


```
In [6]: import pandas as pd
```

```
In [3]: meteorites = pd.read_csv("Meteorite_Landings.csv", nrows = 5)
meteorites
```

```
Out[3]:
```

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
0	Aachen	1	Valid	L5	21	Fell	01/01/1880 12:00:00 AM	50.77500	6.08333
1	Aarhus	2	Valid	H6	720	Fell	01/01/1951 12:00:00 AM	56.18333	10.23333
2	Abee	6	Valid	EH4	107000	Fell	01/01/1952 12:00:00 AM	54.21667	-113.00000
3	Acapulco	10	Valid	Acapulcoite	1914	Fell	01/01/1976 12:00:00 AM	16.88333	-99.90000
4	Achiras	370	Valid	L6	780	Fell	01/01/1902 12:00:00 AM	-33.16667	-64.95000



```
In [4]: meteorites.name
```

```
Out[4]: 0    Aachen
1    Aarhus
2     Abee
3  Acapulco
4   Achiras
Name: name, dtype: object
```

```
In [7]: meteorites["name"]
```

```
Out[7]: 0    Aachen
1    Aarhus
2     Abee
3  Acapulco
4   Achiras
Name: name, dtype: object
```

```
In [5]: meteorites.columns
```

```
Out[5]: Index(['name', 'id', 'nametype', 'recclass', 'mass (g)', 'fall', 'year',
              'reclat', 'reclong', 'GeoLocation'],
              dtype='object')
```

```
In [6]: meteorites.index
```

```
Out[6]: RangeIndex(start=0, stop=5, step=1)
```

```
In [17]: import requests

response = requests.get(
    "https://data.nasa.gov/resource/gh4g-9sfh.json",
    params = {"$limit": 50_000}
)

if response.ok:
    payload = response.json()
else:
    print(f"Request was not successful and returned code: {response.status_code}.")
    payload = None
```


```
In [20]: payload[0]
```

```
Out[20]: {'name': 'Aachen',
'id': '1',
'nametype': 'Valid',
'recclass': 'L5',
'mass': '21',
'fall': 'Fell',
'year': '1880-01-01T00:00:00.000',
'reclat': '50.775000',
'reclong': '6.083330',
'geolocation': {'latitude': '50.775', 'longitude': '6.08333'}}
```

```
In [22]: df = pd.DataFrame(payload)
df.head(3)
```

```
Out[22]:
```

	name	id	nametype	recclass	mass	fall	year	reclat	reclong	g
0	Aachen	1	Valid	L5	21	Fell	1880-01-01T00:00:00.000	50.775000	6.083330	
1	Aarhus	2	Valid	H6	720	Fell	1951-01-01T00:00:00.000	56.183330	10.233330	
2	Abee	6	Valid	EH4	107000	Fell	1952-01-01T00:00:00.000	54.216670	-113.000000	



```
In [25]: meteorites = pd.read_csv("Meteorite_Landings.csv")
meteorites.shape
```

```
Out[25]: (45716, 10)
```

```
In [26]: meteorites.columns
```

```
Out[26]: Index(['name', 'id', 'nametype', 'recclass', 'mass (g)', 'fall', 'year',  
              'reclat', 'reclong', 'GeoLocation'],  
              dtype='object')
```

```
In [27]: meteorites.dtypes
```

```
Out[27]: name          object  
         id            int64  
         nametype      object  
         recclass      object  
         mass (g)      float64  
         fall          object  
         year          object  
         reclat        float64  
         reclong       float64  
         GeoLocation   object  
         dtype: object
```

```
In [30]: meteorites.head(10)
```

Out[30]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
<b>0</b>	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	50.77500	6.08333
<b>1</b>	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	56.18333	10.23333
<b>2</b>	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	54.21667	-113.00000
<b>3</b>	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	16.88333	-99.90000
<b>4</b>	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	-33.16667	-64.95000
<b>5</b>	Adhi Kot	379	Valid	EH4	4239.0	Fell	01/01/1919 12:00:00 AM	32.10000	71.80000
<b>6</b>	Adzhi-Bogdo (stone)	390	Valid	LL3-6	910.0	Fell	01/01/1949 12:00:00 AM	44.83333	95.16667
<b>7</b>	Agen	392	Valid	H5	30000.0	Fell	01/01/1814 12:00:00 AM	44.21667	0.61667
<b>8</b>	Aguada	398	Valid	L6	1620.0	Fell	01/01/1930 12:00:00 AM	-31.60000	-65.23333
<b>9</b>	Aguila Blanca	417	Valid	L	1440.0	Fell	01/01/1920 12:00:00 AM	-30.86667	-64.55000



In [31]:

```
meteorites.tail(5)
```

Out[31]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	r
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	01/01/1990 12:00:00 AM	29.03700	17
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	01/01/1999 12:00:00 AM	13.78333	8
45713	Zlin	30410	Valid	H4	3.3	Found	01/01/1939 12:00:00 AM	49.25000	17
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	01/01/2003 12:00:00 AM	49.78917	41
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	01/01/1976 12:00:00 AM	33.98333	-115

In [32]: meteorites.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 45716 entries, 0 to 45715
Data columns (total 10 columns):
#   Column          Non-Null Count  Dtype
---  -
0   name            45716 non-null  object
1   id              45716 non-null  int64
2   nametype        45716 non-null  object
3   recclass        45716 non-null  object
4   mass (g)        45585 non-null  float64
5   fall            45716 non-null  object
6   year            45425 non-null  object
7   reclat          38401 non-null  float64
8   reclong         38401 non-null  float64
9   GeoLocation     38401 non-null  object
dtypes: float64(3), int64(1), object(6)
memory usage: 3.5+ MB
```

In [43]: meteorites[["name", "year"]]

Out[43]:

	name	year
0	Aachen	01/01/1880 12:00:00 AM
1	Aarhus	01/01/1951 12:00:00 AM
2	Abee	01/01/1952 12:00:00 AM
3	Acapulco	01/01/1976 12:00:00 AM
4	Achiras	01/01/1902 12:00:00 AM
...	...	...
45711	Zillah 002	01/01/1990 12:00:00 AM
45712	Zinder	01/01/1999 12:00:00 AM
45713	Zlin	01/01/1939 12:00:00 AM
45714	Zubkovsky	01/01/2003 12:00:00 AM
45715	Zulu Queen	01/01/1976 12:00:00 AM

45716 rows × 2 columns

```
In [44]: meteorites[["name", "mass (g)"]]
```

Out[44]:

	name	mass (g)
0	Aachen	21.0
1	Aarhus	720.0
2	Abee	107000.0
3	Acapulco	1914.0
4	Achiras	780.0
...	...	...
45711	Zillah 002	172.0
45712	Zinder	46.0
45713	Zlin	3.3
45714	Zubkovsky	2167.0
45715	Zulu Queen	200.0

45716 rows × 2 columns

