

```
In [35]: import numpy as np
from datascience import *

# Configure notebook (happens automatically on data8.berkeley.edu)
%matplotlib inline
import matplotlib.pyplot as plt
plt.style.use('fivethirtyeight')

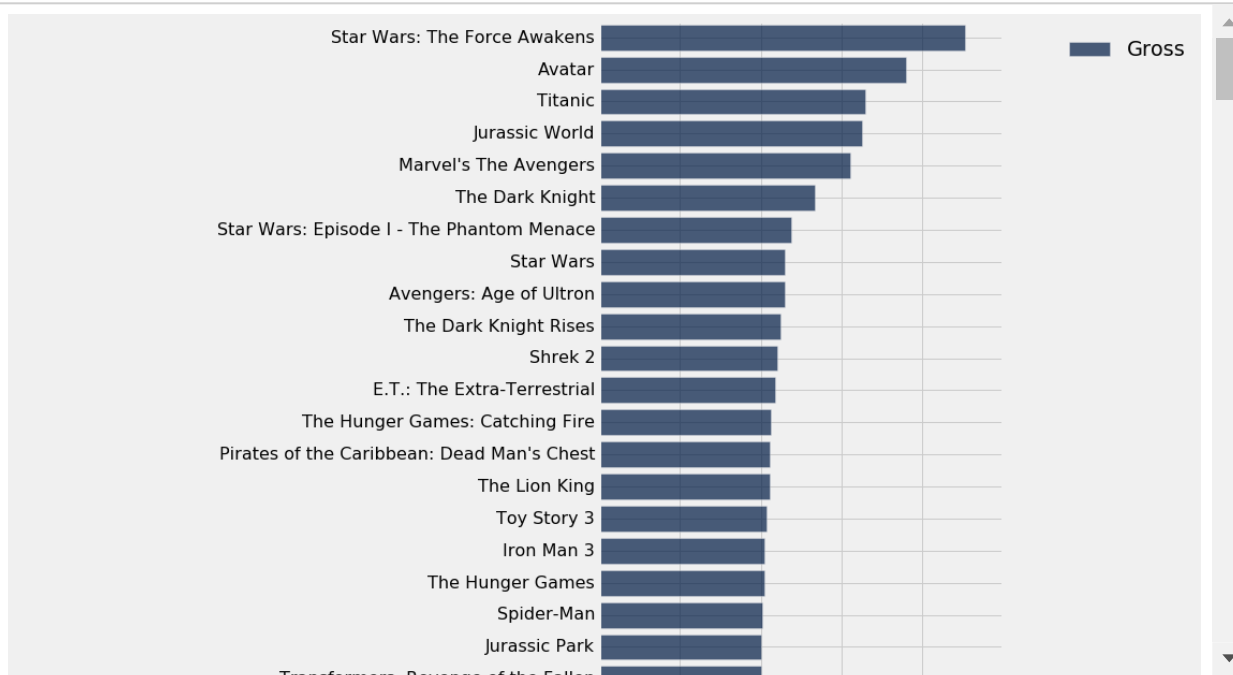
# Configure for presentation
np.set_printoptions(threshold=50, linewidth=50)
import matplotlib as mpl
mpl.rc('font', size=16)
```

```
In [125]: top = Table.read_table('top_movies.csv')
top.set_format([2, 3], NumberFormatter).show(30)
```

