

```

Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning $ " ls
Learning_Python_5th_Edition_Mark_Lutz.pdf
Natural Language Processing with Python.pdf
Notebook_Python
OReilly.Python.for.Finance.2014.12.pdf
Oracle PL_SQL Language Pocket Reference, 5th Edition-O'Reilly(2015).pdf
Programming Python 4th editon.pdf
Python for Data Analysis
Python for Data Analysis.pdf
WebCrawler_Python
functional-programming-python.pdf
lpthw
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning $ " ls -al
total 156672
drwxr-xr-x  14 Hansen  staff      448  9  4  2017  .
drwxr-xr-x  32 Hansen  staff     1024  3 18 15:26  ..
-rw-r--r--@  1 Hansen  staff    14340  3 18 10:53  .DS_Store
-rw-r--r--@  1 Hansen  staff  26015425 12 14  2013  Learning_Python_5th_Edition_Mark_Lutz.pdf
-rw-r--r--@  1 Hansen  staff   3580937  7 16  2017  Natural Language Processing with Python.pdf
drwxr-xr-x   8 Hansen  staff     256  7 25  2016  Notebook_Python
-rw-r--r--@  1 Hansen  staff  11002456  3 28  2015  OReilly.Python.for.Finance.2014.12.pdf
-rw-r--r--@  1 Hansen  staff   1493142  1 17  2016  Oracle PL_SQL Language Pocket Reference, 5th Editi
on-O'Reilly(2015).pdf
-rw-r--r--@  1 Hansen  staff  27410567  6  1  2016  Programming Python 4th editon.pdf
drwxr-xr-x@ 30 Hansen  staff     960  3 17 22:34  Python for Data Analysis
-rw-r--r--@  1 Hansen  staff   8703373  3 18 11:38  Python for Data Analysis.pdf
drwxr-xr-x   6 Hansen  staff     192  5 28  2017  WebCrawler_Python
-rw-r--r--@  1 Hansen  staff   1631017  6  1  2016  functional-programming-python.pdf
drwxr-xr-x   4 Hansen  staff     128  7 16  2015  lpthw
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning $ " cd Python\ for\ Data\ An
ysis
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis $ "
sl
COPYING                ch04                ch07.ipynb            ch11.ipynb
My_Note                 ch04.ipynb          ch08                  ch12.ipynb
README.md               ch05                ch08.ipynb            ch13
appendix_python.ipynb  ch05.ipynb          ch09                  fec_study.ipynb
ch02                    ch06                ch09.ipynb
ch02.ipynb              ch06.ipynb          ch10.ipynb
ch03                    ch07                ch11
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis $ "
d cho2
-bash: cd: cho2: No such file or directory
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis $ "
d ch02
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data 2 $ " ls/ch0
movielens                usagov_bitly_data2012-03-16-1331923249.txt
names
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data 2 $ " cd mov
ielens/
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis/ch0
2/movielens $ " ls
README      movies.dat  ratings.dat  users.dat
Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis/ch0
2/movielens $ " clear

Hansen@Hansens-MacBook-Pro:~/Desktop/Intern_and_Projects/Python_learning/Python for Data Analysis/ch0
2/movielens $ " ipython
Python 2.7.12 |Anaconda custom (x86_64)| (default, Jul  2 2016, 17:43:17)
Type "copyright", "credits" or "license" for more information.

IPython 4.2.0 -- An enhanced Interactive Python.
?          -> Introduction and overview of IPython's features.
%quickref  -> Quick reference.
help       -> Python's own help system.
object?    -> Details about 'object', use 'object??' for extra details.

In [1]: import pandas as pd

```

```
In [2]: ls
README*      movies.dat* ratings.dat* users.dat*
```

```
In [3]: unames = ['user_id', 'gender', 'age', 'occupation', 'zip']
```

```
In [4]: users = pd.read_table('users.dat', sep = '::', header = None, names = rnames)
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-4-3ed10366e8ac> in <module>()
----> 1 users = pd.read_table('users.dat', sep = '::', header = None, names = rnames)
```

```
NameError: name 'rnames' is not defined
```

```
In [5]: users = pd.read_table('users.dat', sep = '::', header = None, names = unames)
/Users/Hansen/anaconda/bin/ipython:1: ParserWarning: Falling back to the 'python' engine because the
'c' engine does not support regex separators (separators > 1 char and different from '\s+' are interpreted as regex); you can avoid this warning by specifying engine='python'.
#!/bin/bash /Users/Hansen/anaconda/bin/python.app
```

```
In [6]: users = pd.read_table('users.dat', sep = '::', header = None, names = unames, engine = 'python')
```

```
In [7]: rnames = ['user_id', 'movie_id', 'rating', 'timestamp']
```

```
In [8]: ratings = pd.read_clipboard pd.read_gbq pd.read_msgpack pd.read_sql_query pd.reset_option
pd.read_csv pd.read_hdf pd.read_pickle pd.read_sql_table
pd.read_excel pd.read_html pd.read_sas pd.read_stata
pd.read_fwf pd.read_json pd.read_sql pd.read_table
```

```
In [8]: ratings = pd.read_table('ratings.dat', sep = '::', header = None, names = rnames, engine = 'python')
```

```
In [9]: mnames = ['movie_id', 'title', 'genres']
```

```
In [10]: movies = pd.read_table("movies.dat", sep='::', header = None, names = mnames)
/Users/Hansen/anaconda/bin/ipython:1: ParserWarning: Falling back to the 'python' engine because the
'c' engine does not support regex separators (separators > 1 char and different from '\s+' are interpreted as regex); you can avoid this warning by specifying engine='python'.
#!/bin/bash /Users/Hansen/anaconda/bin/python.app
```

```
In [11]: movies = pd.read_table("movies.dat", sep='::', header = None, names = mnames, engine = 'python')
```

```
In [12]: users[:5]
```

```
Out[12]:
```