```
In [45]: meteorites[100:104]
```

Out[45]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclon
100	Benton	5026	Valid	LL6	2840.0	Fell	01/01/1949 12:00:00 AM	45.95000	-67.5500
101	Berduc	48975	Valid	L6	270.0	Fell	01/01/2008 12:00:00 AM	-31.91000	-58.3283
102	Béréba	5028	Valid	Eucrite-mmict	18000.0	Fell	01/01/1924 12:00:00 AM	11.65000	-3.6500
103	Berlanguillas	5029	Valid	L6	1440.0	Fell	01/01/1811 12:00:00 AM	41.68333	-3.8000

In [49]: meteorites.iloc[100:104, [0, 3, 4, 6]]

Out[49]:

	name	recclass	mass (g)	year
100	Benton	LL6	2840.0	01/01/1949 12:00:00 AM
101	Berduc	L6	270.0	01/01/2008 12:00:00 AM
102	Béréba	Eucrite-mmict	18000.0	01/01/1924 12:00:00 AM
103	Berlanguillas	L6	1440.0	01/01/1811 12:00:00 AM

In [50]: meteorites.loc[100:104, "mass (g)": "year"]

Out[50]:

	mass (g)	fall	year
100	2840.0	Fell	01/01/1949 12:00:00 AM
101	270.0	Fell	01/01/2008 12:00:00 AM
102	18000.0	Fell	01/01/1924 12:00:00 AM
103	1440.0	Fell	01/01/1811 12:00:00 AM
104	960.0	Fell	01/01/2004 12:00:00 AM

In [56]: meteorites.iloc[-1, [-1]]

Out[56]: GeoLocation (33.98333, -115.68333)  
Name: 45715, dtype: object

In [57]: meteorites.iloc[:, [-1]]

Out[57]:

GeoLocation	
0	(50.775, 6.08333)
1	(56.18333, 10.23333)
2	(54.21667, -113.0)
3	(16.88333, -99.9)
4	(-33.16667, -64.95)
...	...
45711	(29.037, 17.0185)
45712	(13.78333, 8.96667)
45713	(49.25, 17.66667)
45714	(49.78917, 41.5046)
45715	(33.98333, -115.68333)

45716 rows × 1 columns

```
In [60]: (meteorites["mass (g)"] > 50) & (meteorites.fall == "Found")
```

```
Out[60]: 0      False
          1      False
          2      False
          3      False
          4      False
          ...
          45711   True
          45712   False
          45713   False
          45714   True
          45715   True
          Length: 45716, dtype: bool
```

```
In [61]: meteorites[(meteorites["mass (g)"] > 1e6) & (meteorites.fall == "Fell")]
```



Out[61]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
29	Allende	2278	Valid	CV3	2000000.0	Fell	01/01/1969 12:00:00 AM	26.96667	-105.3166
419	Jilin	12171	Valid	H5	4000000.0	Fell	01/01/1976 12:00:00 AM	44.05000	126.1666
506	Kunya-Urgench	12379	Valid	H5	1100000.0	Fell	01/01/1998 12:00:00 AM	42.25000	59.2000
707	Norton County	17922	Valid	Aubrite	1100000.0	Fell	01/01/1948 12:00:00 AM	39.68333	-99.8666
920	Sikhote-Alin	23593	Valid	Iron, IIAB	23000000.0	Fell	01/01/1947 12:00:00 AM	46.16000	134.6533

In [64]:

```
meteorites.query("`mass (g)` > 1e6 and fall == 'Fell'")
```

Out[64]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
29	Allende	2278	Valid	CV3	2000000.0	Fell	01/01/1969 12:00:00 AM	26.96667	-105.3166
419	Jilin	12171	Valid	H5	4000000.0	Fell	01/01/1976 12:00:00 AM	44.05000	126.1666
506	Kunya-Urgench	12379	Valid	H5	1100000.0	Fell	01/01/1998 12:00:00 AM	42.25000	59.2000
707	Norton County	17922	Valid	Aubrite	1100000.0	Fell	01/01/1948 12:00:00 AM	39.68333	-99.8666
920	Sikhote-Alin	23593	Valid	Iron, IIAB	23000000.0	Fell	01/01/1947 12:00:00 AM	46.16000	134.6533

In [65]:

```
meteorites.value_counts()
```

```
Out[65]:
```

name	id	nametype	recclass	mass (g)	fall	year
Aachen	1	Valid	L5	21.00	Fell	01/01/188
0 12:00:00 AM 50.775000 6.083330	(50.775, 6.08333)				1	
Queen Alexandra Range 90215	19020	Valid	L5	358.90	Found	01/01/199
0 12:00:00 AM -84.606700 162.167080	(-84.6067, 162.16708)				1	
Queen Alexandra Range 90217	19022	Valid	L5	327.70	Found	01/01/199
0 12:00:00 AM -84.595030 162.319900	(-84.59503, 162.3199)				1	
Queen Alexandra Range 90218	19023	Valid	L5	926.50	Found	01/01/199
0 12:00:00 AM -84.610170 162.153880	(-84.61017, 162.15388)				1	
Queen Alexandra Range 90219	19024	Valid	L5	316.00	Found	01/01/199
0 12:00:00 AM -84.611160 162.145660	(-84.61116, 162.14566)				1	
..						
Grove Mountains 022004	47902	Valid	H5	0.59	Found	01/01/200
3 12:00:00 AM -72.776389 75.348333	(-72.776389, 75.348333)				1	
Grove Mountains 022005	47903	Valid	L6	8.93	Found	01/01/200
3 12:00:00 AM -72.776389 75.348611	(-72.776389, 75.348611)				1	
Grove Mountains 022006	49930	Valid	L6	1.82	Found	01/01/200
3 12:00:00 AM -72.776390 75.348890	(-72.77639, 75.34889)				1	
Grove Mountains 022011	47904	Valid	L5	2.63	Found	01/01/200
3 12:00:00 AM -72.780556 75.345278	(-72.780556, 75.345278)				1	
Święcany	47342	Valid	L/LL5	8.00	Found	01/01/200
4 12:00:00 AM 49.791390 21.257780	(49.79139, 21.25778)				1	

Name: count, Length: 38115, dtype: int64

```
In [68]: meteorites.fall.value_counts()
```

```
Out[68]:
```

fall	count
Found	44609
Fell	1107

Name: count, dtype: int64

```
In [66]: meteorites.value_counts(subset=["nametype", "fall"], normalize = True)
```

```
Out[66]:
```

nametype	fall	proportion
Valid	Found	0.974145
	Fell	0.024215
Relict	Found	0.001641

Name: proportion, dtype: float64

```
In [82]: meteorites["mass (g)"].mean()
```

```
Out[82]: 13278.078548601512
```

```
In [72]: meteorites["mass (g)"].quantile([0.01, 0.05, 0.5, 0.95, 0.99])
```

```
Out[72]:
```

quantile	mass (g)
0.01	0.44
0.05	1.10
0.50	32.60
0.95	4000.00
0.99	50600.00

Name: mass (g), dtype: float64

