Exercise Part 4

Using the meteorite data from the Meteorite_Landings.csv file, create a pivot table that shows both the number of meteorites and the 95th percentile of meteorite mass for those that were found versus observed falling per year from 2005 through 2009 (inclusive). Hint: Be sure to convert the year column to a number as we did in the previous exercise.

```
In [53]: import pandas as pd
    meteorites = pd.read_csv("Meteorite_Landings.csv")

In [54]: #splicing
    meteorites['year'] = meteorites['year'].str.slice(6,10)

In [55]: # convert it to a numeric data type
    meteorites['year'] = pd.to_numeric(meteorites['year'], errors='coerce')

In [56]: # meteorites falling per year from 2005 through 2009
    meteorites = meteorites[(meteorites['year'] <= 2009) & (meteorites['year'] >= 2005)
    meteorites
```

| Out[56]: | | name | id | nametype | recclass | mass (g) | fall | year |
|----------|----|----------|-------|----------|-------------|----------|------|--------|
| | 30 | Almahata | 48915 | Valid | Ureilite-an | 3950.000 | Fell | 2008.0 |

| | | name | id | nametype | recclass | mass (g) | fall | year | reclat |
|-----|----|----------------------|-------|----------|--------------------|------------|-------|--------|-----------|
| | 30 | Almahata Sitta | 48915 | Valid | Ureilite-an | 3950.000 | Fell | 2008.0 | 20.74575 |
| | 49 | Ash Creek | 48954 | Valid | L6 | 9500.000 | Fell | 2009.0 | 31.80500 |
| | 82 | Bassikounou | 44876 | Valid | H5 | 29560.000 | Fell | 2006.0 | 15.78333 |
| 10 | 01 | Berduc | 48975 | Valid | L6 | 270.000 | Fell | 2008.0 | -31.91000 |
| 1- | 48 | Bunburra Rockhole | 48653 | Valid | Eucrite | 324.000 | Fell | 2007.0 | -31.35000 |
| | | | | | | | | | |
| 383 | 96 | Yabrin 003 | 48974 | Valid | Acapulcoite | 21.048 | Found | 2008.0 | 23.31522 |
| 456 | 64 | Yaringie Hill | 48950 | Valid | H5 | 5750.000 | Found | 2006.0 | -32.08287 |
| 456 | 68 | Yarle Lakes 004 | 52945 | Valid | CK4 | 4.600 | Found | 2009.0 | -30.50000 |
| 456 | 74 | Yelland Dry Lake | 52641 | Valid | H4 | 76000.000 | Found | 2007.0 | 39.35067 |
| 456 | 85 | Youxi | 55793 | Valid | Mesosiderite- C | 218000.000 | Found | 2006.0 | 26.06000 |

6974 rows × 10 columns

```
In [73]: # used a built-in function that generates pivot-style spreadsheet into DataFrame
         pivot_table = meteorites.pivot_table(
             index='year',
             columns='fall',
             values='mass (g)',
             aggfunc={'mass (g)': lambda x: x.quantile(0.95), 'fall': 'count'}
```

```
In [74]: # rename the columns for clarification
         pivot_table.columns = ['count_fell', 'count_found', 'mass_95th_fell', 'mass_95th_fo
         # display the table
         pivot_table
```

| Out[74]: | | count_fell | count_found | mass_95th_fell | mass_95th_found |
|----------|--------|------------|-------------|----------------|-----------------|
| | year | | | | |
| | 2005.0 | NaN | 875.0 | NaN | 4500.00 |
| | 2006.0 | 5.0 | 2451.0 | 25008.0 | 1600.50 |
| | 2007.0 | 8.0 | 1181.0 | 89675.0 | 1126.90 |
| | 2008.0 | 9.0 | 948.0 | 106000.0 | 2274.80 |
| | 2009.0 | 5.0 | 1492.0 | 8333.4 | 1397.25 |

Using the meteorite data from the Meteorite_Landings.csv file, compare summary statistics of the mass column for the meteorites that were found versus observed falling.

```
In [81]: # without using .groupby

fell_meteorites = meteorites[meteorites['fall'] == 'Fell']['mass (g)']
found_meteorites = meteorites[meteorites['fall'] == 'Found']['mass (g)']

# compute summary statistics
fell_stats = fell_meteorites.describe()
found_stats = found_meteorites.describe()

# combine results into a DataFrame for comparison
summary_stats = pd.DataFrame({'Fell': fell_stats, 'Found': found_stats})
summary_stats
```

```
Out[81]:
                          Fell
                                      Found
                     27.000000 6.945000e+03
          count
          mean
                  19029.665185 1.573986e+03
                  34081.623779 4.202089e+04
            std
           min
                     18.410000 0.000000e+00
           25%
                    410.000000 7.500000e+00
           50%
                   3950.000000 3.450000e+01
           75%
                   8206.500000 1.970000e+02
           max 110000.000000 3.000000e+06
```

```
In [80]: # statistics using .groupby
summary_stats = meteorites.groupby('fall')['mass (g)'].describe()
summary_stats
```

| Out[80]: | count | | mean std | | min | 25% | 50% | 75% | max | |
|----------|-------|--------|--------------|--------------|-------|-------|--------|--------|-----------|--|
| | fall | | | | | | | | | |
| | Fell | 27.0 | 19029.665185 | 34081.623779 | 18.41 | 410.0 | 3950.0 | 8206.5 | 110000.0 | |
| | Found | 6945.0 | 1573.986245 | 42020.893987 | 0.00 | 7.5 | 34.5 | 197.0 | 3000000.0 | |

Exercise Part 5

Using the taxi trip data in the 2019_Yellow_Taxi_Trip_Data.csv file, resample the data to an hourly frequency based on the dropoff time. Calculate the total trip_distance, fare_amount, tolls_amount, and tip_amount, then find the 5 hours with the most tips.

```
In [78]: import pandas as pd

taxis = pd.read_csv('2019_Yellow_Taxi_Trip_Data.csv')

# dropoff time column to datetime format convertion
taxis['tpep_dropoff_datetime'] = pd.to_datetime(taxis['tpep_dropoff_datetime'])

# resample data to hourly frequency based on dropoff time
hourly_data = taxis.resample('h', on='tpep_dropoff_datetime').sum()[['trip_distance'])

# top 5 hours with the highest tip amounts
most_tips = hourly_data.nlargest(5, 'tip_amount')

most_tips
```

trip_distance fare_amount tolls_amount tip_amount

| tpep | dropoff | datetime | |
|------|---------|----------|--|

Out[78]:

| 10676.95 | 67797.76 | 699.04 | 12228.64 |
|----------|------------------------------|---|--|
| 16052.83 | 70131.91 | 4044.04 | 12044.03 |
| 3104.56 | 11565.56 | 1454.67 | 1907.64 |
| 14.34 | 213.50 | 0.00 | 51.75 |
| 98.59 | 268.00 | 24.48 | 25.74 |
| | 16052.83 3104.56 14.34 | 16052.83 70131.91 3104.56 11565.56 14.34 213.50 | 16052.83 70131.91 4044.04 3104.56 11565.56 1454.67 14.34 213.50 0.00 |

```
In [ ]:
```