	user_id	gender	age	occupation	zip
0	1	F	1	10	48067
1	2	M	56	16	70072
2	3	M	25	15	55117
3	4	M	45	7	02460
4	5	M	25	20	55455

```
In [13]: ratings
```

```
Out[13]:
```

	user_id	movie_id	rating	timestamp
0	1	1193	5	978300760
1	1	661	3	978302109
2	1	914	3	978301968
3	1	3408	4	978300275
4	1	2355	5	978824291
5	1	1197	3	978302268
6	1	1287	5	978302039
7	1	2804	5	978300719
8	1	594	4	978302268
9	1	919	4	978301368
10	1	595	5	978824268

11	1	938	4	978301752
12	1	2398	4	978302281
13	1	2918	4	978302124
14	1	1035	5	978301753
15	1	2791	4	978302188
16	1	2687	3	978824268
17	1	2018	4	978301777
18	1	3105	5	978301713
19	1	2797	4	978302039
20	1	2321	3	978302205
21	1	720	3	978300760
22	1	1270	5	978300055
23	1	527	5	978824195
24	1	2340	3	978300103
25	1	48	5	978824351
26	1	1097	4	978301953
27	1	1721	4	978300055
28	1	1545	4	978824139
29	1	745	3	978824268
...
1000179	6040	2762	4	956704584
1000180	6040	1036	3	956715455
1000181	6040	508	4	956704972
1000182	6040	1041	4	957717678
1000183	6040	3735	4	960971654
1000184	6040	2791	4	956715569
1000185	6040	2794	1	956716438
1000186	6040	527	5	956704219
1000187	6040	2003	1	956716294
1000188	6040	535	4	964828734
1000189	6040	2010	5	957716795
1000190	6040	2011	4	956716113
1000191	6040	3751	4	964828782
1000192	6040	2019	5	956703977
1000193	6040	541	4	956715288
1000194	6040	1077	5	964828799
1000195	6040	1079	2	956715648
1000196	6040	549	4	956704746
1000197	6040	2020	3	956715288
1000198	6040	2021	3	956716374
1000199	6040	2022	5	956716207
1000200	6040	2028	5	956704519
1000201	6040	1080	4	957717322
1000202	6040	1089	4	956704996
1000203	6040	1090	3	956715518
1000204	6040	1091	1	956716541
1000205	6040	1094	5	956704887
1000206	6040	562	5	956704746
1000207	6040	1096	4	956715648
1000208	6040	1097	4	956715569

[1000209 rows x 4 columns]

In [14]: ratings.size()

```

-----
TypeError                                Traceback (most recent call last)
<ipython-input-14-6eb952cee7c9> in <module>()
----> 1 ratings.size()
```

TypeError: 'numpy.int64' object is not callable

```

In [15]: ra
raise      range      ratings      ratings.dat  raw_input
```

```

In [15]: rat
ratings      ratings.dat
```

In [15]: ratings.

Display all 224 possibilities? (y or n)