```
In [76]: meteorites["mass (g)"].median()
```

Out[76]: 32.6

```
In [77]: meteorites["mass (g)"].max()
```

Out[77]: 60000000.0

```
In [81]: meteorites.loc[meteorites["mass (g)"].idxmax()]
```

Out[81]:

name	Hoba
id	11890
nametype	Valid
recclass	Iron, IVB
mass (g)	60000000.0
fall	Found
year	01/01/1920 12:00:00 AM
reclat	-19.58333
reclong	17.91667
GeoLocation	(-19.58333, 17.91667)
Name	16392, dtype: object

```
In [83]: meteorites.recclass.nunique()
```

Out[83]: 466

```
In [84]: meteorites.recclass.unique()
```

```
Out[84]: array(['L5', 'H6', 'EH4', 'Acapulcoite', 'L6', 'LL3-6', 'H5', 'L',
'Diogenite-pm', 'Unknown', 'H4', 'H', 'Iron, IVA', 'CR2-an', 'LL5',
'CI1', 'L/LL4', 'Eucrite-mmict', 'CV3', 'Ureilite-an',
'Stone-unc1', 'L3', 'Angrite', 'LL6', 'L4', 'Aubrite',
'Iron, IIAB', 'Iron, IAB-sLL', 'Iron, ungrouped', 'CM2', 'OC',
'Mesosiderite-A1', 'LL4', 'C2-ung', 'LL3.8', 'Howardite',
'Eucrite-pmict', 'Diogenite', 'LL3.15', 'LL3.9', 'Iron, IAB-MG',
'H/L3.9', 'Iron?', 'Eucrite', 'H4-an', 'L/LL6', 'Iron, IIIAB',
'H/L4', 'H4-5', 'L3.7', 'LL3.4', 'Martian (chassignite)', 'EL6',
'H3.8', 'H3-5', 'H5-6', 'Mesosiderite', 'H5-7', 'L3-6', 'H4-6',
'Ureilite', 'Iron, IID', 'Mesosiderite-A3/4', 'CO3.3', 'H3',
'EH3/4-an', 'Iron, IIE', 'L/LL5', 'H3.7', 'CBa', 'H4/5', 'H3/4',
'H?', 'H3-6', 'L3.4', 'Iron, IAB-sHL', 'L3.7-6', 'EH7-an', 'Iron',
'CR2', 'CO3.2', 'K3', 'L5/6', 'CK4', 'Iron, IIE-an', 'L3.6',
'LL3.2', 'Pallasite', 'CO3.5', 'Lodranite', 'Mesosiderite-A3',
'L3-4', 'H5/6', 'Pallasite, PMG', 'Eucrite-cm', 'L5-6', 'CO3.6',
'Martian (nakhlite)', 'LL3.6', 'C3-ung', 'H3-4', 'CO3.4', 'EH3',
'Iron, IAB-ung', 'Winonaite', 'LL', 'Eucrite-br', 'Iron, IIF',
'R3.8-6', 'L4-6', 'EH5', 'LL3.00', 'H3.4', 'Martian (shergottite)',
'Achondrite-ung', 'LL3.3', 'C', 'H/L3.6', 'Iron, IIIAB-an', 'LL7',
'Mesosiderite-B2', 'LL4-6', 'CO3.7', 'L/LL6-an',
'Iron, IAB complex', 'Pallasite, PMG-an', 'H3.9/4', 'L3.8',
'LL5-6', 'LL3.8-6', 'L3.9', 'L4-5', 'L3-5', 'LL4/5', 'L4/5',
'H3.9', 'H3.6-6', 'H3.8-5', 'H3.8/4', 'H3.9-5', 'CH3', 'R3.8-5',
'L3.9/4', 'E4', 'CO3', 'Chondrite-ung', 'H~5', 'H~6', 'L/LL3.10',
'EL5', 'LL3', 'L~6', 'L~3', 'H~4', 'L(LL)3.5-3.7', 'Iron, IIIE-an',
'H3.6', 'L3.4-3.7', 'L3.5', 'CM1/2', 'Martian (OPX)', 'Brachinite',
'LL7(?)', 'LL6(?)', 'Eucrite-Mg rich', 'H3.5-4', 'EL3', 'R3.6',
'H3.5', 'CM1', 'L/LL3', 'H7', 'L(?)3', 'L3.2', 'L3.7-3.9',
'Mesosiderite-B1', 'Eucrite-unbr', 'LL3.7', 'CO3.0', 'LL3.5',
'L3.7-4', 'CV3-an', 'Lunar (anorth)', 'L3.3', 'Iron, IAB-sLM',
'Lunar', 'Iron, IC', 'Iron, IID-an', 'Iron, IIIE', 'Iron, IVA-an',
'CK6', 'L3.1', 'CK5', 'H3.3', 'H3.7-6', 'E6', 'H3.0', 'H3.1',
'L3.0', 'L/LL3.4', 'C6', 'LL3.0', 'Lunar (gabbro)', 'R4', 'C4',
'Iron, IIG', 'Iron, IIC', 'C1-ung', 'H5-an', 'EH4/5', 'Iron, IIIF',
'R3-6', 'Mesosiderite-B4', 'L6/7', 'Relict H', 'L-imp melt', 'CK3',
'H3-an', 'Iron, IVB', 'R3.8', 'L~5', 'Mesosiderite-an',
'Mesosiderite-A2', 'Pallasite, PES', 'C4-ung', 'Iron, IAB?',
'Mesosiderite-A', 'R3.5-6', 'H3.9-6', 'Ureilite-pmict', 'LL~6',
'CK4/5', 'EL4', 'Lunar (feldsp. breccia)', 'L3.9-6', 'H-an',
'L/LL3-6', 'L/LL3-5', 'H/L3.5', 'H/L3', 'R3-4', 'CK3-an', 'LL4-5',
'H/L6', 'L3/4', 'H-imp melt', 'CR', 'Chondrite-fusion crust',
'Iron, IAB-sLH', 'H(L)3-an', 'L(LL)3', 'H(L)3', 'R3', 'L7',
'CM-an', 'L/LL~6', 'L/LL~5', 'L~4', 'L/LL~4', 'LL(L)3', 'H3.2',
'L-melt breccia', 'H6-melt breccia', 'H5-melt breccia',
'H-melt rock', 'Eucrite-an', 'Lunar (bas/anor)', 'LL5/6', 'LL3/4',
'H3.4/3.5', 'Lunar (basalt)', 'H/L5', 'H(5?)', 'LL-imp melt',
'Mesosiderite?', 'H~4/5', 'L6-melt breccia', 'L3.5-3.7',
'Iron, IIAB-an', 'L3.3-3.7', 'L3.2-3.6', 'L3.3-3.6',
'Acapulcoite/Lodranite', 'Mesosiderite-B', 'CK5/6', 'L3.05', 'C2',
'C4/5', 'L/LL3.2', 'Iron, IIIAB?', 'L3.5-5', 'L/LL(?)3', 'H4(?)',
'Iron, IAB-sHH', 'Relict iron', 'EL4/5', 'L5-7', 'Diogenite-an',
'L-melt rock', 'CR1', 'H5', 'L5', 'H4', 'L4', 'E', 'L6',
'H3', 'LL6', 'H-metal', 'H6', 'L-metal', 'Relict OC', 'EH',
'Mesosiderite-A4', 'L/LL5/6', 'H3.8-4', 'CBb', 'EL6/7', 'EL7',
'CH/CBb', 'CO3.8', 'H/L~4', 'Mesosiderite-C2', 'R5', 'H4/6',
```

