Applied Analytics and Predictive Modeling Spring 2021

Lecture-3

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Some of the slides adapted from Intro to Data Mining Tan et al. 2nd edition

Today's agenda

- Previous week Data Mining; Python Basics; Numpy
- Data Preprocessing
- Python Packages Pandas
- Class exercises

Overview

Core Ideas of Data Mining

- Data and Dimensionality Reduction
- Data Exploration
- Data Visualization
- Association Rules & Recommendation systems
- Classification
- Clustering
- Prediction

Data Exploration

- Data sets are typically large, complex & messy
- Based on the task at hand, we have to process the data
- Use techniques of Reduction and Visualization

Dimensionality Reduction

- Shrinking the complex/large data into simpler/smaller data
- Reducing the number of variables/columns (e.g., principal components) Dimensionality Reduction
- Reducing the number of records/rows (e.g., clustering) -- Sampling

Data Visualization

- Very important to understand the data in particular, to examine the relationships between the attributes
- Graphs and plots of data
- Histograms, boxplots, bar charts, scatterplots
- Research translation exercise helped get some understanding about this

Association Rules

- To identify rules that define "what goes with what" in transactions
- Example: "If X was purchased, Y was also purchased" given a set of transactions
- Very useful in recommendation systems "Our records show you bought X, you may also like Y"
- Also called as "affinity analysis"

Recommender Systems

- Collaborative filtering Technique used by recommendation systems
- The main goal is to recommend items that we may like
- Various aspects that customers view, select, purchase, rate, etc
- User-based recommendation: Recommend products that "customers like you" purchase
- Item-based recommendation: Recommend products that share a "product purchaser profile" with your purchases

Supervised Learning

- Given training data (where the target value is known), the goal is to predict a single "target" or "outcome" variable
- Can be classified into two types Classification and Prediction

Supervised Learning – Classification

- Main aim is to predict an outcome variable (or target variable)
- The target variable can be binary or multi-class
- Examples: Fraudulent or Non-fraudulent transaction; Pass or Fail; Rainy or Sunny; etc.

Supervised Learning – Prediction

- Main aim is to predict the outcome variable (usually in terms of a probability value)
- Common methods that could perform prediction are Regression
- Examples: performance evaluation, revenue estimation, sales percentage, etc

Unsupervised Learning

- Main aim is to segment data into meaningful segments or detect patterns
- There is no target (outcome) variable to predict or classify
- Common methods include clustering.

Handling data

Pandas

- Most popular python library for data analysis
- Highly optimized performance
- Using: 1) Series; 2) DataFrames

Pandas – Series

• One dimensional array to store any data type

>> import pandas as pd

>> a = pd.Series(data, index = Index)

- data can be:
 - Scalar value integer, string
 - Dictionary <key, value> pair
 - Ndarray

• Index by default is from 0, 1, 2, ... (*n*-1) where *n* is the length of the data

>> data = [1, 2, 3, 4, 5, 6, 7] >> s = pd.Series(data) >> Index = ['a', 'b', 'c', 'd', 'e', 'f', 'g']
>> s1 = pd.Series(data, Index)

0	1	a 1	
1	2	b 2	
-	-	c 3	
2	4	d 4	
-		e 5	
4	5	f 6	
5	6	g 7	
6	7		
dtype: int64 dtype: int64			

Pandas – Series

>> diction = {'a': 1, 'b':2, 'c':3, 'd':4, 'e':5, 'f':6, 'g':7}
>> s = pd.Series(dictionary)



 DataFrames is two-dimensional data structure that consists of rows and columns

>> import pandas as pd
>> s = pd.DataFrame(data)

data can be:

- One or more dictionaries
- One or more series
- 2D-numpy Ndarray

>> diction1 ={'a':1, 'b':2, 'c':3, 'd':4}
>> diction2 ={'a':5, 'b':6, 'c':7, 'd':8, 'e':9}
>> data = {'first':diction1, 'second':diction2}
>> df = pd.DataFrame(data)

	first	second
а	1.0	5
b	2.0	6
С	3.0	7
d	4.0	8
е	NaN	9

Pandas – Series vs DataFrames

- Series is 1-D whereas a DataFrame is 2-D.
- A one column DataFrame can have a name for that one column but a Series cannot have a column name.
- Each column of a DataFrame can be converted to a series.

Pandas – DataFrames – Example1

- Create a dataframe using a list

Pandas – DataFrames – Example2

df = pandas.DataFrame(data)0John12print(df)1William15

2 lan 13

Name

Age

3 Noah 12

• Given a small dataset, create a dataframe and print only two columns Name and Address.

Name	Age	Address	Qualification
John	24	New York	MS
Jim	25	Arizona	BS
Ashley	22	Minnesota	BS
Aimee	23	California	MS
Jeff	27	New York	PhD

Pandas – DataFrames – nba.csv

Name	Team	Number	Position	Age	Height	Weight	College	Salary
Avery Bradley	Boston Celtics	0	PG	25	2-Jun	180	Texas	7730337
Jae Crowder	Boston Celtics	99	SF	25	6-Jun	235	Marquette	6796117
R.J. Hunter	Boston Celtics	28	SG	22	5-Jun	185	Georgia State	1148640
Jonas Jerebko	Boston Celtics	8	PF	29	10-Jun	231		5000000
Amir Johnson	Boston Celtics	90	PF	29	9-Jun	240		12000000
Jordan Mickey	Boston Celtics	55	PF	21	8-Jun	235	LSU	1170960
Kelly Olynyk	Boston Celtics	41	С	25	Jul-00	238	Gonzaga	2165160
Terry Rozier	Boston Celtics	12	PG	22	2-Jun	190	Louisville	1824360