ratings.T	ratings.get_ftype_counts	ratings.replace
ratings.abs	ratings.get_value	ratings.resample
ratings.add	ratings.get_values	ratings.reset_index
ratings.add_prefix	ratings.groupby	ratings.rfloordiv
ratings.add_suffix	ratings.gt	ratings.rmod
ratings.align	ratings.head	ratings.rmul
ratings.all	ratings.hist	ratings.rolling
ratings.any	ratings.iat	ratings.round
ratings.append	ratings.icol	ratings.rpow
ratings.apply	ratings.idxmax	ratings.rsub
ratings.applymap	ratings.idxmin	ratings.rtruediv
ratings.as_blocks	ratings.iget_value	ratings.sample
ratings.as_matrix	ratings.iloc	ratings.select
ratings.asfreq	ratings.index	ratings.select_dtypes
ratings.assign	ratings.info	ratings.sem
ratings.astype	ratings.insert	ratings.set_axis
ratings.at	ratings.interpolate	ratings.set_index
ratings.at_time	ratings.irow	ratings.set_value
ratings.axes	ratings.is_copy	ratings.shape
ratings.between_time	ratings.isin	ratings.shift
ratings.bfill	ratings.isnull	ratings.size
ratings.blocks	ratings.iteritems	ratings.skew
ratings.bool	ratings.iterkv	ratings.slice_shift
ratings.boxplot	ratings.iterrows	ratings.sort
ratings.clip	ratings.itertuples	ratings.sort_index
ratings.clip_lower	ratings.ix	ratings.sort_values
ratings.clip_upper	ratings.join	ratings.sortlevel
ratings.columns	ratings.keys	ratings.squeeze
ratings.combine	ratings.kurt	ratings.stack
ratings.combineAdd	ratings.kurtosis	ratings.std
ratings.combineMult	ratings.last	ratings.style
ratings.combine_first	ratings.last_valid_index	ratings.sub
ratings.compound	ratings.le	ratings.subtract
ratings.consolidate	ratings.loc	ratings.sum
ratings.convert_objects	ratings.lookup	ratings.swapaxes
ratings.copy	ratings.lt	ratings.swaplevel
ratings.corr	ratings.mad	ratings.tail
ratings.corrwith	ratings.mask	ratings.take
ratings.count	ratings.max	ratings.timestamp
ratings.cov	ratings.mean	ratings.to_clipboard
ratings.cummax	ratings.median	ratings.to_csv
ratings.cummin	ratings.memory_usage	ratings.to_dense
ratings.cumprod	ratings.merge	ratings.to_dict
ratings.cumsum	ratings.min	ratings.to_excel
ratings.dat	ratings.mod	ratings.to_gbq
ratings.describe	ratings.mode	ratings.to_hdf
ratings.diff	ratings.movie_id	ratings.to_html
ratings.div	ratings.mul	ratings.to_json
ratings.divide	ratings.multiply	ratings.to_latex
ratings.dot	ratings.ndim	ratings.to_msgpack
ratings.drop	ratings.ne	ratings.to_panel
ratings.drop_duplicates	ratings.nlargest	ratings.to_period
ratings.dropna	ratings.notnull	ratings.to_pickle
ratings.dtypes	ratings.nsmallest	ratings.to_records
ratings.duplicated	ratings.pct_change	ratings.to_sparse
ratings.empty	ratings.pipe	ratings.to_sql
ratings.eq	ratings.pivot	ratings.to_stata
ratings.equals	ratings.pivot_table	ratings.to_string
ratings.eval	ratings.plot	ratings.to_timestamp
ratings.ewm	ratings.pop	ratings.to_wide
ratings.expanding	ratings.pow	ratings.to_xarray
ratings.ffill	ratings.prod	ratings.transpose
ratings.fillna	ratings.product	ratings.truediv
ratings.filter	ratings.quantile	ratings.truncate
ratings.first	ratings.query	ratings.tshift
ratings.first_valid_index	ratings.radd	ratings.tz_convert
ratings.floordiv	ratings.rank	ratings.tz_localize

ratings.from_csv	ratings.rating	ratings.unstack
ratings.from_dict	ratings.rdiv	ratings.update
ratings.from_items	ratings.reindex	ratings.user_id
ratings.from_records	ratings.reindex_axis	ratings.values
ratings.ftypes	ratings.reindex_like	ratings.var
ratings.ge	ratings.rename	ratings.where
ratings.get	ratings.rename_axis	ratings.xs
ratings.get_dtype_counts	ratings.reorder_levels	

In [15]: ratings.size

Out[15]: 4000836

In [16]: data = pd.merge(pd.merge(ratings, users), movies)

In [17]: data

Out[17]:

