In [2]:

```
import pandas
food info = pandas. read csv("food info. csv")
print (type(food_info))
print (food info. dtypes)
print (help(pandas.read_csv))
<class 'pandas.core.frame.DataFrame'>
NDB No
                      int64
Shrt\_Desc
                     object
Water (g)
                    float64
Energ_Kcal
                      int64
Protein_(g)
                    float64
                    float64
Lipid_Tot_(g)
Ash_{g}(g)
                    float64
Carbohydrt_(g)
                    float64
Fiber_TD_(g)
                    float64
                    float64
Sugar Tot (g)
Calcium_(mg)
                    float64
Iron_(mg)
                    float64
Magnesium_(mg)
                    float64
Phosphorus_(mg)
                    float64
Potassium_(mg)
                    float64
Sodium (mg)
                    float64
Zinc_(mg)
                    float64
Copper_(mg)
                    float64
```

In [17]:

```
food_info. head()
#first_rows = food_info. head()
#first_rows
#food_info. tail(4)
#print (food_info. tali(3))
#print (food_info. columns)
#print (food_info. shape)
```

Out[17]:

	NDB_No	Shrt_Desc	Water_(g)	Energ_Kcal	Protein_(g)	Lipid_Tot_(g)	Ash_(g)	Carbohyd
0	1001	BUTTER WITH SALT	15.87	717	0.85	81.11	2.11	
1	1002	BUTTER WHIPPED WITH SALT	15.87	717	0.85	81.11	2.11	
2	1003	BUTTER OIL ANHYDROUS	0.24	876	0.28	99.48	0.00	
3	1004	CHEESE BLUE	42.41	353	21.40	28.74	5.11	
4	1005	CHEESE BRICK	41.11	371	23.24	29.68	3.18	

5 rows × 36 columns

In [20]:

```
#print (food_info. loc[0])

food_info. loc[6]

#food_info. loc[8620]
```

Out[20]:

NDB_No	1007	
Shrt_Desc	CHEESE CAMEMBERT	
Water_(g)	51.8	
Energ_Kcal	300	
Protein_(g)	19.8	
Lipid_Tot_(g)	24. 26	
Ash_(g)	3.68	
Carbohydrt_(g)	0.46	
Fiber_TD_(g)	0	
Sugar_Tot_(g)	0.46	
Calcium_(mg)	388	
Iron_(mg)	0.33	
Magnesium_(mg)	20	
Phosphorus_(mg)	347	
Potassium_(mg)	187	
Sodium_(mg)	842	
Zinc_(mg)	2.38	
Copper (mg)	0. 021	

In [22]:

```
#food_info. loc[3:6]

#two_five_ten = [2, 5, 10]

#food_info. loc[two_five_ten]

food_info. loc[[2, 5, 10]]
```

Out[22]:

	NDB_No	Shrt_Desc	Water_(g)	Energ_Kcal	Protein_(g)	Lipid_Tot_(g)	Ash_(g)	Carbohy
2	1003	BUTTER OIL ANHYDROUS	0.24	876	0.28	99.48	0.00	
5	1006	CHEESE BRIE	48.42	334	20.75	27.68	2.70	
10	1011	CHEESE COLBY	38.20	394	23.76	32.11	3.36	

3 rows × 36 columns

In [26]:

```
ndb_col = food_info["NDB_No"]
print (ndb_col)
#col_name = "NDB_No"
#ndb_col = food_info[aol_name]
...
```

In [33]:

```
columns = ["Zinc_(mg)", "Iron_(mg)"]
zinc_copper = food_info[columns]
print (zinc_copper)
```

_	t (ZINC_copp	
	Zinc_(mg)	Iron_(mg)
0	0.09	0.02
1	0.05	0.16
2	0.01	0.00
3	2.66	0.31
4	2.60	0.43
5	2.38	0.50
6	2.38	0.33
7	2.94	0.64
8	3. 43	0.16
9	2.79	0.21
10	3.07	0.76
11	0.40	0.07
12	0.33	0.16
13	0.47	0.15
14	0.51	0.13
15	0.38	0.14
16	0.51	0.38
17	3.75	0.44
10	9 00	0 65