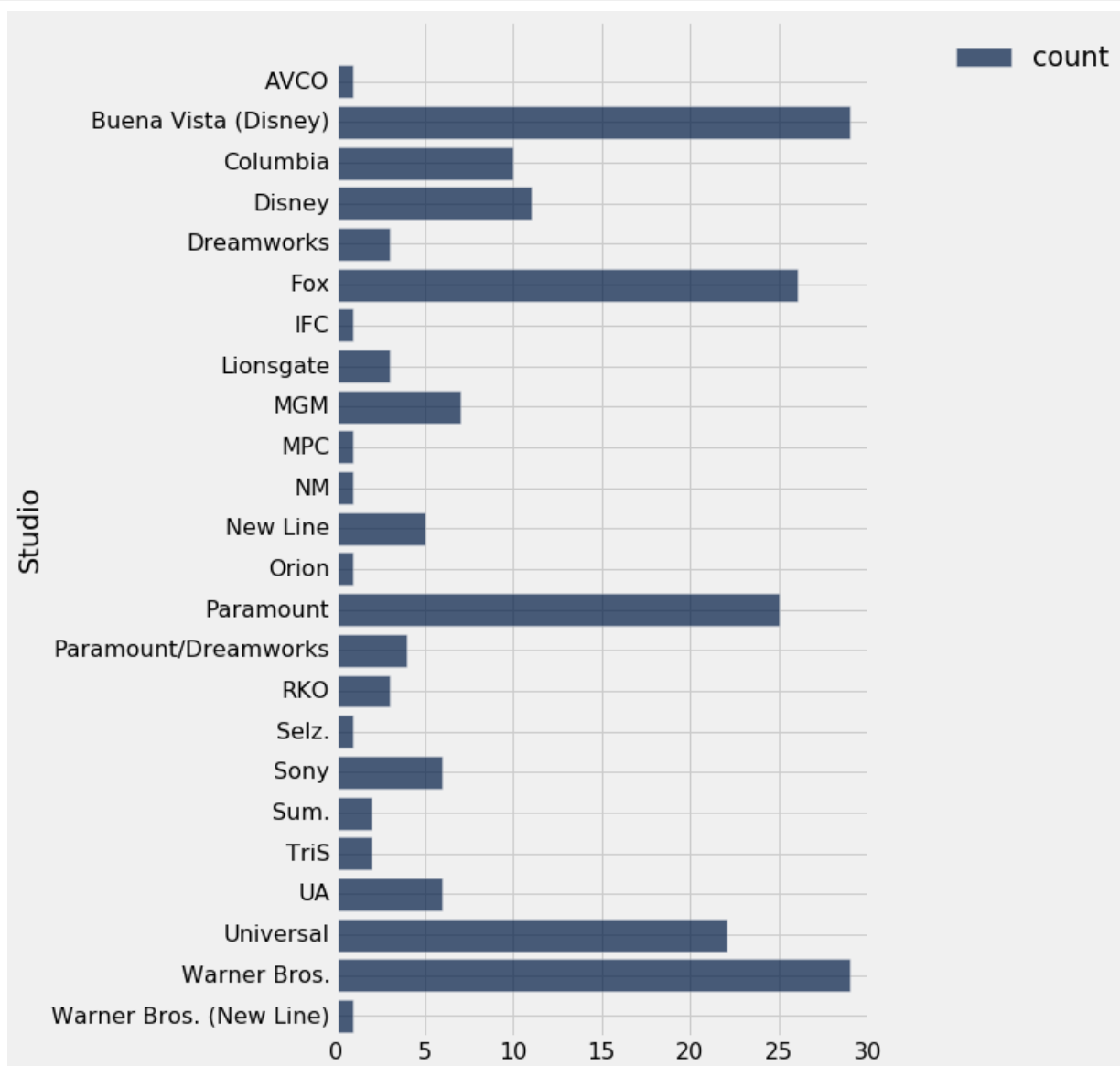
Title	Studio	Gross	Gross (Adjusted)	Year
Star Wars: The Force Awakens	Buena Vista (Disney)	906,723,418	906,723,400	2015
Avatar	Fox	760,507,625	846,120,800	2009
Titanic	Paramount	658,672,302	1,178,627,900	1997
Jurassic World	Universal	652,270,625	687,728,000	2015
Marvel's The Avengers	Buena Vista (Disney)	623,357,910	668,866,600	2012
The Dark Knight	Warner Bros.	534,858,444	647,761,600	2008
Star Wars: Episode I - The Phantom Menace	Fox	474,544,677	785,715,000	1999
Star Wars	Fox	460,998,007	1,549,640,500	1977
Avengers: Age of Ultron	Buena Vista (Disney)	459,005,868	465,684,200	2015
The Dark Knight Rises	Warner Bros.	448,139,099	500,961,700	2012
Shrek 2	Dreamworks	441,226,247	618,143,100	2004
E.T.: The Extra-Terrestrial	Universal	435,110,554	1,234,132,700	1982
The Hunger Games: Catching Fire	Lionsgate	424,668,047	444,697,400	2013
Pirates of the Caribbean: Dead Man's Chest	Buena Vista (Disney)	423,315,812	562,266,800	2006
The Lion King	Buena Vista (Disney)	422,783,777	775,573,900	1994
Toy Story 3	Buena Vista (Disney)	415,004,880	454,156,300	2010
Iron Man 3	Buena Vista (Disney)	409,013,994	424,632,700	2013
The Hunger Games	Lionsgate	408,010,692	442,510,400	2012
Spider-Man	Sony	403,706,375	604,517,300	2002
Jurassic Park	Universal	402,453,882	799,721,000	1993
Transformers: Revenge of the Fallen	Paramount/Dreamworks	402,111,870	468,938,100	2009
Frozen	Buena Vista (Disney)	400,738,009	426,656,900	2013
Harry Potter and the Deathly Hallows Part 2	Warner Bros.	381,011,219	417,512,200	2011
Finding Nemo	Buena Vista (Disney)	380,843,261	535,802,700	2003
Star Wars: Episode III - Revenge of the Sith	Fox	380,270,577	516,123,900	2005
The Lord of the Rings: The Return of the King	New Line	377,845,905	536,265,400	2003
Spider-Man 2	Sony	373,585,825	523,381,100	2004
The Passion of the Christ	NM	370,782,930	519,432,100	2004
Despicable Me 2	Universal	368,061,265	407,978,700	2013
Inside Out	Buena Vista (Disney)	356,461,711	375,723,400	2015