```
'H3.7-5', 'LL3.7-6', 'H3.7/3.8', 'L3.7/3.8', 'EH-imp melt', 'R',
'Fusion crust', 'Aubrite-an', 'R6', 'LL-melt rock', 'L3.5-3.9',
'L3.2-3.5', 'L3.3-3.5', 'L3.0-3.7', 'E3-an', 'K', 'E3',
'Acapulcoite/lodranite', 'CK4-an', 'L(LL)3.05', 'L3.10', 'CB',
'Diogenite-olivine', 'EL-melt rock', 'EH6', 'Pallasite, ungrouped',
'L/LL4/5', 'L3.8-an', 'Iron, IAB-an', 'C5/6-ung', 'CV2',
'Iron, IC-an', 'Lunar (bas. breccia)', 'L3.8-6', 'R3/4', 'R3.9',
'CK', 'LL3.10', 'R4/5', 'L3.8-5', 'Mesosiderite-C', 'Enst achon',
'H/L3-4', 'L(H)3', 'LL6/7', 'LL3.1', 'OC3', 'R3.7', 'CO3 ', 'CH3 ',
'LL~4', 'LL~4/5', 'L(LL)~4', 'H3.05', 'H3.10',
'Impact melt breccia', 'LL3-5', 'H/L3.7', 'LL3-4', 'CK3/4',
'Martian', 'CO3.1', 'Lunar (bas/gab brec)', 'Achondrite-prim',
'LL<3.5', 'CK3.8', 'L/LL-melt rock', 'H6/7', 'EL6 ',
'Iron, IAB-sHL-an', 'CM2-an', 'R3-5', 'L4-melt rock',
'L6-melt rock', 'H/L4/5', 'EL3/4', 'H/L6-melt rock',
'Enst achon-ung', 'L3-7', 'R3.4', 'LL3.05', 'LL4/6', 'LL3.8-4',
'H3.15', 'C3.0-ung', 'LL-melt breccia', 'LL6-melt breccia',
'L5-melt breccia', 'LL(L)3.1', 'LL6-an', 'L4-melt breccia',
'Howardite-an', 'H4-melt breccia', 'Martian (basaltic breccia)',
'L3-melt breccia', 'L~4-6', 'LL~5', 'R3.5-4', 'CR7',
'H-melt breccia', 'Lunar (norite)', 'L3.00', 'H3.0-3.4', 'L/LL4-6',
'CM', 'EH7', 'L4-an', 'E-an', 'H3.8/3.9', 'L3.9-5', 'H3.8-6',
'H3.4-5', 'L3.0-3.9', 'L3.5-3.8', 'H3.2-3.7', 'L3.6-4',
'Iron, IIE?', 'C3/4-ung', 'L/LL3.5', 'L/LL3.6/3.7', 'H/L4-5',
'LL~3', 'Pallasite?', 'LL5-7', 'LL3.9/4', 'H3.8-an', 'CR-an',
'L/LL5-6', 'L(LL)5', 'L(LL)6', 'LL3.1-3.5', 'E5', 'Lodranite-an',
'H3.2-6', 'H(?)4', 'E5-an', 'H3.2-an', 'EH6-an', 'Stone-ung',
'C1/2-ung', 'L/LL'], dtype=object)
```

In [86]: meteorites.describe(include = "all")

Out[86]:

	name	id	nametype	recclass	mass (g)	fall	year	
<b>count</b>	45716	45716.000000	45716	45716	4.558500e+04	45716	45425	3840
<b>unique</b>	45716	NaN	2	466	NaN	2	266	
<b>top</b>	Aachen	NaN	Valid	L6	NaN	Found	01/01/2003 12:00:00 AM	
<b>freq</b>	1	NaN	45641	8285	NaN	44609	3323	
<b>mean</b>	NaN	26889.735104	NaN	NaN	1.327808e+04	NaN	NaN	-39
<b>std</b>	NaN	16860.683030	NaN	NaN	5.749889e+05	NaN	NaN	46
<b>min</b>	NaN	1.000000	NaN	NaN	0.000000e+00	NaN	NaN	-87
<b>25%</b>	NaN	12688.750000	NaN	NaN	7.200000e+00	NaN	NaN	-76
<b>50%</b>	NaN	24261.500000	NaN	NaN	3.260000e+01	NaN	NaN	-77
<b>75%</b>	NaN	40656.750000	NaN	NaN	2.026000e+02	NaN	NaN	(
<b>max</b>	NaN	57458.000000	NaN	NaN	6.000000e+07	NaN	NaN	87

## Exercise (Part 1)

Using the 2019\_Yellow\_Taxi\_Trip\_Data.csv dataset, accomplish the following items and submit a PDF of the notebook:


1. Create a DataFrame by reading in the 2019\_Yellow\_Taxi\_Trip\_Data.csv file. Examine the first 5 rows.
2. Find the dimensions (number of rows and number of columns) in the data.
3. Using the data in the 2019\_Yellow\_Taxi\_Trip\_Data.csv file, calculate summary statistics for the fare\_amount, tip\_amount, tolls\_amount, and total\_amount columns.
4. Isolate the fare\_amount, tip\_amount, tolls\_amount, and total\_amount for the longest trip by distance (trip\_distance).

1. Create a DataFrame by reading in the 2019\_Yellow\_Taxi\_Trip\_Data.csv file. Examine the first 5 rows.

```
In [61]: data = pd.read_csv("2019_Yellow_Taxi_Trip_Data.csv")
dframe = pd.DataFrame(data)
dframe.head()
```

```
Out[61]:
```

	vendorid	tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	trip_distance
0	2	2019-10-23T16:39:42.000	2019-10-23T17:14:10.000	1	7.93
1	1	2019-10-23T16:32:08.000	2019-10-23T16:45:26.000	1	2.00
2	2	2019-10-23T16:08:44.000	2019-10-23T16:21:11.000	1	1.36
3	2	2019-10-23T16:22:44.000	2019-10-23T16:43:26.000	1	1.00
4	2	2019-10-23T16:45:11.000	2019-10-23T16:58:49.000	1	1.96



2. Find the dimensions (number of rows and number of columns) in the data.

