

Data Preprocessing

Importing Data

`df.to_csv(filename)` # Writes to a CSV file
`df.to_excel(filename)` # Writes to an Excel file
`df.to_sql(table_name, connection_object)` # Writes to a SQL table
`df.to_json(filename)` # Writes to a file in JSON format
`df.to_html(filename)` # Saves as an HTML table
`df.to_clipboard()` # Writes to the clipboard

Exploring Data

`df.shape()` # Prints number of rows and columns in Dataframe
`df.head(n)` # Prints first n rows of the DataFrame
`df.tail(n)` # Prints last n rows of the DataFrame
`df.info()` # Index, Datatype and Memory information
`df.describe()` # Summary statistics for numerical columns
`s.value_counts(dropna=False)` # Views unique values and counts
`df.apply(pd.Series.value_counts)` # Unique values and counts for all columns
`df.describe()` # Summary statistics for numerical columns
`df.mean()` # Returns the mean of all columns
`df.corr()` # Returns the correlation between columns in a DataFrame
`df.count()` # Returns the number of non-null values in each DataFrame column
`df.max()` # Returns the highest value in each column
`df.min()` # Returns the lowest value in each column
`df.median()` # Returns the median of each column
`df.std()` # Returns the standard deviation of each column

Data Selecting

`df[col]` # Returns column with label col as Series
`df[[col1, col2]]` # Returns Columns as a new DataFrame
`s.iloc[0]` # Selection by position (selects first element)
`s.loc[0]` # Selection by index (selects element at index 0)

```
df.iloc[0,:] # First row
df.iloc[0,0] # First element of first column
df.iloc[:, :-1].values # Extract the independent variables (Features)
df.iloc[:, -1].values # Extract the dependent variable (Target)
```

Data Cleaning

```
df.columns = ['a','b','c'] # Renames columns
pd.isnull() # Checks for null Values, Returns Boolean Array
pd.notnull() # Opposite of s.isnull()
df.dropna() # Drops all rows that contain null values
df.dropna(axis=1) # Drops all columns that contain null values
df.dropna(axis=1,thresh=n) # Drops all rows have have less than n non null values
df.fillna(x) # Replaces all null values with x
s.fillna(s.mean()) # Replaces all null values with the mean (mean can be replaced with almost any function from the statistics section)
s.astype(float) # Converts the datatype of the series to float
s.replace(1,'one') # Replaces all values equal to 1 with 'one'
s.replace([1,3],['one','three']) # Replaces all 1 with 'one' and 3 with 'three'
df.rename(columns=lambda x: x + 1) # Mass renaming of columns
df.rename(columns={'old_name': 'new_ name'}) # Selective renaming
df.set_index('column_one') # Changes the index
df.rename(index=lambda x: x + 1) # Mass renaming of index
```

Filter, Sort and Group By

```
df[df[col] > 0.5] # Rows where the col column is greater than 0.5
df[(df[col] > 0.5) & (df[col] < 0.7)] # Rows where 0.5 < col < 0.7
df.sort_values(col1) # Sorts values by col1 in ascending order
df.sort_values(col2,ascending=False) # Sorts values by col2 in descending order
df.sort_values([col1,col2], ascending=[True,False]) # Sorts values by col1 in ascending order then col2 in descending order
df.groupby(col) # Returns a groupby object for values from one column
df.groupby([col1,col2]) # Returns a groupby object values from multiple columns
```

`df.groupby(col1)[col2].mean()` # Returns the mean of the values in col2, grouped by the values in col1 (mean can be replaced with almost any function from the statistics section)

`df.pivot_table(index=col1, values= col2,col3], aggfunc=mean)` # Creates a pivot table that groups by col1 and calculates the mean of col2 and col3

`df.groupby(col1).agg(np.mean)` # Finds the average across all columns for every unique column 1 group

`df.apply(np.mean)` # Applies a function across each column

`df.apply(np.max, axis=1)` # Applies a function across each row

Joining and Combining

`df1.append(df2)` # Adds the rows in df1 to the end of df2 (columns should be identical)

`pd.concat([df1, df2],axis=1)` # Adds the columns in df1 to the end of df2 (rows should be identical)

`df1.join(df2,on=col1,how='inner')` # SQL-style joins the columns in df1 with the columns on df2 where the rows for col have identical values. how can be one of 'left', 'right', 'outer', 'inner'

Writing Data

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Data Transformation and Splitting using sklearn Package

Filling missing value with mean

`from sklearn.preprocessing import Imputer`

`imputer= Imputer(missing_values ='NaN', strategy='mean', axis = 0)`

`imputerimputer= imputer.fit(df[:, 1:3])`

`df[:, 1:3]= imputer.transform(df[:, 1:3])`

`df`

Encoding a variable

`from sklearn.preprocessing import LabelEncoder`

`label_encoder_df= LabelEncoder()`

`df[:, 0]= label_encoder_df.fit_transform(df[:, 0])`

df

Dummy encoding

labelencoder_y= LabelEncoder()

y= labelencoder_y.fit_transform(y)

Train Test split

from sklearn.model_selection import train_test_split

train, test = train_test_split(df, test_size=0.2)

X_train, X_test, y_train, y_test = train_test_split(df[list_of_x_cols], df[y_col], test_size=0.2)