```
In [26]: punctuations = '''!()-{}[];:"'\,<>./?@#$%^&*_~`'''
In [27]: my_str = "Hello!!!, he said ...and went."
In [31]: no_punct =""
In [34]: for char in my_str:
          if(char not in punctuations):
             no_punct = no_punct + char
             print(no_punct)
         Н
         He
         Hel
         Hell
         Hello
         Hello
         Hello h
         Hello he
         Hello he
         Hello he s
         Hello he sa
         Hello he sai
         Hello he said
         Hello he said
         Hello he said a
         Hello he said an
         Hello he said and
         Hello he said and
         Hello he said and w
         Hello he said and we
         Hello he said and wen
         Hello he said and went
In [35]: import re
         s = "Hello!!!, he said ...and went."
         s = re.sub(r'[^\w\s]','',s)
         # not of word and space character
         print(s)
```

Hello he said and went

```
In [40]: import nltk
          nltk.download('punkt')
          from nltk.tokenize import sent tokenize, word tokenize
          example_text = "Hello Mr. Pravin, how are you doing today? The weather in lona
          print(sent tokenize(example text))
          print(word tokenize(example text))
          ['Hello Mr. Pravin, how are you doing today?', 'The weather in lonavala israi
          ny.', 'The sky is full of cloud.']
          ['Hello', 'Mr.', 'Pravin', ',', 'how', 'are', 'you', 'doing', 'today', '?',
          'The', 'weather', 'in', 'lonavala', 'israiny', '.', 'The', 'sky', 'is', 'ful
          l', 'of', 'cloud', '.']
          [nltk data] Downloading package punkt to
                           C:\Users\Pranali\AppData\Roaming\nltk_data...
          [nltk data]
                         Package punkt is already up-to-date!
          [nltk data]
In [41]: import nltk
          nltk.download('stopwords')
          from nltk.corpus import stopwords
          stop_words = stopwords.words('english')
          print(stop words)
          ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you'r
          e", "you've", "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves',
          'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'i
          t', "it's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves',
          'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'ha
          d', 'having', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but',
          'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'wit
          h', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'af
          ter', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off',
          'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when',
          'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'such', 'no', 'nor', 'only', 'own', 'same', 'so', 'th
```

an', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'shoul d', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'might n', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'sh ouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'w ouldn', "wouldn't"]