```
In [62]: dframe.shape
```

```
Out[62]: (10000, 18)
```

The rows is 10000. The columns is 18.

3. Using the data in the 2019\_Yellow\_Taxi\_Trip\_Data.csv file, calculate summary statistics for the fare\_amount, tip\_amount, tolls\_amount, and total\_amount columns.

```
In [63]: result = dframe.iloc[:, [4, 10, 13, 14, 16]]
result.describe()
```

```
Out[63]:
```

	trip_distance	fare_amount	tip_amount	tolls_amount	total_amount
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	3.015250	15.106313	2.634494	0.623447	22.564659
std	4.148063	13.954762	3.409800	6.437507	19.209255
min	0.000000	-52.000000	0.000000	-6.120000	-65.920000
25%	0.920000	7.000000	0.000000	0.000000	12.375000
50%	1.500000	10.000000	2.000000	0.000000	16.300000
75%	2.760000	16.000000	3.250000	0.000000	22.880000
max	38.110000	176.000000	43.000000	612.000000	671.800000

4. Isolate the fare\_amount, tip\_amount, tolls\_amount, and total\_amount for the longest trip by distance (trip\_distance).

```
In [64]: result.loc[result["trip_distance"].idxmax()]
```

```
Out[64]: trip_distance    38.11
fare_amount    176.00
tip_amount     18.29
tolls_amount     6.12
total_amount    201.21
Name: 8338, dtype: float64
```

Reflection: In this activity, I had a hard time on the last part as I get confused on how would I isolate the fare\_amount, tip\_amount, tolls\_amount, and total\_amount for the longest trip by distance. All throughout the activity, I was able to apply all what I have understood on the lecture.

```
In [107... taxi = pd.read_csv("2019_Yellow_Taxi_Trip_Data.csv")
```

```
In [108... mask = taxi.columns.str.contains("id$|store_and_fwd_flag", regex = True)
columns_to_drop = taxi.columns[mask]
columns_to_drop
```

```
Out[108... Index(['vendorid', 'ratecodeid', 'store_and_fwd_flag', 'pulocationid',
      'dolocationid'],
      dtype='object')
```

```
In [109... taxi = taxi.drop(columns = columns_to_drop)
taxi.head()
```

Out[109...

	tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	trip_distance	payment_t
0	2019-10-23T16:39:42.000	2019-10-23T17:14:10.000	1	7.93	
1	2019-10-23T16:32:08.000	2019-10-23T16:45:26.000	1	2.00	
2	2019-10-23T16:08:44.000	2019-10-23T16:21:11.000	1	1.36	
3	2019-10-23T16:22:44.000	2019-10-23T16:43:26.000	1	1.00	
4	2019-10-23T16:45:11.000	2019-10-23T16:58:49.000	1	1.96	

In [110...

```
taxis = taxis.rename(
    columns={
        'tpep_pickup_datetime': 'pickup',
        'tpep_dropoff_datetime': 'dropoff'
    }
)
taxis.columns
```

Out[110...

```
Index(['pickup', 'dropoff', 'passenger_count', 'trip_distance', 'payment_type',
       'fare_amount', 'extra', 'mta_tax', 'tip_amount', 'tolls_amount',
       'improvement_surcharge', 'total_amount', 'congestion_surcharge'],
      dtype='object')
```

In [111...

```
taxis[['pickup', 'dropoff']] = taxis[['pickup', 'dropoff']].apply(pd.to_datetime)
taxis.dtypes
```

Out[111...

```
pickup          datetime64[ns]
dropoff         datetime64[ns]
passenger_count      int64
trip_distance      float64
payment_type        int64
fare_amount        float64
extra              float64
mta_tax            float64
tip_amount         float64
tolls_amount       float64
improvement_surcharge float64
total_amount       float64
congestion_surcharge float64
dtype: object
```

In [112...

```
taxis = taxis.assign(
    elapsed_time = lambda x: x.dropoff - x.pickup,
    cost_before_tip = lambda x: x.total_amount - x.tip_amount,
    tip_pct = lambda x: x.tip_amount / x.cost_before_tip,
    fees = lambda x: x.cost_before_tip - x.fare_amount,
    avg_speed = lambda x: x.trip_distance.div(
        x.elapsed_time.dt.total_seconds()/60/60
```



```
)
)
#It create columns with computed values

taxi.dtypes
```

```
Out[112...] pickup          datetime64[ns]
dropoff          datetime64[ns]
passenger_count      int64
trip_distance        float64
payment_type         int64
fare_amount          float64
extra                float64
mta_tax              float64
tip_amount           float64
tolls_amount         float64
improvement_surcharge float64
total_amount         float64
congestion_surcharge float64
elapsed_time         timedelta64[ns]
cost_before_tip       float64
tip_pct              float64
fees                 float64
avg_speed            float64
dtype: object
```

```
In [113...] taxi.head(2)
```

```
Out[113...]   pickup  dropoff  passenger_count  trip_distance  payment_type  fare_amount  extra  m
```

<b>0</b>	2019-10-23 16:39:42	2019-10-23 17:14:10	1	7.93	1	29.5	1.0	
<b>1</b>	2019-10-23 16:32:08	2019-10-23 16:45:26	1	2.00	1	10.5	1.0	



```
In [114...] taxi.sort_values(["passenger_count", "pickup"], ascending = [False, True]).head()
```

Out[114...

	pickup	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra
<b>5997</b>	2019-10-23 15:55:19	2019-10-23 16:08:25	6	1.58	2	10.0	1.0
<b>443</b>	2019-10-23 15:56:59	2019-10-23 16:04:33	6	1.46	2	7.5	1.0
<b>8722</b>	2019-10-23 15:57:33	2019-10-23 16:03:34	6	0.62	1	5.5	1.0
<b>4198</b>	2019-10-23 15:57:38	2019-10-23 16:05:07	6	1.18	1	7.0	1.0
<b>8238</b>	2019-10-23 15:58:31	2019-10-23 16:29:29	6	3.23	2	19.5	1.0



In [115...

```
taxis.nlargest(3, "elapsed_time")
#It shows the largest value in elapsed_time
```

Out[115...

	pickup	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra
<b>7576</b>	2019-10-23 16:52:51	2019-10-24 16:51:44	1	3.75	1	17.5	1.0
<b>6902</b>	2019-10-23 16:51:42	2019-10-24 16:50:22	1	11.19	2	39.5	1.0
<b>4975</b>	2019-10-23 16:18:51	2019-10-24 16:17:30	1	0.70	2	7.0	1.0



In [84]:

```
taxis.nlargest(3, "fare_amount")
```

Out[84]:

	pickup	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra
<b>8338</b>	2019-10-23 16:50:53	2019-10-24 15:32:55	1	38.11	1	176.00	0.0
<b>853</b>	2019-10-23 16:07:39	2019-10-23 17:37:05	3	19.09	2	160.00	0.0
<b>4714</b>	2019-10-23 16:33:17	2019-10-23 17:56:49	2	26.30	1	111.75	0.0



### Exercise (Part 2)

Read in the meteorite data from the Meteorite\_Landings.csv file, rename the mass (g) column to mass, and drop all the latitude and longitude columns. Sort the result by mass in descending order.

