

CAP4770/5771

Lab 1

# Joining DataFrames in Pandas

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# Quick Review: Pandas(Lab0-2)

DataFrame: two dimensional data structures(like SQL table or spreadsheet)

SQL-like operators

- Row filters (selection)
- Column filters (project)
- Form groups (group by)

And more from Python:

1. lambda functions
2. Visualization: matplotlib



# Goal

## Join

- Join operation in Pandas
- Fuzzy join using string edit distance

## Data analysis and visualization

- precision
- Recall



# Join operation in Pandas

SQL:

Join combines records from two or more tables in a relational database

Dataframe in Pandas (equijoin):

```
import pandas as pd
```

```
Students = pd.DataFrame({'student_id': [1, 2], 'name':  
['Alice', 'Bob']})
```

```
Grades = pd.DataFrame({'student_id': [1, 1, 2, 2],  
'class_id': [1, 2, 1, 3], 'grade': ['A', 'C', 'B', 'B']})
```

```
pd.merge(Students, Grades, on='student_id')
```



# Fuzzy join using string edit distance

1. Levenshtein distance
  - a. metric to measure difference between two string sequences
  - b. e.g. kitten → sitting
    - i. kitten → sitten (substitution of "s" for "k")
    - ii. sitten → sittin (substitution of "i" for "e")
    - iii. sittin → sitting (insertion of "g" at the end)

The Levenshtein distance is 3



# Fuzzy join using string edit distance - cont

2. Restaurant data set contains 4 fields:

- a. id: unique for each row
- b. cluster: indicating if two rows are duplicate, i.e. about the same restaurant
- c. name: name of the restaurant(not 100% accurate).
- d. city: location of the restaurant

To find duplicate records: join the restaurant table with itself (self-join) on column 'cluster' (exact search, 100% correct)



# Fuzzy join using string edit distance - cont

3. Fuzzy join and finding duplicate pair
  - a. Cartesian product  
add a **dummy column** to enable product
  - b. Add join criterion  
add a column to product table to store the Levenshtein distance of the names
  - c. Filter the Cartesian product based on join criterion

This way we can also find duplicate records in the data set based on edit distance of names (not 100% correct)



# Data analysis and visualization

To evaluate the fuzzy join accuracy in terms of finding duplicate pairs of records, we use

- Precision  
=  $(\# \text{ correct duplicate pairs found}) / (\# \text{ all similar pairs found})$
- Recall  
=  $(\# \text{ correct duplicate pairs found}) / (\# \text{ all duplicate pairs})$

(All duplicates pairs are stored in DataFrame 'clusters')

(All pairs with different L-distance are stored in 'prod')





# Data analysis and visualization

## - cont

For different Levenshtein-distance thresholds as join filter criterion, visualize a “precision, recall-threshold” graph

1. Another way to visualize to show precision/recall trade-off: precision-recall graph. (DIY4)
2. There is another metric - Levenshtein. ratio that can also be used. Try to compare the two metrics used in the restaurant dataset. (DIY5)