Retrieving a player's information
>> Import pandas as pd
>> data = pd.read_csv("nba.csv",
index_col="Name")
>> first = data.loc["Avery Bradley"]
>> second = data.loc["R.J. Hunter"]

>> print(first)
>> print(second)

Team	Boston Ce	ltics	
Number		0	
Position		PG	
Age		25	
Height		6-2	
Weight		180	
College		Texas	
Salary	7.73034	4e+06	
Name: Avery	Bradley,	dtype:	object

Team	Boston Celtics
Number	28
Position	SG
Age	22
Height	6-5
Weight	185
College	Georgia State
Salary	1.14864e+06
Name: R.J.	Hunter, dtype: object

Retrieving a single column
> Import pandas as pd
> data = pd.read_csv("nba.csv", index_col="Name")

>> first = data["Age"]

Jonas Jerebko	29.0
Amir Johnson	29.0
Jordan Mickey	21.0
Kelly Olynyk	25.0
Terry Rozier	22.0
Marcus Smart	22.0
Jared Sullinger	24.0
Isaiah Thomas	27.0
Evan Turner	27.0
James Young	20.0
Tyler Zeller	26.0
Bojan Bogdanovic	27.0
Markel Brown	24.0
Wayne Ellington	28.0
Rondae Hollis-Jefferson	21.0
Jarrett Jack	32.0
Sergey Karasev	22.0
Sean Kilpatrick	26.0
Shane Larkin	23.0
Brook Lopez	28.0
Chris McCullough	21.0
Willie Reed	26.0
Thomas Robinson	25.0
Henry Sims	26.0
Donald Sloan	28.0
Thaddeus Young	27.0
Al-Farouq Aminu	25.0
Pat Connaughton	23.0
Allen Crabbe	24.0
Ed Davis	27.0
	27.0

• How can we print the entire dataframe?

for i, j in df.iterrows(): print(i, j)

Pandas – Missing values

• To check if there are any missing values

Pandas – Fill Missing Values

dict1 = {'First Score':[100, 90, np.nan, 95], 'Second Score': [30, 45, 56, np.nan], 'Third Score':[np.nan, 40, 80, 98]}

df = pd.DataFrame(dict1) df.fillna(0)

Pandas – Drop the rows with missing values

dict1 = {'First Score':[100, 90, np.nan, 95], 'Second Score': [30, 45, 56, np.nan], 'Third Score':[np.nan, 40, 80, 98]}

df = pd.DataFrame(dict1)

df.dropna()

Pandas – ffill()

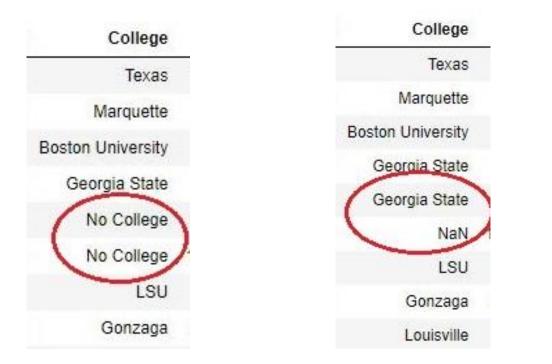
• Missing values are replaced with the previous row's column value



nba["College"].fillna(method ='ffill', inplace = True)

Pandas – ffill()

• We can set a limit (by using *limit*) on successful replacement of NaN values.



• nba["College"].fillna(method ='ffill', limit = 1, inplace = True)

Pandas - Groupby()

- It is used to split the data into groups based on some criteria.
- For example, use the nba.csv to group the data based on the "Team"

```
>> import pandas as pd
>> df = pd.read_csv("nba.csv")
>> df
>> gbdata = df.groupby('Team')
>> gbdata.first()
>> gbmean = df.groupby('age').mean()
>> gbmean
```

Summary of the dataframe

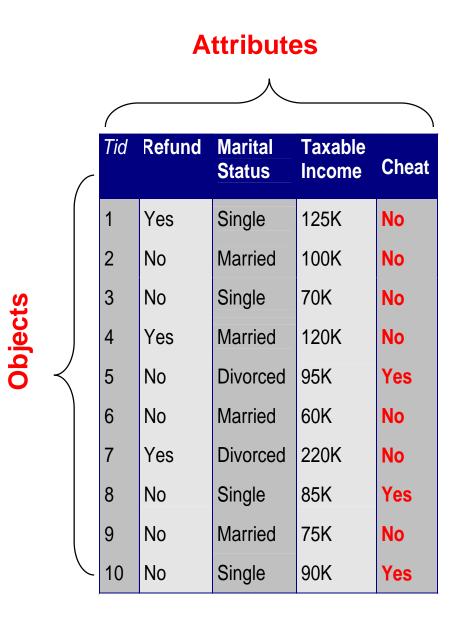
>> print(df.info())

Pandas – Class Exercise

- 1. How can we retrieve a row by their index number? For example, index=3?
- 2. How do you find the number of rows and number of columns in the dataframe?
- 3. For nba.csv dataset, find the number of rows that have missing values.
- 4. Replace all the missing values in nba.csv with 0.

What is data?

- Collection of data objects and their attributes
- According to Tan et al.,
- An **attribute** is a property or characteristic of an object
 - Also known as variable, field, characteristic, dimension, or feature
- A collection of attributes describe an object
 - Also known as tuple, record, point, case, sample, etc.



More views of data

- Data may have parts
- The different parts of data may have relationships
- More generally, data may have structure
- Data can be incomplete