... (170 rows omitted)

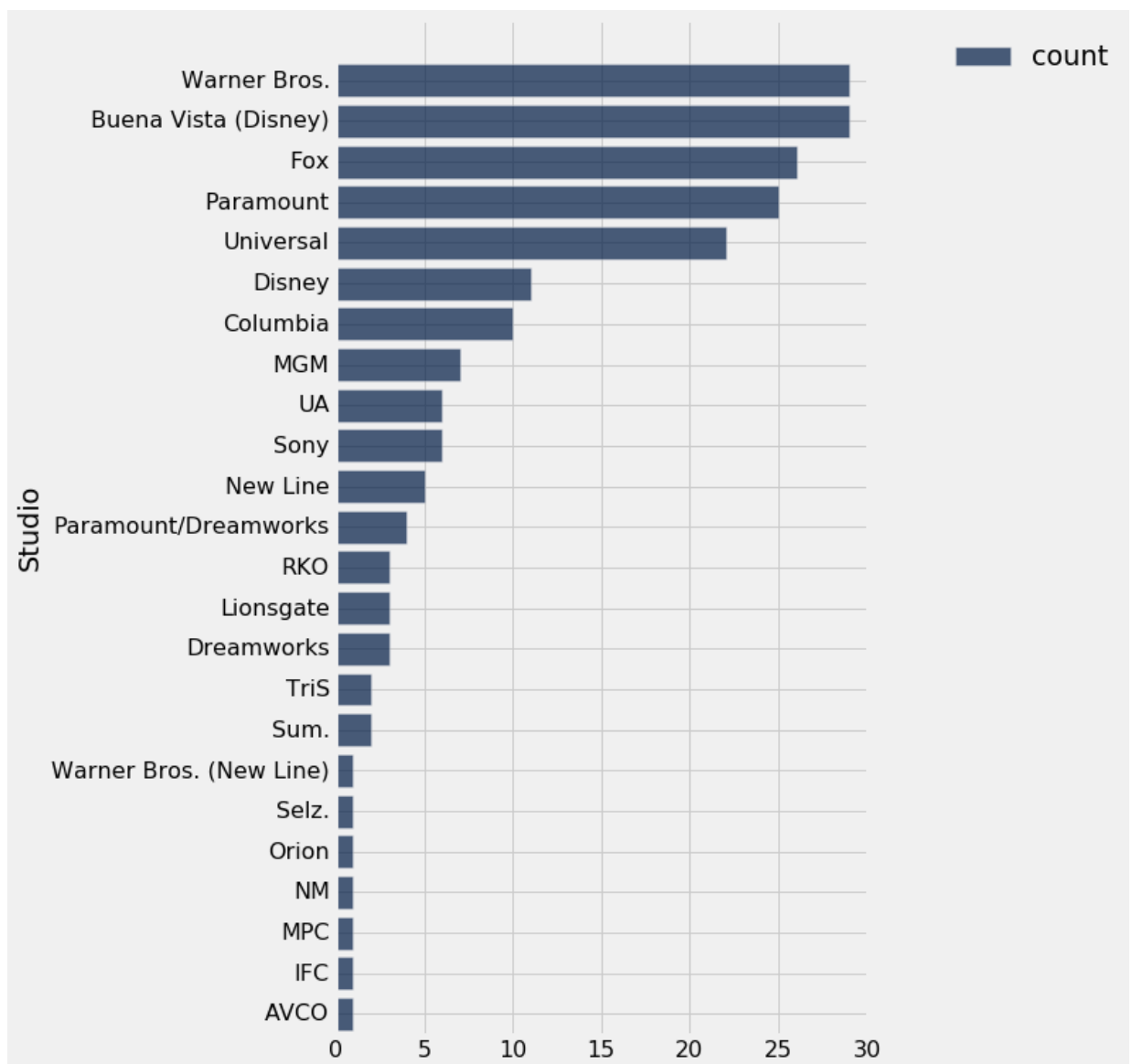
```
In [127]: top.barh('Title', 'Gross')
```



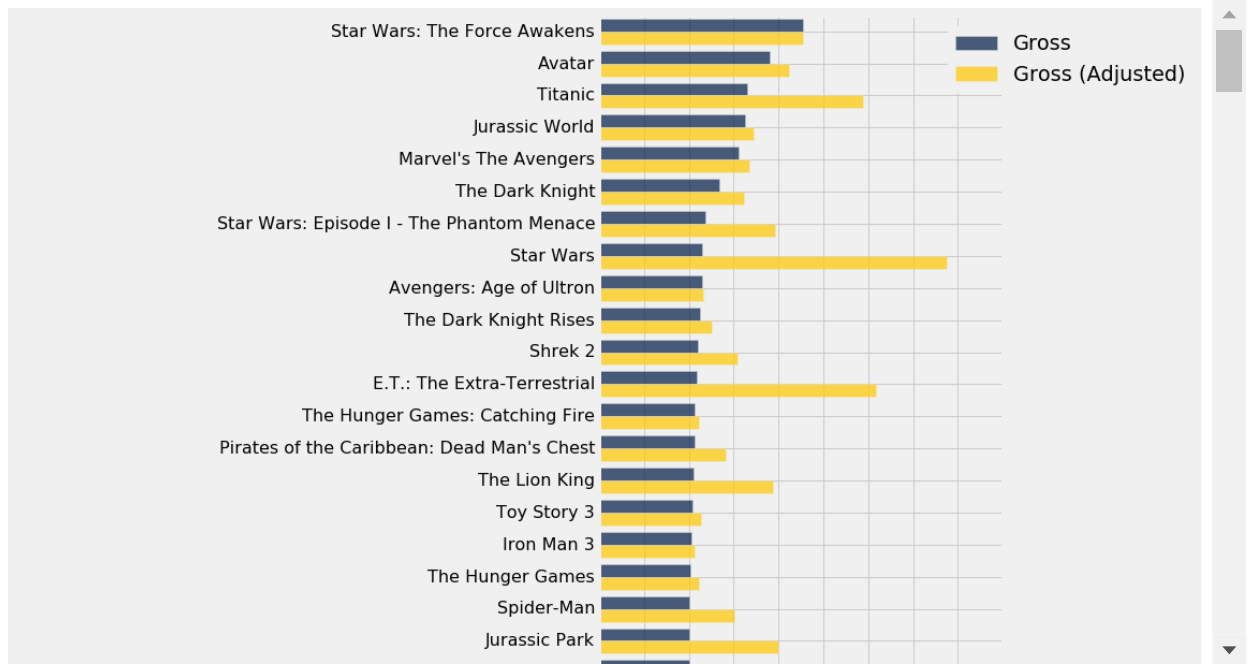
```
In [129]: top.group('Studio').barh('Studio', 'count')
```



```
In [130]: top.group('Studio').sort('count', descending=True).barh('Studio', 'count')
```



```
In [132]: top.select([0, 2, 3]).barh('Title')
```