	user_id	movie_id	rating	timestamp	gender	age	occupation	zip	\
0	1	1193	5	978300760	F	1	10	48067	
1	2	1193	5	978298413	M	56	16	70072	
2	12	1193	4	978220179	M	25	12	32793	
3	15	1193	4	978199279	M	25	7	22903	
4	17	1193	5	978158471	M	50	1	95350	
5	18	1193	4	978156168	F	18	3	95825	
6	19	1193	5	982730936	M	1	10	48073	
7	24	1193	5	978136709	F	25	7	10023	
8	28	1193	3	978125194	F	25	1	14607	
9	33	1193	5	978557765	M	45	3	55421	
10	39	1193	5	978043535	M	18	4	61820	
11	42	1193	3	978038981	M	25	8	24502	
12	44	1193	4	978018995	M	45	17	98052	
13	47	1193	4	977978345	M	18	4	94305	
14	48	1193	4	977975061	M	25	4	92107	
15	49	1193	4	978813972	M	18	12	77084	
16	53	1193	5	977946400	M	25	0	96931	
17	54	1193	5	977944039	M	50	1	56723	
18	58	1193	5	977933866	M	25	2	30303	
19	59	1193	4	977934292	F	50	1	55413	
20	62	1193	4	977968584	F	35	3	98105	
21	80	1193	4	977786172	M	56	1	49327	
22	81	1193	5	977785864	F	25	0	60640	
23	88	1193	5	977694161	F	45	1	02476	
24	89	1193	5	977683596	F	56	9	85749	
25	95	1193	5	977626632	M	45	0	98201	
26	96	1193	3	977621789	F	25	16	78028	
27	99	1193	2	982791053	F	1	10	19390	
28	102	1193	5	1040737607	M	35	19	20871	
29	104	1193	2	977546620	M	25	12	00926	
...	
1000179	4933	3084	3	962757020	M	25	15	94040	
1000180	4802	2218	2	1014866656	M	56	1	40601	
1000181	4812	2308	2	962932391	M	18	14	25301	
1000182	4874	624	4	962781918	F	25	4	70808	
1000183	5059	1434	4	962484364	M	45	16	22652	
1000184	5947	1434	4	957190428	F	45	16	97215	
1000185	5077	1868	3	962417299	M	25	2	20037	
1000186	5944	1868	1	957197520	F	18	10	27606	
1000187	5105	404	3	962337582	M	50	7	18977	
1000188	5185	404	4	963402617	F	35	4	44485	
1000189	5532	404	5	959619841	M	25	17	27408	
1000190	5543	404	3	960127592	M	25	17	97401	
1000191	5220	2543	3	961546137	M	25	7	91436	
1000192	5754	2543	4	958272316	F	18	1	60640	
1000193	5227	591	3	961475931	M	18	10	64050	
1000194	5795	591	1	958145253	M	25	1	92688	
1000195	5313	3656	5	960920392	M	56	0	55406	
1000196	5328	2438	4	960838075	F	25	4	91740	
1000197	5334	3323	3	960796159	F	56	13	46140	
1000198	5334	127	1	960795494	F	56	13	46140	

1000199	5334	3382	5	960796159	F	56	13	46140
1000200	5420	1843	3	960156505	F	1	19	14850
1000201	5433	286	3	960240881	F	35	17	45014
1000202	5494	3530	4	959816296	F	35	17	94306
1000203	5556	2198	3	959445515	M	45	6	92103
1000204	5949	2198	5	958846401	M	18	17	47901
1000205	5675	2703	3	976029116	M	35	14	30030
1000206	5780	2845	1	958153068	M	18	17	92886
1000207	5851	3607	5	957756608	F	18	20	55410
1000208	5938	2909	4	957273353	M	25	1	35401

					title \	
0					One Flew Over the Cuckoo's Nest (1975)	
1					One Flew Over the Cuckoo's Nest (1975)	
2					One Flew Over the Cuckoo's Nest (1975)	
3					One Flew Over the Cuckoo's Nest (1975)	
4					One Flew Over the Cuckoo's Nest (1975)	
5					One Flew Over the Cuckoo's Nest (1975)	
6					One Flew Over the Cuckoo's Nest (1975)	
7					One Flew Over the Cuckoo's Nest (1975)	
8					One Flew Over the Cuckoo's Nest (1975)	
9					One Flew Over the Cuckoo's Nest (1975)	
10					One Flew Over the Cuckoo's Nest (1975)	
11					One Flew Over the Cuckoo's Nest (1975)	
12					One Flew Over the Cuckoo's Nest (1975)	
13					One Flew Over the Cuckoo's Nest (1975)	
14					One Flew Over the Cuckoo's Nest (1975)	
15					One Flew Over the Cuckoo's Nest (1975)	
16					One Flew Over the Cuckoo's Nest (1975)	
17					One Flew Over the Cuckoo's Nest (1975)	
18					One Flew Over the Cuckoo's Nest (1975)	
19					One Flew Over the Cuckoo's Nest (1975)	
20					One Flew Over the Cuckoo's Nest (1975)	
21					One Flew Over the Cuckoo's Nest (1975)	
22					One Flew Over the Cuckoo's Nest (1975)	
23					One Flew Over the Cuckoo's Nest (1975)	
24					One Flew Over the Cuckoo's Nest (1975)	
25					One Flew Over the Cuckoo's Nest (1975)	
26					One Flew Over the Cuckoo's Nest (1975)	
27					One Flew Over the Cuckoo's Nest (1975)	
28					One Flew Over the Cuckoo's Nest (1975)	
29					One Flew Over the Cuckoo's Nest (1975)	
...					...	
1000179					Home Page (1999)	
1000180					Juno and Paycock (1930)	
1000181					Detroit 9000 (1973)	
1000182					Condition Red (1995)	
1000183					Stranger, The (1994)	
1000184					Stranger, The (1994)	
1000185					Truce, The (1996)	
1000186					Truce, The (1996)	
1000187	Brother Minister: The Assassination of Malcolm...					
1000188	Brother Minister: The Assassination of Malcolm...					
1000189	Brother Minister: The Assassination of Malcolm...					
1000190	Brother Minister: The Assassination of Malcolm...					
1000191					Six Ways to Sunday (1997)	
1000192					Six Ways to Sunday (1997)	
1000193					Tough and Deadly (1995)	
1000194					Tough and Deadly (1995)	
1000195					Lured (1947)	
1000196					Outside Ozona (1998)	
1000197					Chain of Fools (2000)	
1000198	Silence of the Palace, The (Saimt el Qusur) (1...					
1000199					Song of Freedom (1936)	
1000200					Slappy and the Stinkers (1998)	
1000201					Nemesis 2: Nebula (1995)	
1000202					Smoking/No Smoking (1993)	
1000203					Modulations (1998)	