[nltk data] Downloading package stopwords to [nltk data] C:\Users\Pranali\AppData\Roaming\nltk data... Package stopwords is already up-to-date! [nltk data]

```
In [43]: import nltk
         from nltk.stem import PorterStemmer
         from nltk.tokenize import word_tokenize
         stemmer = PorterStemmer()
         input_str = "There are several types of stemming algorithms."
         input_str = nltk.word_tokenize(input_str)
         print (input_str)
         for word in input str:
          print(stemmer.stem(word))
         ['There', 'are', 'several', 'types', 'of', 'stemming', 'algorithms', '.']
         there
         are
         sever
         type
         of
         stem
         algorithm
In [45]: | import nltk
         nltk.download('wordnet')
         from nltk.stem import WordNetLemmatizer
         from nltk.tokenize import word_tokenize
         lemmatizer = WordNetLemmatizer()
         input_str = "There are several cities with mice."
         input_str = nltk.word_tokenize(input_str)
         print (input str)
         for word in input_str:
          print(lemmatizer.lemmatize(word))
         [nltk data] Downloading package wordnet to
                         C:\Users\Pranali\AppData\Roaming\nltk data...
         [nltk data]
         ['There', 'are', 'several', 'cities', 'with', 'mice', '.']
         There
         are
         several
         city
         with
         mouse
```

```
In [50]:
         import pandas as pd
         import numpy as np
         df = pd.read_csv("C:\\Users\\Pranali\\Downloads\\records.csv")
         df.head()
         print(df)
         df.info()
            ID
                  Name
                          Role Salary
         0
             1
               Pankaj Editor
                                 10000
         1
             2
                  Lisa Editor
                                  8000
         2
             3
                 David Author
                                  6000
         3
             4
                   Ram Author
                                  4000
         4
             5 Anupam Author
                                  5000
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5 entries, 0 to 4
         Data columns (total 4 columns):
              Column Non-Null Count Dtype
          0
              ID
                    5 non-null
                                      int64
          1
              Name
                      5 non-null
                                      object
          2
              Role
                      5 non-null
                                      object
              Salary 5 non-null
                                      int64
         dtypes: int64(2), object(2)
         memory usage: 292.0+ bytes
In [51]:
         updated_df = df.dropna(axis=1)
         print(updated_df)
         updated_df.info()
            ID
                  Name
                          Role Salary
                Pankaj Editor
         0
             1
                                 10000
         1
             2
                  Lisa Editor
                                  8000
         2
             3
                 David Author
                                  6000
         3
             4
                   Ram Author
                                  4000
             5 Anupam Author
                                  5000
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5 entries, 0 to 4
         Data columns (total 4 columns):
              Column Non-Null Count Dtype
          0
              ID
                      5 non-null
                                      int64
                      5 non-null
                                      object
          1
              Name
          2
                      5 non-null
              Role
                                      object
                                      int64
          3
              Salary 5 non-null
         dtypes: int64(2), object(2)
         memory usage: 292.0+ bytes
```

```
In [52]:
         updated df = df.dropna(axis=0)
         print(updated df)
         updated df.info()
             ΙD
                   Name
                           Role
                                 Salary
             1
                Pankaj
                         Editor
                                  10000
             2
                         Editor
         1
                   Lisa
                                   8000
         2
             3
                 David Author
                                   6000
         3
             4
                    Ram Author
                                   4000
         4
             5
                Anupam Author
                                   5000
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5 entries, 0 to 4
         Data columns (total 4 columns):
          #
               Column Non-Null Count Dtype
          0
               ID
                       5 non-null
                                       int64
              Name
          1
                       5 non-null
                                       object
          2
               Role
                       5 non-null
                                       object
              Salary 5 non-null
                                       int64
         dtypes: int64(2), object(2)
         memory usage: 292.0+ bytes
In [53]:
         updated df = df
         updated_df['Salary']=updated_df['Salary'].fillna(updated_df['Salary'].mean())
         print(updated_df)
            ID
                   Name
                           Role
                                 Salary
         0
                Pankaj
                         Editor
                                  10000
             1
         1
             2
                   Lisa
                         Editor
                                   8000
             3
                                   6000
         2
                 David Author
         3
             4
                    Ram Author
                                   4000
         4
                Anupam Author
                                   5000
In [54]:
         updated df = df
         updated df['Salaryismissing'] = updated df['Salary'].isnull()
         print(updated df)
            ID
                   Name
                           Role
                                 Salary
                                         Salaryismissing
         0
             1
                Pankaj
                         Editor
                                  10000
                                                    False
         1
             2
                   Lisa
                         Editor
                                   8000
                                                    False
         2
             3
                 David
                        Author
                                   6000
                                                    False
         3
             4
                    Ram
                        Author
                                   4000
                                                    False
         4
                Anupam Author
                                   5000
                                                    False
```

```
In [55]: testdf = df[df['Salary'].isnull()==True]
    traindf = df[df['Salary'].isnull()==False]
    traindf.drop("Salary",axis=1,inplace=True)
    testdf.drop("Salary",axis=1,inplace=True)
    print(traindf)
```

```
ID
        Name
                Role Salaryismissing
   1 Pankaj Editor
                               False
0
        Lisa Editor
                               False
1
   2
2
   3
       David Author
                               False
3
   4
         Ram Author
                               False
   5 Anupam Author
                               False
```

C:\Users\Pranali\AppData\Local\Temp\ipykernel\_24728\201493639.py:4: SettingWi
thCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user\_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy)

testdf.drop("Salary",axis=1,inplace=True)

```
In [56]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

- In [57]: from sklearn.linear\_model import LinearRegression
  from sklearn.model\_selection import train\_test\_split
- In [58]: data=pd.read\_csv("C:\\Users\\Pranali\\Downloads\\house.csv")