```
In [101... # Import meteorite data from the Meteorite_Landings.csv file
meteorite1 = pd.read_csv("Meteorite_Landings.csv")
meteorite1
```

Out[101...

	name	id	nametype	recclass	mass (g)	fall	year	reclat
0	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	50.77500
1	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	56.18333
2	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	54.21667
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	16.88333
4	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	-33.16667
...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	01/01/1990 12:00:00 AM	29.03700
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	01/01/1999 12:00:00 AM	13.78333
45713	Zlin	30410	Valid	H4	3.3	Found	01/01/1939 12:00:00 AM	49.25000
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	01/01/2003 12:00:00 AM	49.78917
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	01/01/1976 12:00:00 AM	33.98333

45716 rows × 10 columns



In [102...

```
# Rename the mass (g) column to mass
meteorite1 = meteorite1.rename(
    columns={
        'mass (g)': 'mass',
    }
)
meteorite1
```

Out[102...

	name	id	nametype	recclass	mass	fall	year	reclat
0	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	50.77500
1	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	56.18333
2	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	54.21667
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	16.88333
4	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	-33.16667
...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	01/01/1990 12:00:00 AM	29.03700
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	01/01/1999 12:00:00 AM	13.78333
45713	Zlin	30410	Valid	H4	3.3	Found	01/01/1939 12:00:00 AM	49.25000
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	01/01/2003 12:00:00 AM	49.78917
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	01/01/1976 12:00:00 AM	33.98333

45716 rows × 10 columns



In [103...

```
# Drop all the Latitude and Longitude columns
meteorite1 = meteorite1.drop(columns = ["reclat", "reclong"])
meteorite1
```

Out[103...

	name	id	nametype	recclass	mass	fall	year	GeoLocation
0	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	(50.775 6.08333
1	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	(56.18333 10.23333
2	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	(54.21667 -113.0
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	(16.88333 -99.9
4	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	(-33.16667 -64.95
...	...	...	...	...	...	...	...	.
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	01/01/1990 12:00:00 AM	(29.037 17.0185
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	01/01/1999 12:00:00 AM	(13.78333 8.96667
45713	Zlin	30410	Valid	H4	3.3	Found	01/01/1939 12:00:00 AM	(49.25 17.66667
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	01/01/2003 12:00:00 AM	(49.78917 41.5046
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	01/01/1976 12:00:00 AM	(33.98333 -115.68333

45716 rows × 8 columns



In [104...

```
meteorite1 = meteorite1.sort_values("mass", ascending = False)
meteorite1
```

Out[104...

	name	id	nametype	recclass	mass	fall	year	GeoLocation
<b>16392</b>	Hoba	11890	Valid	Iron, IVB	60000000.0	Found	01/01/1920 12:00:00 AM	(-19.5833 17.9166
<b>5373</b>	Cape York	5262	Valid	Iron, IIIAB	58200000.0	Found	01/01/1818 12:00:00 AM	(76.1333 -64.9333
<b>5365</b>	Campo del Cielo	5247	Valid	Iron, IAB- MG	50000000.0	Found	12/22/1575 12:00:00 AM	(-27.4666 -60.5833
<b>5370</b>	Canyon Diablo	5257	Valid	Iron, IAB- MG	30000000.0	Found	01/01/1891 12:00:00 AM	(35.0 -111.0333
<b>3455</b>	Armanty	2335	Valid	Iron, IIIE	28000000.0	Found	01/01/1898 12:00:00 AM	(47.0, 88
...	...	...	...	...	...	...	...	...
<b>38282</b>	Wei- hui-fu (a)	24231	Valid	Iron	NaN	Found	01/01/1931 12:00:00 AM	Na
<b>38283</b>	Wei- hui-fu (b)	24232	Valid	Iron	NaN	Found	01/01/1931 12:00:00 AM	Na
<b>38285</b>	Weiyuan	24233	Valid	Mesosiderite	NaN	Found	01/01/1978 12:00:00 AM	(35.2666 104.3166
<b>41472</b>	Yamato 792768	28117	Valid	CM2	NaN	Found	01/01/1979 12:00:00 AM	(-71 35.6666
<b>45698</b>	Zapata County	30393	Valid	Iron	NaN	Found	01/01/1930 12:00:00 AM	(27.0, -99

45716 rows × 8 columns




In [116...

```
taxis = taxis.set_index("pickup")
taxis.head(3)
```

Out[116...

	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	mta
pickup							
<b>2019-10-23 16:39:42</b>	2019-10-23 17:14:10	1	7.93	1	29.5	1.0	
<b>2019-10-23 16:32:08</b>	2019-10-23 16:45:26	1	2.00	1	10.5	1.0	
<b>2019-10-23 16:08:44</b>	2019-10-23 16:21:11	1	1.36	1	9.5	1.0	



In [129...

```
taxis.sort_index(axis = 0)  
# This sort the by index, but axis = 1 is sorted by columns
```



Out[129...

	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	mta
pickup							
<b>2019-10-23 07:05:34</b>	2019-10-23 08:03:16	3	14.68	1	50.0	1.0	
<b>2019-10-23 07:48:58</b>	2019-10-23 07:52:09	1	0.67	2	4.5	1.0	
<b>2019-10-23 08:02:09</b>	2019-10-24 07:42:32	1	8.38	1	32.0	1.0	
<b>2019-10-23 08:18:47</b>	2019-10-23 08:36:05	1	2.39	2	12.5	1.0	
<b>2019-10-23 09:27:16</b>	2019-10-23 09:33:13	2	1.11	2	6.0	1.0	
...	...	...	...	...	...	...	...
<b>2019-10-24 07:23:52</b>	2019-10-24 08:08:52	1	0.00	1	36.2	0.0	
<b>2019-10-24 07:29:52</b>	2019-10-24 07:33:24	1	0.54	2	4.0	0.0	
<b>2019-10-24 07:58:31</b>	2019-10-24 08:47:05	1	0.00	1	22.2	0.0	
<b>2019-10-24 08:07:45</b>	2019-10-24 08:07:50	2	0.00	2	52.0	0.0	
<b>2019-10-24 08:19:11</b>	2019-10-24 09:00:35	0	13.20	2	42.0	0.0	

10000 rows × 17 columns



In [139...

```
taxis["2019-10-23 07:45": "2019-10-23 08"]
```

Out[139...

	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	mta
pickup							
<b>2019-10-23 07:48:58</b>	2019-10-23 07:52:09	1	0.67	2	4.5	1.0	
<b>2019-10-23 08:02:09</b>	2019-10-24 07:42:32	1	8.38	1	32.0	1.0	
<b>2019-10-23 08:18:47</b>	2019-10-23 08:36:05	1	2.39	2	12.5	1.0	

In [140...