## Histograms

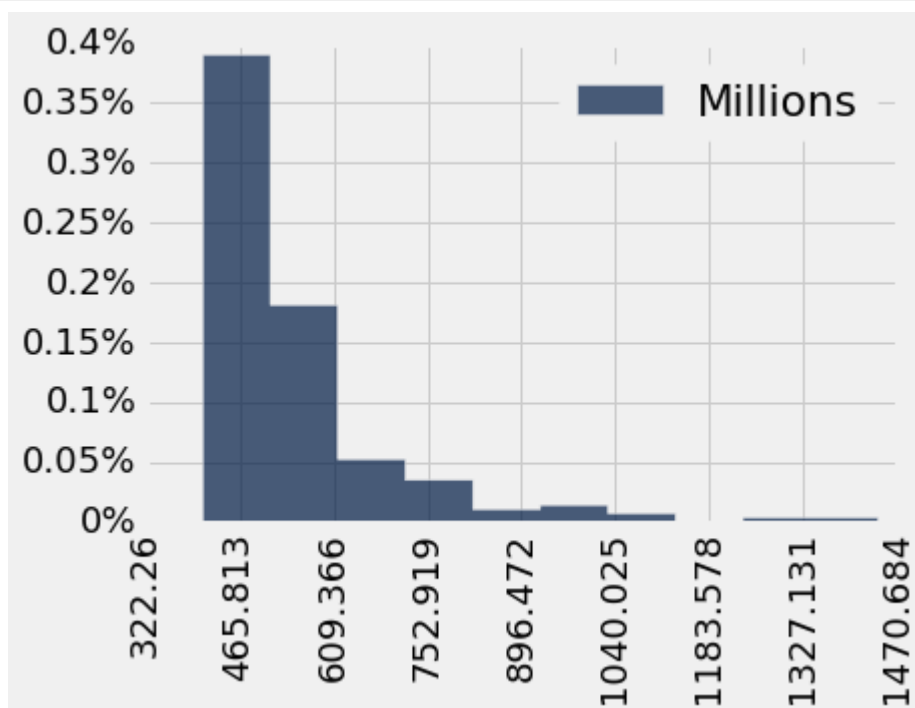
```
In [32]: in_millions = np.round(top.column(3)/1000000, 2)
millions = top.select(['Title', 'Year']).with_column('Millions', in_millions)
millions
```

Out[32]:

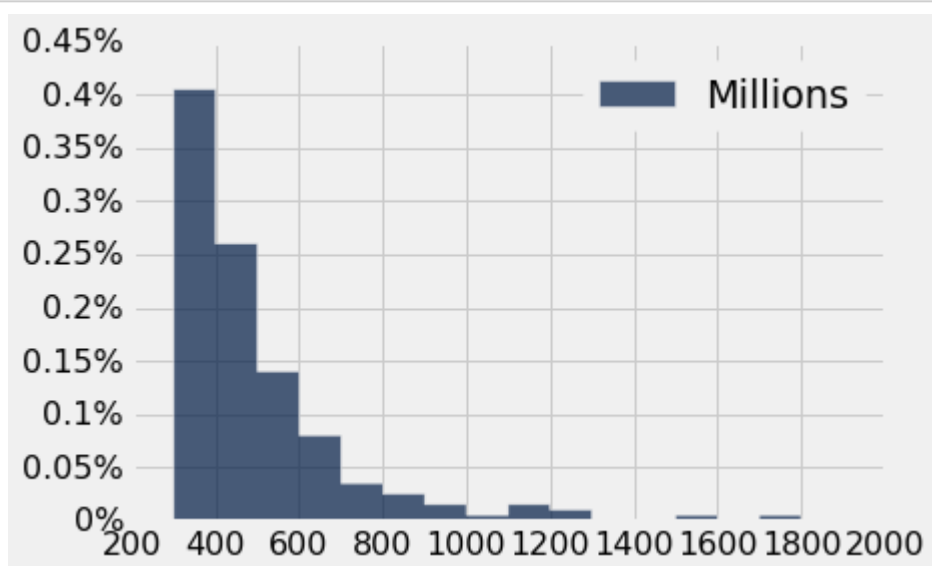
	Title	Year	Millions
	Star Wars: The Force Awakens	2015	906.72
	Avatar	2009	846.12
	Titanic	1997	1178.63
	Jurassic World	2015	687.73
	Marvel's The Avengers	2012	668.87
	The Dark Knight	2008	647.76
	Star Wars: Episode I - The Phantom Menace	1999	785.72
	Star Wars	1977	1549.64
	Avengers: Age of Ultron	2015	465.68
	The Dark Knight Rises	2012	500.96

... (190 rows omitted)

```
In [33]: millions.hist('Millions')
```



```
In [45]: millions.hist('Millions', bins=np.arange(300,2001,100))
```



```
In [55]: millions.bin('Millions', bins=np.arange(300,2001,100))
```

```
Out[55]:
```

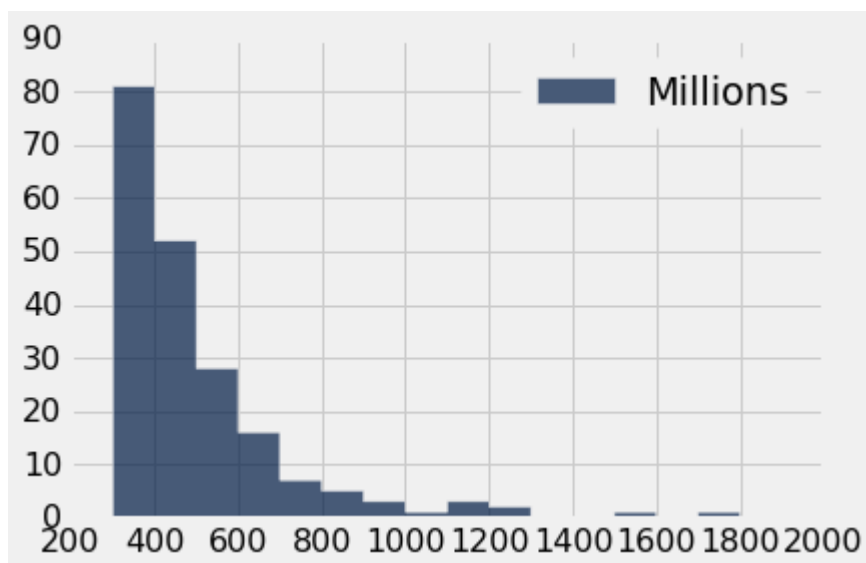
bin	Millions count
300	81
400	52
500	28
600	16
700	7
800	5
900	3
1000	1
1100	3
1200	2

... (8 rows omitted)

```
In [81]: PercentFormatter().format_value(81/millions.num_rows/(400-300))
```

```
Out[81]: '0.41%'
```

```
In [82]: millions.hist('Millions', bins=np.arange(300,2001,100), normed=False)
```