1000204	Modulations (1998)
1000205	Broken Vessels (1998)
1000206	White Boys (1999)
1000207	One Little Indian (1973)
1000208	Five Wives, Three Secretaries and Me (1998)

	genres
0	Drama
1	Drama
2	Drama
3	Drama
4	Drama
5	Drama
6	Drama
7	Drama
8	Drama
9	Drama
10	Drama
11	Drama
12	Drama
13	Drama
14	Drama
15	Drama
16	Drama
17	Drama
18	Drama
19	Drama
20	Drama
21	Drama
22	Drama
23	Drama
24	Drama
25	Drama
26	Drama
27	Drama
28	Drama
29	Drama
...	...
1000179	Documentary
1000180	Drama
1000181	Action Crime
1000182	Action Drama Thriller
1000183	Action
1000184	Action
1000185	Drama War
1000186	Drama War
1000187	Documentary
1000188	Documentary
1000189	Documentary
1000190	Documentary
1000191	Comedy
1000192	Comedy
1000193	Action Drama Thriller
1000194	Action Drama Thriller
1000195	Crime
1000196	Drama Thriller
1000197	Comedy Crime
1000198	Drama
1000199	Drama
1000200	Children's Comedy
1000201	Action Sci-Fi Thriller
1000202	Comedy
1000203	Documentary
1000204	Documentary
1000205	Drama
1000206	Drama
1000207	Comedy Drama Western
1000208	Documentary

[1000209 rows x 10 columns]

```
In [18]: data.ix[0]
```

```
Out[18]:
```

```
user_id          1
movie_id        1193
rating           5
timestamp       978300760
gender           F
age             1
occupation      10
zip             48067
title           One Flew Over the Cuckoo's Nest (1975)
genres          Drama
Name: 0, dtype: object
```

```
In [19]: data.ix?
```

```
Type:          property
```

```
String form: <property object at 0x112e0c788>
```

```
Docstring:
```

A primarily label-location based indexer, with integer position fallback.

```ix[]``` supports mixed integer and label based access. It is primarily label based, but will fall back to integer positional access unless the corresponding axis is of integer type.

```ix``` is the most general indexer and will support any of the inputs in ```loc``` and ```iloc```. ```ix``` also supports floating point label schemes. ```ix``` is exceptionally useful when dealing with mixed positional and label based hierarchical indexes.

However, when an axis is integer based, ONLY label based access and not positional access is supported. Thus, in such cases, it's usually better to be explicit and use ```iloc``` or ```loc```.

See more at :ref:`Advanced Indexing <advanced>`.

```
In [20]: mean_ratings = data.pivot_table('rating', row = 'title', cols = 'gender', aggfunc='mean')
```

```
TypeError
```

```
Traceback (most recent call last)
```

```
<ipython-input-20-7a3b42bb1510> in <module>()
```

```
----> 1 mean_ratings = data.pivot_table('rating', row = 'title', cols = 'gender', aggfunc='mean')
```

```
TypeError: pivot_table() got an unexpected keyword argument 'row'
```

```
In [21]: mean_ratings = data.pivot_table('rating', rows = 'title', cols = 'gender', aggfunc='mean')
```

```
TypeError
```

```
Traceback (most recent call last)
```

```
<ipython-input-21-818478d17263> in <module>()
```

```
----> 1 mean_ratings = data.pivot_table('rating', rows = 'title', cols = 'gender', aggfunc='mean')
```

```
TypeError: pivot_table() got an unexpected keyword argument 'rows'
```

```
In [22]: mean_ratings = data.pivot_table('rating', index = 'title', columns = 'gender', aggfunc='mean')
'
```

```
In [23]: mean_ratings[:5]
```

```
Out[23]:
```