In [34]:

```
col names = food info.columns.tolist()
print (col_names)
gram columns = []
for c in col names:
     if c. endswith ("(g)"):
          gram columns. append (c)
gram_df = food_info[gram_columns]
print (gram_df.head(3))
['NDB_No', 'Shrt_Desc', 'Water_(g)', 'Energ_Kcal', 'Protein_(g)', 'Lipid_Tot_(g)',
'Ash_(g)', 'Carbohydrt_(g)', 'Fiber_TD_(g)', 'Sugar_Tot_(g)', 'Calcium_(mg)', 'Iron_(mg)', 'Magnesium_(mg)', 'Phosphorus_(mg)', 'Potassium_(mg)', 'Sodium_(mg)', 'Zinc_
```

```
(mg)', 'Copper_(mg)', 'Manganese_(mg)', 'Selenium_(mcg)', 'Vit_C_(mg)', 'Thiamin_(mg)', 'Riboflavin_(mg)', 'Niacin_(mg)', 'Vit_B6_(mg)', 'Vit_B12_(mcg)', 'Vit_A_RAE', 'Vit_E_(mg)', 'Vit_D_IU', 'Vit_K_(mcg)', 'FA_Sat_(g)', 'FA_Sat_(g)', 'FA_Sat_(g)', 'FA_Sat_(g)', 'FA_Sat_(g)', 'RA_Sat_(g)', 'RA_
_Mono_(g)', 'FA_Poly_(g)', 'Cholestrl_(mg)']
                    Water (g) Protein (g) Lipid Tot (g) Ash (g) Carbohydrt (g)
                                                                                                                                      0.85
                                                                                                                                                                                                                                                                                                2.11
0
                                            15.87
                                                                                                                                                                                                                                81.11
                                                                                                                                                                                                                                                                                                                                                                                                       0.06
                                            15.87
                                                                                                                                      0.85
                                                                                                                                                                                                                                81.11
                                                                                                                                                                                                                                                                                                2.11
                                                                                                                                                                                                                                                                                                                                                                                                       0.06
1
2
                                                  0.24
                                                                                                                                      0.28
                                                                                                                                                                                                                                99.48
                                                                                                                                                                                                                                                                                                0.00
                                                                                                                                                                                                                                                                                                                                                                                                       0.00
                                                                                                          Sugar_Tot_(g) FA_Sat_(g) FA_Mono_(g) FA_Poly_(g)
                    Fiber_TD_(g)
0
                                                                             0.0
                                                                                                                                                                      0.06
                                                                                                                                                                                                                                        51.368
                                                                                                                                                                                                                                                                                                                          21.021
                                                                                                                                                                                                                                                                                                                                                                                                                    3.043
```

```
50.489
             0.0
                            0.06
                                                      23.426
                                                                     3.012
1
2
             0.0
                            0.00
                                        61.924
                                                      28.732
                                                                     3.694
```

[36]: In

```
print (food info["Iron (mg)"])
div_1000 = food_info["Iron_(mg)"] / 10000
print (div_1000)
add_100 = food_info["Iron_(mg)"] + 100
```

```
29
          0.41
         . . .
          9.00
8588
8589
          0.30
          0.10
8590
          1.63
8591
8592
         34.82
8593
          2.28
8594
          0.17
8595
          0.17
8596
          4.86
8597
          0.25
8598
          0.23
          0.13
8599
8600
          0.11
8601
          0.68
8602
          7.83
          3.11
8603
          0.30
8604
```

In [37]:

```
water_energy = food_info["Water_(g)"] * food_info["Energ_Kcal"]
water_energy = food_info["Water_(g)"] * food_info["Energ_Kcal"]
iron_grams = food_info["Iron_(mg)"] / 1000
print (food_info.shape)
food_info["Iron_(g)"] = iron_grams
print (food_info.shape)
(8618, 36)
(8618, 37)
```

