

# Stage 3

Dan Howe and Keith Funkhouser  
CS838: Data Science  
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- Describe the type of entity you want to match, briefly describe the two tables (e.g., where did you obtain these tables), list the number of tuples per table.
  - We used the provided Silicon Valley dataset. We chose to match the 'songs' table (downsampled to 3038 tuples) and the 'tracks' table (downsampled to 3500 tuples).
- Describe the blocker that you use and list the number of tuple pairs in the candidate set obtained after the blocking step.
  - We used the overlap blocker on both song name and artist name, with a minimum overlap of one word. That successfully reduced the number of candidates by approximately two orders of magnitude, after which we used a rule based blocker based on the Jaccard similarity of the song title. This reduced the number of tuple pairs to 543.
- List the number of tuple pairs in the sample G that you have labeled.
  - We eliminated 10 tuple pairs due to missing data. That left us with 533 tuple pairs, all of which we labeled.
- For each of the six learning methods provided in Magellan (Decision Tree, Random Forest, SVM, Naive Bayes, Logistic Regression, Linear Regression), report the precision, recall, and F-1 that you obtain when you perform cross validation **for the first time** for these methods on I.
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| Method              | Precision | Recall   | F1       |
|---------------------|-----------|----------|----------|
| Decision Tree       | 0.886904  | 0.885568 | 0.884977 |
| Random Forest       | 0.913180  | 0.942320 | 0.926725 |
| SVM                 | 0.789615  | 0.984525 | 0.875998 |
| Linear Regression   | 0.920335  | 0.937755 | 0.928635 |
| Logistic Regression | 0.926929  | 0.948289 | 0.937370 |
| Naive Bayes         | 0.883408  | 0.653312 | 0.748647 |

- Report which learning based matcher you selected after that cross validation.
  - We selected logistic regression, based on it having the highest average precision score.

- Report all debugging iterations and cross validation iterations that you performed. For each debugging iteration, report (a) what is the matcher that you are trying to debug, and its precision/recall/F-1, (b) what kind of problems you found, and what you did to fix them, (c) the final precision/recall/F-1 that you reached. For each cross validation iteration, report (a) what matchers were you trying to evaluate using the cross validation, and (b) precision/recall/F-1 of those.
  - No debugging iterations were performed.
- Report the final best matcher that you selected, and its precision/recall/F-1.
  - (See above)
- **It is important to note that all precision/recall/F-1 numbers asked for in the aboves are supposed to be numbers obtained via CV on the set I. Do not yet use set J.**
- Now report these numbers:

| Method            | Precision | Recall | F1     |
|-------------------|-----------|--------|--------|
| Decision Tree     | 90.16%    | 94.02% | 92.05% |
| Random Forest     | 95.73%    | 95.73% | 95.73% |
| SVM               | 81.69%    | 99.15% | 89.58% |
| Linear Regression | 96.64%    | 98.29% | 97.46% |
| Naive Bayes       | 97.22%    | 59.83% | 74.07% |

- For the final best matcher Y selected, train it on I, then **report its precision/recall/F-1 on J.**

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|---------------------|--------|--------|--------|
| Logistic Regression | 95.76% | 96.58% | 96.17% |
|---------------------|--------|--------|--------|

- Report approximate time estimates: (a) to do the blocking, (b) to label the data, (c) to find the best matcher.
  - Blocking: 3 hours
  - Label data: 1 hour
  - Best matcher: 2 hours
- Provide a discussion on why you didn't reach higher recall, and what you can do in the future to obtain higher recall.
  - To obtain higher recall, we should reduce our false negatives. There were only 4 false negatives produced by the logistic regression model. The major causes were: misspellings in data, short artist names which caused the overlap between artist (in the song table) and list of artists (in the tracks table) to be small, and a mislabeling by us.