gender	F	M
title		
\$1,000,000 Duck (1971)	3.375000	2.761905
'Night Mother (1986)	3.388889	3.352941
'Til There Was You (1997)	2.675676	2.733333
'burbs, The (1989)	2.793478	2.962085
...And Justice for All (1979)	3.828571	3.689024


```
In [24]: ratings_by_title = data.groupby('title').size()
```

```
In [25]: ra
raise      ratings      ratings_by_title
range      ratings.dat   raw_input
```

```
In [25]: ratings_by_title[:10]
```

```
Out[25]:
title
$1,000,000 Duck (1971)      37
'Night Mother (1986)      70
'Til There Was You (1997)  52
'burbs, The (1989)        303
...And Justice for All (1979) 199
1-900 (1994)              2
10 Things I Hate About You (1999) 700
101 Dalmatians (1961)      565
101 Dalmatians (1996)      364
12 Angry Men (1957)       616
dtype: int64
```

```
In [26]: active_titles = ratings_by_title.index[ratings_by_title >= 250]
```

```
In [27]: active_titles[:10]
```

```
Out[27]:
Index([u'burbs, The (1989)', u'10 Things I Hate About You (1999)',
      u'101 Dalmatians (1961)', u'101 Dalmatians (1996)',
      u'12 Angry Men (1957)', u'13th Warrior, The (1999)',
      u'2 Days in the Valley (1996)', u'20,000 Leagues Under the Sea (1954)',
      u'2001: A Space Odyssey (1968)', u'2010 (1984)'],
      dtype='object', name=u'title')
```

```
In [28]: active_titles.size
```

```
Out[28]: 1216
```

```
In [29]: active_titles
```

```
Out[29]:
Index([u'burbs, The (1989)', u'10 Things I Hate About You (1999)',
      u'101 Dalmatians (1961)', u'101 Dalmatians (1996)',
      u'12 Angry Men (1957)', u'13th Warrior, The (1999)',
      u'2 Days in the Valley (1996)', u'20,000 Leagues Under the Sea (1954)',
      u'2001: A Space Odyssey (1968)', u'2010 (1984)',
      ...,
      u'X-Men (2000)', u'Year of Living Dangerously (1982)',
      u'Yellow Submarine (1968)', u'You've Got Mail (1998)',
      u'Young Frankenstein (1974)', u'Young Guns (1988)',
      u'Young Guns II (1990)', u'Young Sherlock Holmes (1985)',
      u'Zero Effect (1998)', u'eXistenZ (1999)'],
      dtype='object', name=u'title', length=1216)
```

```
In [30]: mean_ratings = mean_ratings.ix[active_titles]
```

```
In [31]: mean_ratings
```

```
Out[31]:
gender      F      M
title
'burbs, The (1989)      2.793478  2.962085
10 Things I Hate About You (1999)  3.646552  3.311966
101 Dalmatians (1961)      3.791444  3.500000
101 Dalmatians (1996)      3.240000  2.911215
12 Angry Men (1957)      4.184397  4.328421
13th Warrior, The (1999)  3.112000  3.168000
2 Days in the Valley (1996)  3.488889  3.244813
20,000 Leagues Under the Sea (1954)  3.670103  3.709205
2001: A Space Odyssey (1968)  3.825581  4.129738
2010 (1984)      3.446809  3.413712
28 Days (2000)      3.209424  2.977707
39 Steps, The (1935)      3.965517  4.107692
```