```
taxis["2019-10-23": "2019-10-23"]
```

Out[140...

	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	mta
pickup							
<b>2019-10-23 07:05:34</b>	2019-10-23 08:03:16	3	14.68	1	50.0	1.0	
<b>2019-10-23 07:48:58</b>	2019-10-23 07:52:09	1	0.67	2	4.5	1.0	
<b>2019-10-23 08:02:09</b>	2019-10-24 07:42:32	1	8.38	1	32.0	1.0	
<b>2019-10-23 08:18:47</b>	2019-10-23 08:36:05	1	2.39	2	12.5	1.0	
<b>2019-10-23 09:27:16</b>	2019-10-23 09:33:13	2	1.11	2	6.0	1.0	
...	...	...	...	...	...	...	...
<b>2019-10-23 17:59:53</b>	2019-10-23 18:12:56	2	1.60	1	10.0	1.0	
<b>2019-10-23 17:59:53</b>	2019-10-23 18:19:12	1	2.39	2	14.0	1.0	
<b>2019-10-23 18:00:03</b>	2019-10-23 18:04:56	1	0.94	2	5.5	1.0	
<b>2019-10-23 18:01:21</b>	2019-10-23 18:08:00	5	1.25	1	6.5	1.0	
<b>2019-10-23 18:03:03</b>	2019-10-23 18:10:45	1	0.76	1	6.5	1.0	

9993 rows × 17 columns



In [148...

```
taxis.loc["2019-10-23 08"]
```

Out[148...

	pickup	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	mta
<b>2019-10-23 08:02:09</b>	2019-10-24 07:42:32		1	8.38	1	32.0	1.0	
<b>2019-10-23 08:18:47</b>	2019-10-23 08:36:05		1	2.39	2	12.5	1.0	

In [149...

```
taxis = taxis.reset_index()
taxis.head()
# This the index
```

Out[149...

	pickup	dropoff	passenger_count	trip_distance	payment_type	fare_amount	extra	m
<b>0</b>	2019-10-23 07:05:34	2019-10-23 08:03:16	3	14.68	1	50.0	1.0	
<b>1</b>	2019-10-23 07:48:58	2019-10-23 07:52:09	1	0.67	2	4.5	1.0	
<b>2</b>	2019-10-23 08:02:09	2019-10-24 07:42:32	1	8.38	1	32.0	1.0	
<b>3</b>	2019-10-23 08:18:47	2019-10-23 08:36:05	1	2.39	2	12.5	1.0	
<b>4</b>	2019-10-23 09:27:16	2019-10-23 09:33:13	2	1.11	2	6.0	1.0	

### Exercise (Part 3)

In [235...

```
meteorite2 = pd.read_csv("Meteorite_Landings.csv")
meteorite2
```

Out[235...

	name	id	nametype	recclass	mass (g)	fall	year	reclat
0	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	50.77500
1	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	56.18333
2	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	54.21667
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	16.88333
4	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	-33.16667
...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	01/01/1990 12:00:00 AM	29.03700
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	01/01/1999 12:00:00 AM	13.78333
45713	Zlin	30410	Valid	H4	3.3	Found	01/01/1939 12:00:00 AM	49.25000
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	01/01/2003 12:00:00 AM	49.78917
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	01/01/1976 12:00:00 AM	33.98333

45716 rows × 10 columns



In [236...

```
# Update the year column to only contain the year# Import meteorite data from the M
meteorite2_1 = lambda x: x.str[6:10]
meteorite2["year"] = meteorite2_1(meteorite2["year"])
meteorite2
```

Out[236...

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
0	Aachen	1	Valid	L5	21.0	Fell	1880	50.77500	6.08
1	Aarhus	2	Valid	H6	720.0	Fell	1951	56.18333	10.23
2	Abee	6	Valid	EH4	107000.0	Fell	1952	54.21667	-113.00
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	1976	16.88333	-99.90
4	Achiras	370	Valid	L6	780.0	Fell	1902	-33.16667	-64.95
...	...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	1990	29.03700	17.01
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	1999	13.78333	8.96
45713	Zlin	30410	Valid	H4	3.3	Found	1939	49.25000	17.66
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	2003	49.78917	41.50
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	1976	33.98333	-115.66

45716 rows × 10 columns



In [238...

```
# Convert it to a numeric data type
meteorite2["year"] = meteorite2["year"].apply(pd.to_numeric)
meteorite2.dtypes
```

Out[238...

```
name          object
id            int64
nametype      object
recclass      object
mass (g)      float64
fall          object
year          float64
reclat        float64
reclong       float64
GeoLocation   object
dtype: object
```

In [242...

```
# Create a new column indicating whether the meteorite was observed falling before
meteorite2["Observed_Falling"] = (meteorite2.year < 1970.0) & (meteorite2.fall == "
meteorite2
```

Out[242...

	name	id	nametype	recclass	mass (g)	fall	year	reclat	re
0	Aachen	1	Valid	L5	21.0	Fell	1880.0	50.77500	6.
1	Aarhus	2	Valid	H6	720.0	Fell	1951.0	56.18333	10.
2	Abee	6	Valid	EH4	107000.0	Fell	1952.0	54.21667	-113.
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	1976.0	16.88333	-99.
4	Achiras	370	Valid	L6	780.0	Fell	1902.0	-33.16667	-64.
...	...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	1990.0	29.03700	17.
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	1999.0	13.78333	8.
45713	Zlin	30410	Valid	H4	3.3	Found	1939.0	49.25000	17.
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	2003.0	49.78917	41.
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	1976.0	33.98333	-115.

45716 rows × 11 columns



In [245...

```
# Set the index to the id column and extract all the rows with IDs between 10036 and 10040
meteorite2.set_index("id")
```

Out[245...

	name	nametype	recclass	mass (g)	fall	year	reclat	reclong
id								
1	Aachen	Valid	L5	21.0	Fell	1880.0	50.77500	6.08333
2	Aarhus	Valid	H6	720.0	Fell	1951.0	56.18333	10.23333
6	Abee	Valid	EH4	107000.0	Fell	1952.0	54.21667	-113.00000
10	Acapulco	Valid	Acapulcoite	1914.0	Fell	1976.0	16.88333	-99.90000
370	Achiras	Valid	L6	780.0	Fell	1902.0	-33.16667	-64.95000
...	...	...	...	...	...	...	...	...
31356	Zillah 002	Valid	Eucrite	172.0	Found	1990.0	29.03700	17.01850
30409	Zinder	Valid	Pallasite, ungrouped	46.0	Found	1999.0	13.78333	8.96667
30410	Zlin	Valid	H4	3.3	Found	1939.0	49.25000	17.66667
31357	Zubkovsky	Valid	L6	2167.0	Found	2003.0	49.78917	41.50460
30414	Zulu Queen	Valid	L3.7	200.0	Found	1976.0	33.98333	-115.68333

45716 rows × 10 columns



In [246...