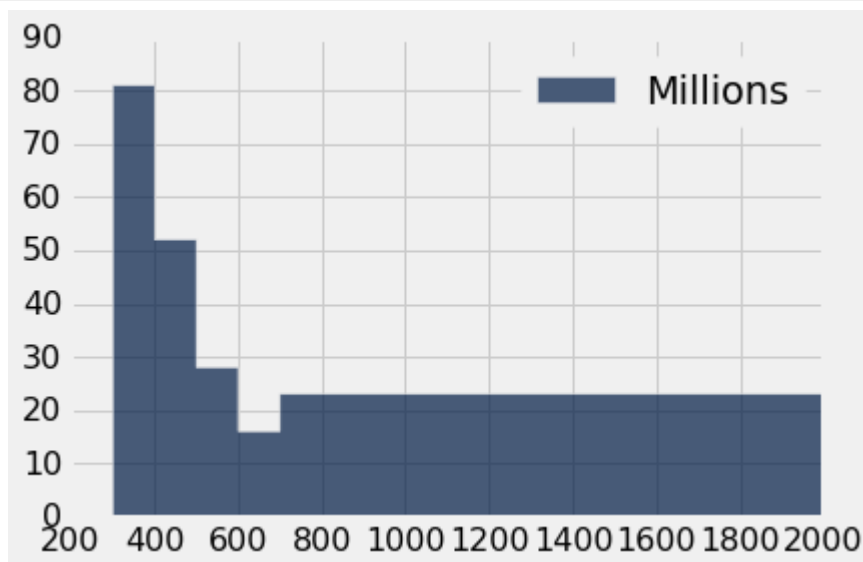


```
In [90]: millions.bin('Millions', bins=[300, 400, 500, 600, 700, 2000])
```

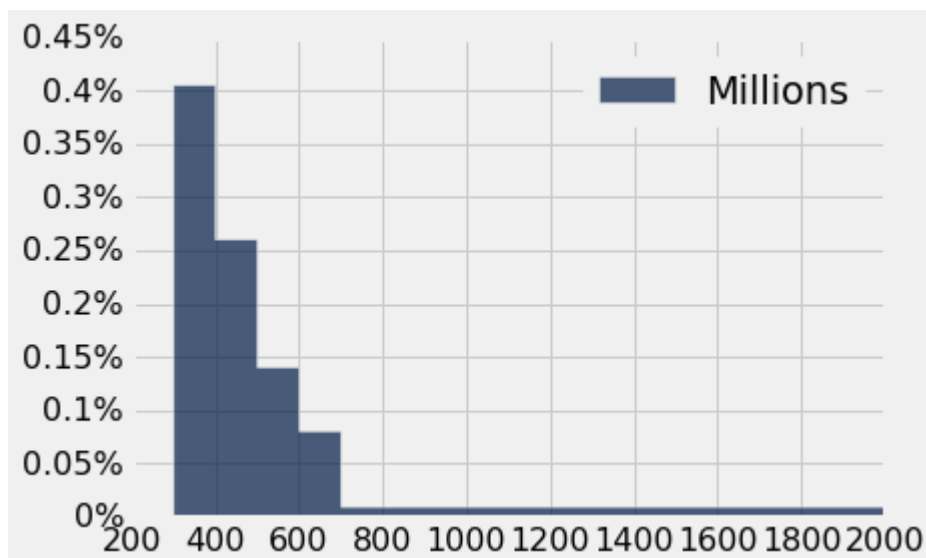
```
Out[90]:
```

bin	Millions count
300	81
400	52
500	28
600	16
700	23
2000	0

```
In [92]: millions.hist('Millions', bins=[300, 400, 500, 600, 700, 2000], normed=False)
```



```
In [91]: millions.hist('Millions', bins=[300, 400, 500, 600, 700, 2000])
```



```
In [104]: trips = Table.read_table('trip.csv')
trips
```