54 (1998)	2.701754	2.782178
7th Voyage of Sinbad, The (1958)	3.409091	3.658879
8MM (1999)	2.906250	2.850962
About Last Night... (1986)	3.188679	3.140909
Absent Minded Professor, The (1961)	3.469388	3.446809
Absolute Power (1997)	3.469136	3.327759
Abyss, The (1989)	3.659236	3.689507
Ace Ventura: Pet Detective (1994)	3.000000	3.197917
Ace Ventura: When Nature Calls (1995)	2.269663	2.543333
Addams Family Values (1993)	3.000000	2.878531
Addams Family, The (1991)	3.186170	3.163498
Adventures in Babysitting (1987)	3.455782	3.208122
Adventures of Buckaroo Bonzai Across the 8th Di...	3.308511	3.402321
Adventures of Priscilla, Queen of the Desert, T...	3.989071	3.688811
Adventures of Robin Hood, The (1938)	4.166667	3.918367
African Queen, The (1951)	4.324232	4.223822
Age of Innocence, The (1993)	3.827068	3.339506
Agnes of God (1985)	3.534884	3.244898
...
White Men Can't Jump (1992)	3.028777	3.231061
Who Framed Roger Rabbit? (1988)	3.569378	3.713251
Who's Afraid of Virginia Woolf? (1966)	4.029703	4.096939
Whole Nine Yards, The (2000)	3.296552	3.404814
Wild Bunch, The (1969)	3.636364	4.128099
Wild Things (1998)	3.392000	3.459082
Wild Wild West (1999)	2.275449	2.131973
William Shakespeare's Romeo and Juliet (1996)	3.532609	3.318644
Willow (1988)	3.658683	3.453543
Willy Wonka and the Chocolate Factory (1971)	4.063953	3.789474
Witness (1985)	4.115854	3.941504
Wizard of Oz, The (1939)	4.355030	4.203138
Wolf (1994)	3.074074	2.899083
Women on the Verge of a Nervous Breakdown (1988)	3.934307	3.865741
Wonder Boys (2000)	4.043796	3.913649
Working Girl (1988)	3.606742	3.312500
World Is Not Enough, The (1999)	3.337500	3.388889
Wrong Trousers, The (1993)	4.588235	4.478261
Wyatt Earp (1994)	3.147059	3.283898
X-Files: Fight the Future, The (1998)	3.489474	3.493797
X-Men (2000)	3.682310	3.851702
Year of Living Dangerously (1982)	3.951220	3.869403
Yellow Submarine (1968)	3.714286	3.689286
You've Got Mail (1998)	3.542424	3.275591
Young Frankenstein (1974)	4.289963	4.239177
Young Guns (1988)	3.371795	3.425620
Young Guns II (1990)	2.934783	2.904025
Young Sherlock Holmes (1985)	3.514706	3.363344
Zero Effect (1998)	3.864407	3.723140
eXistenZ (1999)	3.098592	3.289086

[1216 rows x 2 columns]

```
In [32]: top_female_ratings = mean_ratings.sort_index(by = 'F', ascending = False)
/Users/Hansen/anaconda/bin/ipython:1: FutureWarning: by argument to sort_index is deprecated, pls use
.sort_values(by=...)
#!/bin/bash /Users/Hansen/anaconda/bin/python.app
```

```
In [33]: top_female_ratings = mean_ratings.sort_values(by = 'F', ascending = False)
```

```
In [34]: top_female_ratings[:10]
File "<ipython-input-34-f39e8306b1fb>", line 1
    top_female_ratings[:10]
    ^
```

SyntaxError: invalid syntax

```
In [35]: top_female_ratings[:10]
Out[35]:
```

gender	F	M
title		
Close Shave, A (1995)	4.644444	4.473795
Wrong Trousers, The (1993)	4.588235	4.478261
Sunset Blvd. (a.k.a. Sunset Boulevard) (1950)	4.572650	4.464589
Wallace & Gromit: The Best of Aardman Animation...	4.563107	4.385075
Schindler's List (1993)	4.562602	4.491415
Shawshank Redemption, The (1994)	4.539075	4.560625
Grand Day Out, A (1992)	4.537879	4.293255
To Kill a Mockingbird (1962)	4.536667	4.372611
Creature Comforts (1990)	4.513889	4.272277
Usual Suspects, The (1995)	4.513317	4.518248

```
In [36]: mean_ratings['diff'] = mean_ratings['M'] - mean_ratings['F']
```

```
In [37]: sorted_by_diff = mean_ratings.sort_value(by = 'diff')
```

```
AttributeError                                Traceback (most recent call last)
```

```
<ipython-input-37-1136462f41db> in <module>()
```

```
----> 1 sorted_by_diff = mean_ratings.sort_value(by = 'diff')
```

```
/Users/Hansen/anaconda/lib/python2.7/site-packages/pandas/core/generic.pyc in __getattr__(self, name)
```

```
2670         if name in self._info_axis:
```

```
2671             return self[name]
```

```
-> 2672         return object.__getattr__(self, name)
```

```
2673
```

```
2674         def __setattr__(self, name, value):
```

```
AttributeError: 'DataFrame' object has no attribute 'sort_value'
```

```
In [38]: sorted_by_diff = mean_ratings.sort_values(by = 'diff')
```

```
In [39]: sorted_by_diff
```

```
Out[39]:
```