```
meteorite2.sort_index()
```



Out[246...

	name	id	nametype	recclass	mass (g)	fall	year	reclat	re
0	Aachen	1	Valid	L5	21.0	Fell	1880.0	50.77500	6.
1	Aarhus	2	Valid	H6	720.0	Fell	1951.0	56.18333	10.
2	Abee	6	Valid	EH4	107000.0	Fell	1952.0	54.21667	-113.
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	1976.0	16.88333	-99.
4	Achiras	370	Valid	L6	780.0	Fell	1902.0	-33.16667	-64.
...	...	...	...	...	...	...	...	...	...
45711	Zillah 002	31356	Valid	Eucrite	172.0	Found	1990.0	29.03700	17.
45712	Zinder	30409	Valid	Pallasite, ungrouped	46.0	Found	1999.0	13.78333	8.
45713	Zlin	30410	Valid	H4	3.3	Found	1939.0	49.25000	17.
45714	Zubkovsky	31357	Valid	L6	2167.0	Found	2003.0	49.78917	41.
45715	Zulu Queen	30414	Valid	L3.7	200.0	Found	1976.0	33.98333	-115.

45716 rows × 11 columns



In [247...

```
meteorite2.iloc[10036: 10040]
```

Out[247...

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong	Ge
10036	Elephant Moraine 90022	8432	Valid	CK5	15.5	Found	1990.0	-76.28573	156.45721	(
10037	Elephant Moraine 90023	8433	Valid	CK5	31.5	Found	1990.0	-76.27507	156.41038	(
10038	Elephant Moraine 90024	8434	Valid	Eucrite- br	22.8	Found	1990.0	-76.28843	156.47872	(
10039	Elephant Moraine 90025	8435	Valid	CK5	45.8	Found	1990.0	-76.28200	156.39926	1

#### Exercise (Part 4)

In [3]: `import pandas as pd`

In [5]: `data = pd.read_csv("Meteorite_Landings.csv")  
meteorites = pd.DataFrame(data)  
meteorites.head()`

Out[5]:

	name	id	nametype	recclass	mass (g)	fall	year	reclat	reclong
0	Aachen	1	Valid	L5	21.0	Fell	01/01/1880 12:00:00 AM	50.77500	6.08333
1	Aarhus	2	Valid	H6	720.0	Fell	01/01/1951 12:00:00 AM	56.18333	10.23333
2	Abee	6	Valid	EH4	107000.0	Fell	01/01/1952 12:00:00 AM	54.21667	-113.00000
3	Acapulco	10	Valid	Acapulcoite	1914.0	Fell	01/01/1976 12:00:00 AM	16.88333	-99.90000
4	Achiras	370	Valid	L6	780.0	Fell	01/01/1902 12:00:00 AM	-33.16667	-64.95000

1. 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 [7]: filter = lambda x: x.str[6:10]
meteorites["year"] = filter(meteorites["year"])
meteorites["year"] = meteorites["year"].apply(pd.to_numeric)
meteorites.dtypes
```

```
Out[7]: name          object
id            int64
nametype      object
recclass      object
mass (g)      float64
fall          object
year          float64
reclat        float64
reclong       float64
GeoLocation   object
dtype: object
```

```
In [9]: meteorites_filtered = meteorites[(meteorites["year"] >= 2005) & (meteorites["year"]
meteorites_filtered = meteorites_filtered.set_index("year")
```

```
In [11]: meteorites_Fell = meteorites_filtered[(meteorites_filtered["fall"] == "Fell")]
meteorites_Fell.iloc[:, 4].quantile([0.95])
```

```
Out[11]: 0.95    100000.0
Name: mass (g), dtype: float64
```

```
In [13]: meteorites_Found = meteorites_filtered[(meteorites_filtered["fall"] == "Found")]
meteorites_Found.iloc[:, 4].quantile([0.95])
```

```
Out[13]: 0.95    1841.64
Name: mass (g), dtype: float64
```

```
In [15]: Number_1_1 = meteorites_filtered.groupby(["year", "fall"])["mass (g)"].quantile(0)
Number_1_1
```

Out[15]:

mass (g)		
year	fall	
2005.0	Found	4500.00
2006.0	Fell	25008.00
	Found	1600.50
2007.0	Fell	89675.00
	Found	1126.90
2008.0	Fell	106000.00
	Found	2274.80
2009.0	Fell	8333.40
	Found	1397.25

```
In [17]: Number_1_2 = meteorites_filtered.groupby(["year", "fall"])[["name"]].count()  
Number_1_2
```

Out[17]:

name		
year	fall	
2005.0	Found	875
2006.0	Fell	5
	Found	2451
2007.0	Fell	8
	Found	1181
2008.0	Fell	9
	Found	948
2009.0	Fell	5
	Found	1492

```
In [19]: Number_1_1["95th Percentile"] = Number_1_1  
Number_1_1["Count"] = Number_1_2  
Number_1_1
```

Out[19]:

		mass (g)	95th Percentile	Count
year	fall			
2005.0	Found	4500.00	4500.00	875
2006.0	Fell	25008.00	25008.00	5
	Found	1600.50	1600.50	2451
2007.0	Fell	89675.00	89675.00	8
	Found	1126.90	1126.90	1181
2008.0	Fell	106000.00	106000.00	9
	Found	2274.80	2274.80	948
2009.0	Fell	8333.40	8333.40	5
	Found	1397.25	1397.25	1492

- 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 [21]: `Number_2 = meteorites_filtered.groupby("fall")["mass (g)"].describe()  
Number_2`

Out[21]:


	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)

In [25]: `data1 = pd.read_csv("2019_Yellow_Taxi_Trip_Data.csv")  
taxi = pd.DataFrame(data1)  
taxi.head()`

Out[25]:

	vendorid	tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	trip_distance
0	2	2019-10-23T16:39:42.000	2019-10-23T17:14:10.000	1	7.93
1	1	2019-10-23T16:32:08.000	2019-10-23T16:45:26.000	1	2.00
2	2	2019-10-23T16:08:44.000	2019-10-23T16:21:11.000	1	1.36
3	2	2019-10-23T16:22:44.000	2019-10-23T16:43:26.000	1	1.00
4	2	2019-10-23T16:45:11.000	2019-10-23T16:58:49.000	1	1.96



1. 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 [27]: # resample the data to an hourly frequency based on the dropoff time
taxi['tpep_dropoff_datetime'] = pd.to_datetime(taxi['tpep_dropoff_datetime'])
taxi.set_index("tpep_dropoff_datetime", inplace = True)
taxi.index = pd.to_datetime(taxi.index)
taxi['Hour'] = taxi.index.hour

# Calculate the total trip_distance, fare_amount, tolls_amount, and tip_amount, the
taxi = taxi.groupby('Hour')[['trip_distance', 'fare_amount', 'tolls_amount', 'tip_a
taxi = taxi["tip_amount"].nlargest(5)
taxi
```

```
Out[27]: Hour
16    12249.32
17    12044.03
18     1907.64
15       75.10
19       25.74
Name: tip_amount, dtype: float64
```

In [ ]: