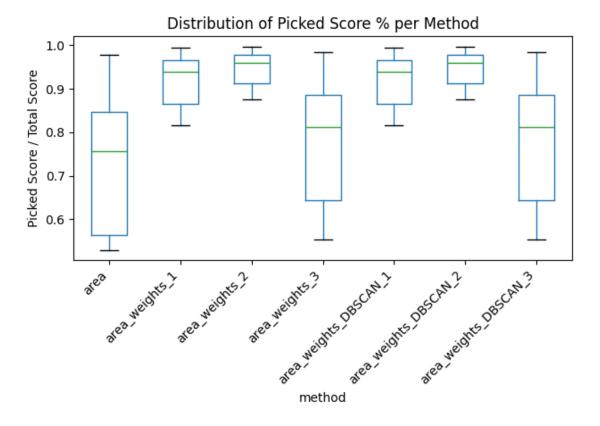
[]: plot_label_diversity(comparison_df)



This is self explanatory. It's only here because I'll be running the same script with different images and hopefully get different results.

[]: plot_score_distribution_boxplot(comparison_df)

<Figure size 1200x600 with 0 Axes>



```
[]: def compute_method_agreement(df):
         df_unique = df.drop_duplicates(subset=["patch_id", "method"])
         pivot = df_unique.pivot_table(index="patch_id", columns="method", __
      ⇔values="picked_label", aggfunc="first")
         pivot["agreement_count"] = pivot.nunique(axis=1)
         total_patches = len(pivot)
         full_agreement = (pivot["agreement_count"] == 1).sum()
         print(f"Total patches: {total_patches}")
         print(f"Full agreement (same label across all methods): {full_agreement}_\( \)
      ⇔patches")
         print(f"Agreement rate: {full_agreement / total_patches:.2%}")
         return pivot
     pivot_df = compute_method_agreement(comparison_df)
     disagreements = pivot_df[pivot_df["agreement_count"] > 1]
    Total patches: 8
    Full agreement (same label across all methods): 7 patches
    Agreement rate: 87.50%
[]: def flag_unique_dominance(pivot_df):
         unique_methods = []
         for idx, row in pivot_df.drop(columns="agreement_count").iterrows():
             label_counts = row.value_counts()
             if label counts.max() == 1:
                 unique_methods.append((idx, row.to_dict()))
         print(f"Unique dominance in {len(unique_methods)} patches (i.e., each_
      →method picked a different label)")
         return pd.DataFrame([{"source_img": idx[0], "patch_no": idx[1], **labels}
                              for idx, labels in unique_methods])
     unique_df = flag_unique_dominance(pivot_df)
```

Unique dominance in 0 patches (i.e., each method picked a different label)

```
[]: # comparison_df.to_csv("comparison_summary.csv", index=False)
```