gender	F	M	diff
title			
Dirty Dancing (1987)	3.790378	2.959596	-0.830782
Jumpin' Jack Flash (1986)	3.254717	2.578358	-0.676359
Grease (1978)	3.975265	3.367041	-0.608224
Little Women (1994)	3.870588	3.321739	-0.548849
Steel Magnolias (1989)	3.901734	3.365957	-0.535777
Anastasia (1997)	3.800000	3.281609	-0.518391
Rocky Horror Picture Show, The (1975)	3.673016	3.160131	-0.512885
Color Purple, The (1985)	4.158192	3.659341	-0.498851
Age of Innocence, The (1993)	3.827068	3.339506	-0.487561
Free Willy (1993)	2.921348	2.438776	-0.482573
French Kiss (1995)	3.535714	3.056962	-0.478752
Little Shop of Horrors, The (1960)	3.650000	3.179688	-0.470312
Guys and Dolls (1955)	4.051724	3.583333	-0.468391
Mary Poppins (1964)	4.197740	3.730594	-0.467147
Patch Adams (1998)	3.473282	3.008746	-0.464536
Grease 2 (1982)	2.243478	1.792553	-0.450925
Sound of Music, The (1965)	4.233677	3.783418	-0.450259
Never Been Kissed (1999)	3.452174	3.002538	-0.449636
Casper (1995)	3.291139	2.844444	-0.446695
Jewel of the Nile, The (1985)	3.494253	3.048096	-0.446157
South Pacific (1958)	3.887850	3.445087	-0.442764
Gone with the Wind (1939)	4.269841	3.829371	-0.440471
Sommersby (1993)	3.482270	3.046667	-0.435603
Skulls, The (2000)	3.053333	2.621145	-0.432188
Corrina, Corrina (1994)	3.523490	3.092593	-0.430897
Iron Eagle (1986)	2.968750	2.543933	-0.424817
Hope Floats (1998)	3.206186	2.782383	-0.423802
Beverly Hills Cop III (1994)	2.865385	2.443333	-0.422051
Sense and Sensibility (1995)	4.233333	3.819277	-0.414056
Parent Trap, The (1961)	3.805556	3.393333	-0.412222
...
Tommy Boy (1995)	3.170732	3.585366	0.414634

Eyes Wide Shut (1999)	2.891213	3.305949	0.414736
South Park: Bigger, Longer and Uncut (1999)	3.422481	3.846686	0.424206
Godzilla (Gojira) (1954)	2.948718	3.374046	0.425328
Raging Bull (1980)	3.831933	4.263441	0.431508
Reservoir Dogs (1992)	3.769231	4.213873	0.444642
Where Eagles Dare (1969)	3.384615	3.833333	0.448718
Kingpin (1996)	3.092308	3.557480	0.465173
Me, Myself and Irene (2000)	2.629310	3.096847	0.467537
Akira (1988)	3.511111	3.980344	0.469233
Wild Bunch, The (1969)	3.636364	4.128099	0.491736
From Dusk Till Dawn (1996)	2.735714	3.232558	0.496844
Big Trouble in Little China (1986)	2.987952	3.485030	0.497078
Beavis and Butt-head Do America (1996)	2.637363	3.135417	0.498054
Rocky II (1979)	2.741379	3.242894	0.501515
Barb Wire (1996)	1.585366	2.100386	0.515020
Fright Night (1985)	2.973684	3.500000	0.526316
Exorcist, The (1973)	3.537634	4.067239	0.529605
Animal House (1978)	3.628906	4.167192	0.538286
Porky's (1981)	2.296875	2.836364	0.539489
For a Few Dollars More (1965)	3.409091	3.953795	0.544704
Caddyshack (1980)	3.396135	3.969737	0.573602
Rocky III (1982)	2.361702	2.943503	0.581801
Hidden, The (1987)	3.137931	3.745098	0.607167
Evil Dead II (Dead By Dawn) (1987)	3.297297	3.909283	0.611985
Cable Guy, The (1996)	2.250000	2.863787	0.613787
Longest Day, The (1962)	3.411765	4.031447	0.619682
Dumb & Dumber (1994)	2.697987	3.336595	0.638608
Kentucky Fried Movie, The (1977)	2.878788	3.555147	0.676359
Good, The Bad and The Ugly, The (1966)	3.494949	4.221300	0.726351

[1216 rows x 3 columns]

```
In [40]: ratings_std_by_title = data.groupby('title')['rating'].std()
```

```
In [41]: ratings_std_by_title = ratings_
ratings_by_title      ratings_std_by_title
```

```
In [41]: ratings_std_by_title = ratings_std_by_title.ix[active_titles]
```

```
In [42]: ra
raise          ratings          ratings_by_title      raw_input
range          ratings.dat      ratings_std_by_title
```

```
In [42]: ratings_std_by_title.order(ascending=False)[:10]
/Users/Hansen/anaconda/bin/ipython:1: FutureWarning: order is deprecated, use sort_values(...)
#!/bin/bash /Users/Hansen/anaconda/bin/python.app
```

```
Out[42]:
title
Dumb & Dumber (1994)          1.321333
Blair Witch Project, The (1999)  1.316368
Natural Born Killers (1994)    1.307198
Tank Girl (1995)              1.277695
Rocky Horror Picture Show, The (1975)  1.260177
Eyes Wide Shut (1999)        1.259624
Evita (1996)                  1.253631
Billy Madison (1995)          1.249970
Fear and Loathing in Las Vegas (1998)  1.246408
Bicentennial Man (1999)       1.245533
Name: rating, dtype: float64
```

```
In [43]:
```