```
In [59]: print(data.columns)
    print(data.head(10))
    print(data.describe())

    print(data.shape)
    print(data.isnull().sum())
```

```
Index(['longitude', 'latitude', 'housing_median_age', 'total_rooms',
        'total_bedrooms', 'population', 'households', 'median_income',
        'median house value'],
      dtype='object')
                         housing_median_age
              latitude
                                               total rooms
                                                             total bedrooms
   longitude
0
     -114.31
                  34.19
                                           15
                                                       5612
                                                                         1283
1
     -114.47
                  34.40
                                           19
                                                       7650
                                                                         1901
2
                                           17
     -114.56
                  33.69
                                                        720
                                                                         174
3
     -114.57
                  33.64
                                           14
                                                       1501
                                                                         337
4
                                           20
                                                                         326
     -114.57
                  33.57
                                                       1454
5
                                           29
     -114.58
                  33.63
                                                       1387
                                                                         236
6
     -114.58
                  33.61
                                           25
                                                       2907
                                                                         680
7
                                           41
     -114.59
                  34.83
                                                        812
                                                                         168
8
     -114.59
                  33.61
                                           34
                                                       4789
                                                                         1175
9
                                                       1497
     -114.60
                  34.83
                                           46
                                                                          309
                households
                             median income
                                             median house value
   population
0
         1015
                        472
                                     1.4936
                                                           66900
1
         1129
                        463
                                     1.8200
                                                           80100
2
           333
                        117
                                     1.6509
                                                           85700
3
           515
                        226
                                     3.1917
                                                           73400
4
           624
                        262
                                     1.9250
                                                           65500
5
           671
                        239
                                     3.3438
                                                           74000
6
         1841
                        633
                                     2.6768
                                                           82400
7
           375
                        158
                                     1.7083
                                                           48500
8
          3134
                       1056
                                     2.1782
                                                            58400
9
           787
                        271
                                     2.1908
                                                           48100
                                     housing_median_age
           longitude
                           latitude
                                                            total_rooms
       17000.000000
                       17000.000000
                                            17000.000000
count
                                                           17000.000000
                          35.625225
        -119.562108
mean
                                                28.589353
                                                            2643.664412
std
            2.005166
                           2.137340
                                                12.586937
                                                             2179.947071
min
        -124.350000
                          32.540000
                                                 1.000000
                                                                2.000000
25%
        -121.790000
                          33.930000
                                                18.000000
                                                             1462.000000
50%
                                                             2127.000000
        -118.490000
                          34.250000
                                                29.000000
75%
                          37.720000
                                                37.000000
                                                             3151.250000
        -118.000000
        -114.310000
                          41.950000
                                                52.000000
                                                           37937.000000
max
       total bedrooms
                           population
                                          households
                                                       median income
count
         17000.000000
                         17000.000000
                                        17000.000000
                                                        17000.000000
                          1429.573941
                                          501.221941
mean
            539.410824
                                                            3.883578
std
            421.499452
                          1147.852959
                                          384.520841
                                                            1.908157
min
              1.000000
                             3.000000
                                            1.000000
                                                            0.499900
25%
            297.000000
                           790.000000
                                          282.000000
                                                            2.566375
50%
            434.000000
                          1167.000000
                                          409.000000
                                                            3.544600
75%
            648.250000
                          1721.000000
                                          605.250000
                                                            4.767000
max
           6445.000000
                         35682.000000
                                         6082.000000
                                                           15.000100
       median house value
count
              17000.000000
mean
             207300.912353
std
             115983.764387
min
              14999.000000
25%
             119400.000000
50%
             180400.000000
75%
             265000.000000
max
             500001.000000
(17000, 9)
```

```
longitude
latitude
                       0
housing_median_age
                       0
total rooms
                       0
total_bedrooms
                       0
population
                       0
households
                       0
median income
                       0
median_house_value
                       0
dtype: int64
```

In [60]: sns.relplot(x='median\_house\_value', y= 'median\_income', hue= 'total\_rooms', da
plt.show()

