



# Preprocessing Data for Machine Learning

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### What is data preprocessing?

- Beyond cleaning and exploratory data analysis
- Prepping data for modeling
- Modeling in Python requires numerical input



### Refresher on Pandas basics



#### Refresher on Pandas basics

```
In [5]: print(hiking.dtypes)
                   object
Accessible
Difficulty
                   object
Length
                   object
Limited Access
                   object
                   object
Location
                   object
Name
Other Details
                   object
Park Name
                   object
Prop ID
                   object
                  float64
lat
                  float64
lon
dtype: object
```



#### Refresher on Pandas basics

```
In [6]: print(wine.describe())
                                                         Alcalinity of ash
                                Malic acid
                       Alcohol
                                                    Ash
             Type
       178.000000
                    178.000000
                                178.000000
                                             178.000000
count
                                                                 178.000000
         1.938202
                     13.000618
                                  2.336348
                                               2.366517
                                                                  19.494944
mean
         0.775035
                      0.811827
                                  1.117146
                                               0.274344
                                                                   3.339564
std
                     11.030000
                                  0.740000
                                               1.360000
                                                                  10.600000
min
         1.000000
25%
                                                                  17.200000
         1.000000
                     12.362500
                                  1.602500
                                               2.210000
50%
                     13.050000
                                  1.865000
                                               2.360000
                                                                  19.500000
         2.000000
                                  3.082500
                                               2.557500
                                                                  21.500000
75%
         3.000000
                     13.677500
                                  5.800000
                                                                  30.000000
         3.000000
                     14.830000
                                               3.230000
max
```



```
In [7]: print(df)
  1.0
      NaN 2.0
  4.0
       7.0
            3.0
  7.0
       NaN
            NaN
      7.0
  NaN
           NaN
4 5.0
       9.0
           7.0
In [8]: print(df.dropna())
  4.0 7.0 3.0
       9.0 7.0
  5.0
```



```
In [9]: print(df)
  1.0
      NaN 2.0
       7.0
  4.0
            3.0
  7.0
       NaN
            NaN
      7.0
  NaN
           NaN
4 5.0
       9.0
           7.0
In [10]: print(df.drop([1, 2, 3]))
         В
  1.0 NaN 2.0
  5.0
       9.0 7.0
```



```
In [11]: print(df)
         В
  1.0
       NaN 2.0
  4.0
       7.0
            3.0
  7.0
       NaN
            NaN
       7.0
  NaN
            NaN
4 5.0
       9.0
            7.0
In [12]: print(df.drop("A", axis=1))
  NaN 2.0
  7.0
       3.0
   NaN
       NaN
  7.0
       NaN
       7.0
4 9.0
```



```
In [13]: print(df)
  1.0
      NaN 2.0
  4.0
       7.0
            3.0
  7.0
       NaN
            NaN
      7.0
  NaN
           NaN
4 5.0
      9.0
           7.0
In [14]: print(df[df["B"] == 7])
         В
  4.0 7.0 3.0
      7.0 NaN
  NaN
```



```
In [15]: print(df)
  1.0
       NaN 2.0
  4.0
       7.0
            3.0
  7.0
       NaN
            NaN
  NaN
       7.0
            NaN
4 5.0
       9.0
            7.0
In [16]: print(df["B"].isnull().sum())
2
In [17]: print(df[df["B"].notnull()])
  4.0
       7.0
            3.0
       7.0
            NaN
  NaN
  5.0
       9.0
            7.0
```



# Let's practice!





# Working With Data Types

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### Why are types important?

```
In [1]: print(volunteer.dtypes)
opportunity_id
                         int64
content id
                         int64
vol requests
                         int64
event time
                         int64
                        object
title
hits
                         int64
                        object
summary
is_priority
                        object
category id
                       float64
. . .
```

- object: string/mixed types
- int64: integer
- float64: float
- datetype64 (or timedelta): datetime



### Converting column types

```
In [2]: print(df)

A B C
0 1 string 1.0
1 2 string2 2.0
2 3 string3 3.0

In [3]: print(df.dtypes)

A int64
B object
C object
dtype: object
```



### Converting column types

```
In [4]: print(df)

A          B     C
0     1     string    1.0
1     2     string2    2.0
2     3     string3    3.0

In [5]: df["C"] = df["C"].astype("float")
In [6]: print(df.dtypes)

A     int64
B     object
C     float64
dtype: object
```



# Let's practice!





## **Training and Test Sets**

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### Splitting up your dataset

```
In [1]: from sklearn.model selection import train test split
In [2]: X_train, X_test, y_train, y_test = train_test_split(X, y)
   X_train y_train
       1.0
       4.0
       7.0
       2.0
       5.0
       5.0
       6.0
   X_test y_test
      9.0
      1.0
      4.0
               n
```



### Stratified sampling

- 100 samples, 80 class 1 and 20 class 2
- Training set: 75 samples, 60 class 1 and 15 class 2
- Test set: 25 samples, 20 class 1 and 5 class 2



### Stratified sampling

```
In [3]: y["labels"].value counts()
class1
          80
class2
          20
Name: labels, dtype: int64
In [4]: X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y)
In [5]: y train["labels"].value counts()
class1
          60
class2
          15
Name: labels, dtype: int64
In [6]: y_test["labels"].value counts()
class1
          20
class2
Name: labels, dtype: int64
```



# Let's practice!