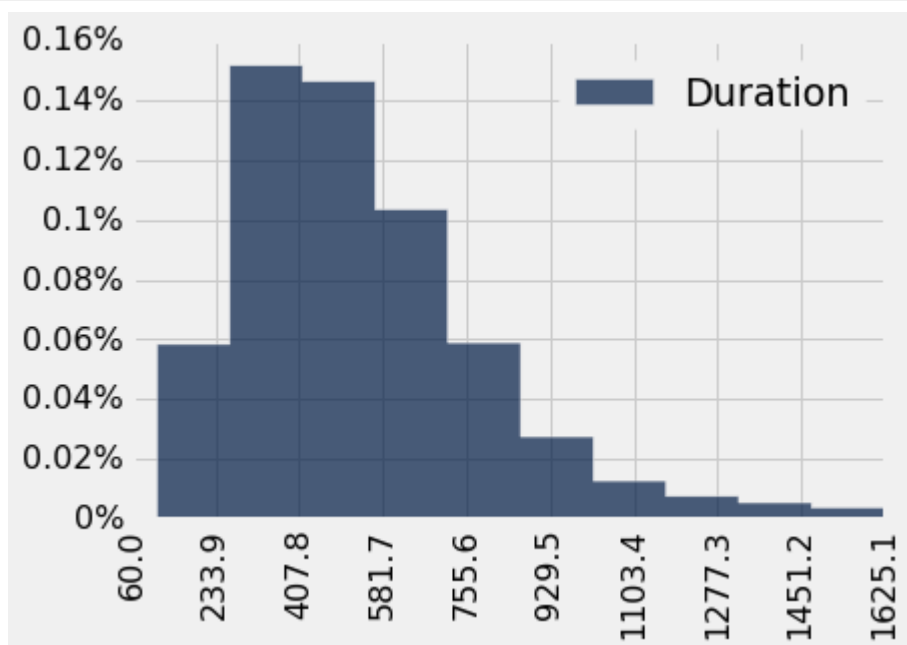
Out[104]:

Trip ID	Duration	Start Date	Start Station	Start Terminal	End Date	End Station	End Terminal	Bike #	Subsc
913460	765	8/31/2015 23:26	Harry Bridges Plaza (Ferry Building)	50	8/31/2015 23:39	San Francisco Caltrain (Townsend at 4th)	70	288	Subsc
913459	1036	8/31/2015 23:11	San Antonio Shopping Center	31	8/31/2015 23:28	Mountain View City Hall	27	35	Subsc
913455	307	8/31/2015 23:13	Post at Kearny	47	8/31/2015 23:18	2nd at South Park	64	468	Subsc
913454	409	8/31/2015 23:10	San Jose City Hall	10	8/31/2015 23:17	San Salvador at 1st	8	68	Subsc
913453	789	8/31/2015 23:09	Embarcadero at Folsom	51	8/31/2015 23:22	Embarcadero at Sansome	60	487	Cust
913452	293	8/31/2015 23:07	Yerba Buena Center of the Arts (3rd @ Howard)	68	8/31/2015 23:12	San Francisco Caltrain (Townsend at 4th)	70	538	Subsc
913451	896	8/31/2015 23:07	Embarcadero at Folsom	51	8/31/2015 23:22	Embarcadero at Sansome	60	363	Cust
913450	255	8/31/2015 22:16	Embarcadero at Sansome	60	8/31/2015 22:20	Steuart at Market	74	470	Subsc
913449	126	8/31/2015 22:12	Beale at Market	56	8/31/2015 22:15	Temporary Transbay Terminal (Howard at Beale)	55	439	Subsc
913448	932	8/31/2015 21:57	Post at Kearny	47	8/31/2015 22:12	South Van Ness at Market	66	472	Subsc

... (354142 rows omitted)



```
In [124]: commute = trips.where(trips.column('Duration') < 1800)  
commute.hist('Duration')
```



```
In [123]: commute.hist('Duration', bins=np.arange(60, 1801, 10))
```

