Approach

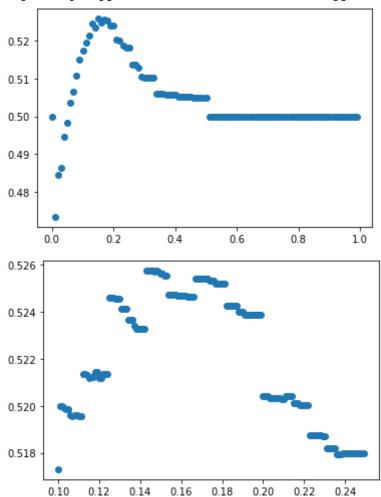
Find a weight for Jaccard similarity (see baseline.ipynb) and recipe popularity percentile (see baselines.ipy), and calculate the weighed average.

If the weighed average >= a value we set (here we use 0.5), the user_id and recipe_id combination is predicted to be actually have happened.

```
In []: # make ipynb files importable
    from jupyter_utils_tddschn import notebook_importer

# import the baseline.ipynb file
    import baseline
```

importing Jupyter notebook from baseline.ipynb



```
0.5255 -

0.5250 -

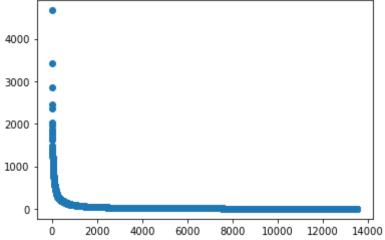
0.5245 -

0.5240 -

0.5235 -

0.1400 0.1425 0.1450 0.1475 0.1500 0.1525 0.1550 0.1575 0.1600
```

```
from baseline import *
In []:
In []:
        d.shape
        X train.shape
        recipes_train = X_train[:, :-1]
        recipes_train = recipes_train.copy()
        recipes_train.resize((X_train.shape[0],))
        recipes_train.shape
        (490000,)
Out[]:
In [ ]:
        from collections import Counter, defaultdict
        import numpy as np
        recipe counts = Counter(recipes train)
In [ ]:
        recipe_counts.most_common(3)
        [(32445558, 4671), (95482435, 3418), (54496210, 2847)]
Out[ ]:
In [ ]:
        most common recipes = recipe counts.most common()
In [ ]:
        most common recipes id normalized = [(i + 1, most common recipes[i][1]) for i :
In [ ]:
        len(most_common_recipes_id_normalized)
        most common recipes id normalized[:3], most common recipes id normalized[-3:]
        ([(1, 4671), (2, 3418), (3, 2847)], [(13514, 1), (13515, 1), (13516, 1)])
Out[ ]:
In [ ]: plt.scatter(*zip(*most common recipes id normalized, strict=True))
        plt.show()
```



```
In [ ]: recipe_quantiles: dict[int, float] = {}
        cum_count = 0
        for recipe, count in most_common_recipes:
                recipe_quantiles[recipe] = cum_count / len(recipes_train)
                cum_count += count
In [ ]: import itertools
        dict(itertools.islice(recipe_quantiles.items(), 5))
Out[]: {32445558: 0.0,
         95482435: 0.00953265306122449,
         54496210: 0.016508163265306124,
         43615275: 0.022318367346938775,
         59129763: 0.027351020408163264}
In []; recipe quantiles values = list(recipe quantiles.values())
        recipe_quantiles_id_normalized: list[int, float] = [(i + 1, recipe_quantiles_va
In [ ]: plt.scatter(*zip(*recipe_quantiles_id_normalized, strict=True))
        plt.show()
        1.0
         0.8
         0.6
         0.4
         0.2
         0.0
```

```
In []: from collections import defaultdict
    jaccard_max_sims_dict: dict[tuple[int, int], float] = defaultdict(float)
    for u, r in X_test_and_random:
          jaccard_max_sims_dict[(u, r)] = jaccard_sim_max(u, r)
```

10000

12000

14000

2000

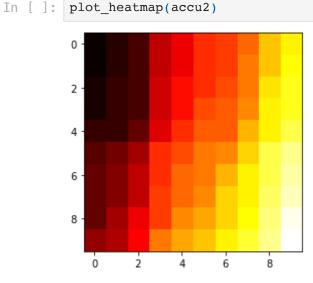
4000

6000

8000

```
In []: def get prediction score(weight1: float, weight2: float, user id: int, recipe i
                 # sim_max = jaccard_sim_max(user_id, recipe id)
                sim_max = jaccard_max_sims_dict[(user_id, recipe_id)]
                return weight1 * sim_max + weight2 * (1 - recipe_quantiles.get(recipe_i
        def get_prediction(weight1: float, weight2: float, user_id: int, recipe_id: int
                return get prediction score(weight1, weight2, user id, recipe id) >= 0.
In [ ]: # get the number of elements in np.arange(start, stop, step)
        def numpy_arange_element_count(start: float, stop: float, step: float) -> int:
                mod = (stop - start) / step
                return int(mod) + int(mod != int(mod))
In [ ]: import seaborn as sns
        def get_accuracies(start, stop, step, start2, stop2, step2) -> list[tuple[float
                accuracies: list[tuple[float, float, float]] = []
                d1, d2 = numpy_arange_element_count(start, stop, step), numpy_arange_el
                for weight1 in np.arange(start, stop, step):
                         for weight2 in np.arange(start2, stop2, step2):
                                 predictions binary = [get prediction(weight1, weight2,
                                 accuracy = sum(predictions_binary) / len(predictions_bi
                                 accuracies.append((weight1, weight2, accuracy))
                arr = np.zeros((d1, d2))
                return accuracies
        def plot_heatmap(accuracies: list[tuple[float, float, float]]) -> None:
                x, y, z = zip(*accuracies)
                heatmap, _, _ = np.histogram2d(x, y, weights=z)
                plt.clf()
                plt.imshow(heatmap, cmap='hot')
                plt.show()
In [ ]: accu = get accuracies(0, 1, 0.1, 0, 1, 0.1)
In [ ]: | accu[:3], accu[8:12]
        accu[:5], accu[-5:]
Out[]: ([(0.0, 0.0, 0.0),
          (0.0, 0.1, 0.0),
          (0.0, 0.2, 0.0),
          (0.0, 0.300000000000004, 0.0),
          (0.0, 0.4, 0.0)],
         [(0.9, 0.5, 0.0004),
          (0.9, 0.60000000000001, 0.00085),
          (0.9, 0.70000000000001, 0.00135),
          (0.9, 0.8, 0.0016),
          (0.9, 0.9, 0.002)])
In [ ]: plot_heatmap(accu)
```

```
0 - 2 - 4 - 6 - 8 - 0 2 4 6 8
```



```
In []: accu3 = get_accuracies(1, 11, 0.5, 1, 11, 0.5)
In []: # accu3[:5], accu3[-5:]
accu3
```

```
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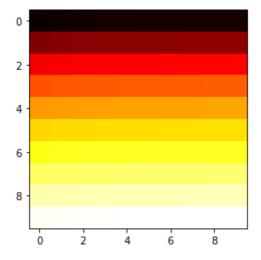
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(10.0, 5.0, 0.42795),
(10.0, 5.5, 0.42815),
(10.0, 6.0, 0.4282),
(10.0, 6.5, 0.42825),
(10.0, 7.0, 0.42845),
(10.0, 7.5, 0.4285),
(10.0, 8.0, 0.42885),
(10.0, 8.5, 0.42895),
(10.0, 9.0, 0.4292),
(10.0, 9.5, 0.4294),
(10.0, 10.0, 0.4295),
(10.0, 10.5, 0.42965),
(10.5, 1.0, 0.43915),
(10.5, 1.5, 0.4401),
(10.5, 2.0, 0.4411),
(10.5, 2.5, 0.44155),
(10.5, 3.0, 0.44205),
(10.5, 3.5, 0.44235),
(10.5, 4.0, 0.4427),
(10.5, 4.5, 0.44305),
(10.5, 5.0, 0.44345),
(10.5, 5.5, 0.4437),
(10.5, 6.0, 0.44385),
(10.5, 6.5, 0.444),
(10.5, 7.0, 0.4441),
(10.5, 7.5, 0.4442),
(10.5, 8.0, 0.44455),
(10.5, 8.5, 0.4447),
(10.5, 9.0, 0.4449),
(10.5, 9.5, 0.44495),
(10.5, 10.0, 0.44505),
(10.5, 10.5, 0.4452)]
```

In []: plot_heatmap(accu3)



In the weight ranges used by accu3, weight1 dominates.

```
In [ ]: |
        accu4 = get_accuracies(11, 21, 0.5, 15, 16, 1)
        accu4
        [(11.0, 15, 0.4598),
Out[]:
          (11.5, 15, 0.4725),
          (12.0, 15, 0.48775),
          (12.5, 15, 0.5004),
          (13.0, 15, 0.513),
          (13.5, 15, 0.52475),
          (14.0, 15, 0.5362),
          (14.5, 15, 0.54705),
          (15.0, 15, 0.55455),
          (15.5, 15, 0.5641),
          (16.0, 15, 0.5726),
          (16.5, 15, 0.5813),
          (17.0, 15, 0.5894),
          (17.5, 15, 0.59615),
          (18.0, 15, 0.6026),
          (18.5, 15, 0.60945),
          (19.0, 15, 0.61465),
          (19.5, 15, 0.6197),
          (20.0, 15, 0.6243),
          (20.5, 15, 0.6285)]
In [ ]:
        accu5 = get_accuracies(21, 41, 1, 15, 16, 1)
         accu5
        [(21, 15, 0.633),
Out[]:
          (22, 15, 0.64095),
          (23, 15, 0.64825),
          (24, 15, 0.65405),
          (25, 15, 0.6597),
          (26, 15, 0.66435),
          (27, 15, 0.66885),
          (28, 15, 0.67275),
          (29, 15, 0.6766),
          (30, 15, 0.6794),
          (31, 15, 0.6824),
          (32, 15, 0.6851),
          (33, 15, 0.68785),
          (34, 15, 0.69075),
          (35, 15, 0.69325),
          (36, 15, 0.69515),
          (37, 15, 0.69745),
          (38, 15, 0.69985),
          (39, 15, 0.70185),
          (40, 15, 0.70395)]
In []:
        accu6 = get_accuracies(41, 61, 0.5, 15, 16, 1)
        accu6
```

```
Out[]: [(41.0, 15, 0.7052),
          (41.5, 15, 0.7057),
          (42.0, 15, 0.70665),
          (42.5, 15, 0.7072),
          (43.0, 15, 0.70795),
          (43.5, 15, 0.7086),
          (44.0, 15, 0.70945),
          (44.5, 15, 0.7101),
          (45.0, 15, 0.71075),
          (45.5, 15, 0.7115),
          (46.0, 15, 0.71205),
          (46.5, 15, 0.7126),
          (47.0, 15, 0.7132),
          (47.5, 15, 0.7135),
          (48.0, 15, 0.71375),
          (48.5, 15, 0.7143),
          (49.0, 15, 0.71435),
          (49.5, 15, 0.71495),
          (50.0, 15, 0.71535),
          (50.5, 15, 0.7159),
          (51.0, 15, 0.7162),
          (51.5, 15, 0.7163),
          (52.0, 15, 0.71675),
          (52.5, 15, 0.71715),
          (53.0, 15, 0.71755),
          (53.5, 15, 0.71755),
          (54.0, 15, 0.71815),
          (54.5, 15, 0.71855),
          (55.0, 15, 0.7189),
          (55.5, 15, 0.7194),
          (56.0, 15, 0.71965),
          (56.5, 15, 0.71995),
          (57.0, 15, 0.7202),
          (57.5, 15, 0.72065),
          (58.0, 15, 0.7211),
          (58.5, 15, 0.72135),
          (59.0, 15, 0.72195),
          (59.5, 15, 0.7225),
          (60.0, 15, 0.7227),
          (60.5, 15, 0.7229)]
In [ ]: accu7 = get accuracies(11, 101, 2, 15, 16, 1)
         w1_{to} = [(x, y) \text{ for } x, _, y \text{ in } accu7]
         w1 to accu
```

```
Out[]: [(11, 0.4598),
          (13, 0.513),
          (15, 0.55455),
          (17, 0.5894),
          (19, 0.61465),
          (21, 0.633),
          (23, 0.64825),
          (25, 0.6597),
          (27, 0.66885),
          (29, 0.6766),
          (31, 0.6824),
          (33, 0.68785),
          (35, 0.69325),
          (37, 0.69745),
          (39, 0.70185),
          (41, 0.7052),
          (43, 0.70795),
          (45, 0.71075),
          (47, 0.7132),
          (49, 0.71435),
          (51, 0.7162),
          (53, 0.71755),
          (55, 0.7189),
          (57, 0.7202),
          (59, 0.72195),
          (61, 0.7235),
          (63, 0.72495),
          (65, 0.72625),
          (67, 0.72785),
          (69, 0.7287),
          (71, 0.7295),
          (73, 0.7305),
          (75, 0.7311),
          (77, 0.7321),
          (79, 0.7333),
          (81, 0.73385),
          (83, 0.7347),
          (85, 0.73545),
          (87, 0.73655),
          (89, 0.73725),
          (91, 0.73805),
          (93, 0.73875),
          (95, 0.73905),
          (97, 0.73965),
          (99, 0.74)]
In [ ]: accu8 = get accuracies(101, 201, 10, 15, 16, 1)
         w1_{to} = [(x, y) \text{ for } x, _, y \text{ in } accu8]
         w1 to accu
        [(101, 0.7405),
Out[]:
          (111, 0.74235),
          (121, 0.74335),
          (131, 0.74415),
          (141, 0.7448),
          (151, 0.74595),
          (161, 0.74615),
          (171, 0.7479),
          (181, 0.7488),
          (191, 0.74945)]
```

```
In [ ]: accu9 = get_accuracies(201, 2701, 100, 15, 16, 1)
         w1_{to} = [(x, y) \text{ for } x, _, y \text{ in } accu9]
         w1_to_accu
Out[]: [(201, 0.74975),
         (301, 0.7559),
          (401, 0.75715),
          (501, 0.75715),
          (601, 0.75715),
          (701, 0.75715),
          (801, 0.75715),
          (901, 0.75715),
          (1001, 0.75715),
          (1101, 0.75715),
          (1201, 0.75715),
          (1301, 0.75715),
          (1401, 0.75715),
          (1501, 0.75715),
          (1601, 0.75715),
          (1701, 0.75715),
          (1801, 0.75715),
          (1901, 0.75715),
          (2001, 0.75715),
          (2101, 0.75715),
          (2201, 0.75715),
          (2301, 0.75715),
          (2401, 0.75715),
          (2501, 0.75715),
          (2601, 0.75715)]
         w1 doesn't seem to matter when it reaches 400.
In []: accu9 = get accuracies(400, 401, 1, 10, 200, 10)
         w2_{to} = [(x, y) \text{ for } _, x, y \text{ in } accu9]
         w2 to accu
Out[]: [(10, 0.75695),
          (20, 0.7573),
          (30, 0.7576),
          (40, 0.75775),
          (50, 0.7581),
          (60, 0.7583),
          (70, 0.75835),
          (80, 0.7585),
          (90, 0.7585),
          (100, 0.7586),
          (110, 0.7587),
          (120, 0.75885),
          (130, 0.7589),
          (140, 0.7589),
          (150, 0.759),
          (160, 0.75905),
          (170, 0.75915),
          (180, 0.75935),
          (190, 0.75945)
In [ ]: accu10 = get_accuracies(400, 401, 1, 200, 3000, 100)
         w2_{to} = [(x, y) \text{ for } x, y \text{ in } accu10]
         w2_to_accu
```

```
Out[]: [(200, 0.75945),
          (300, 0.76005),
          (400, 0.7605),
          (500, 0.76075),
          (600, 0.76085),
          (700, 0.76095),
          (800, 0.7611),
          (900, 0.7612),
          (1000, 0.7612),
          (1100, 0.7612),
          (1200, 0.7612),
          (1300, 0.7613),
          (1400, 0.76135),
          (1500, 0.76135),
          (1600, 0.7614),
          (1700, 0.76145),
          (1800, 0.76145),
          (1900, 0.7615),
          (2000, 0.7615),
          (2100, 0.7615),
          (2200, 0.7615),
          (2300, 0.7615),
          (2400, 0.7615),
          (2500, 0.7615),
          (2600, 0.7616),
          (2700, 0.7616),
          (2800, 0.7616),
          (2900, 0.7616)]
```

Conclusion

The accuracy increases as both weights increase, and stays 0.76 when weight1 set to >=400 and weight2 set to >= 2600.