In [40]:

```
food_info.sort_values("Sodium_(mg)", inplace=True)
print (food_info["Sodium_(mg)"])
food_info.sort_values("Sodium_(mg)", inplace=True, ascending=False)
print (food_info["Sodium_(mg)"])
2231
        0.0
407
        0.0
        0.0
6827
758
        0.0
748
        0.0
747
        0.0
701
        0.0
704
        0.0
705
        0.0
706
        0.0
707
        0.0
730
        0.0
738
        0.0
       . . .
8153
        NaN
8155
        NaN
8156
        NaN
        NaN
8157
        NaN
8158
8159
        NaN
```

In [69]:

```
import pandas as pd
import numpy as np
titanic_surivival = pd.read_csv("titanic_train.csv")
titanic_surivival.head()
```

Out[69]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	(
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	_
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	
4)	•

In [44]:

```
age = titanic_surivival["Age"]
print (age.loc[0:10])
age_is_null = pd.isnull(age)
#print (ageage_is_null)
age_null_true = age[age_is_null]
#print (ageage_null_true)
age_null_count = len(age_null_true)
print (age_null_count)
```

```
0
      22.0
      38.0
1
2
      26.0
3
      35.0
      35.0
4
5
      NaN
6
      54.0
7
       2.0
      27.0
8
      14.0
9
       4.0
10
Name: Age, dtype: float64
177
```

```
In [45]:
```

```
mean_age = sum(titanic_surivival["Age"]) / len(titanic_surivival["Age"])
print (mean_age)
```

nan

In [46]:

```
good_ages = titanic_surivival["Age"][age_is_null == False]
#print good_ages
correct_mean_age = sum(good_ages) / len(good_ages)
print (correct_mean_age)
```

29, 69911764705882

In [47]:

```
correct_mean_age = titanic_surivival["Age"].mean()
print (correct_mean_age)
```

29.69911764705882

In [70]:

```
passenger_classes = [1, 2, 3]
fares_by_class = {}
for this_class in passenger_classes:
    pclsaa_rows = titanic_surivival[titanic_surivival["Pclass"] == this_class]
    pclass_fares = pclass_rows["Fare"]
    fare_for_class = pclass_fares.mean()
    fares_by_class[this_class] = fare_for_class
    print (fares_by_class)
```

NameError: name 'pclass rows' is not defined

In [51]:

```
passenger_surival = titanic_surivival.pivot_table(index="Pclass", values="Survived", aggfunc=np.mean)
print (passenger_surival)
```

```
Survived Pclass 1 0.629630 2 0.472826 3 0.242363
```

In [52]:

```
passenger_age = titanic_surivival.pivot_table(index="Pclass", values="Age")
print (passenger_age)
```

Age

Pclass

1 38. 233441 2 29. 877630

3 25. 140620

In [55]:

port_stats = titanic_surivival.pivot_table(index="Embarked", values=["Fare", "Survived"], aggfunc=np. print (port_stats)

	Fare	Survived
Embarked		
C	10072. 2962	93
Q	1022. 2543	30
S	17439, 3988	217

In [57]:

```
drop_na_columns = titanic_surivival.dropna(axis=1)
new_titanic_survival = titanic_survival.dropna(axis=0, subset=["Age", "Sex"])
print (new_titanic_survival)
870
          0
                        349248
                                   7.8958
                                                     NaN
                                                                 S
                                                                S
                                  52. 5542
                                                     D35
871
          1
                         11751
872
         0
                           695
                                   5.0000
                                           B51 B53 B55
                                                                 S
                                   9.0000
873
                                                                 S
         0
                        345765
                                                     NaN
                                                                C
874
         0
                    P/PP 3381
                                  24.0000
                                                     NaN
                                                                C
875
         0
                          2667
                                   7. 2250
                                                     NaN
876
         0
                          7534
                                   9.8458
                                                     NaN
                                                                 S
                                                                 S
877
         0
                        349212
                                   7.8958
                                                     NaN
                                                     C50
                                                                 C
879
          1
                         11767
                                  83.1583
                                                                 S
880
          1
                        230433
                                  26.0000
                                                     NaN
                                                                 S
881
         0
                        349257
                                   7.8958
                                                     NaN
                                                                 S
882
         0
                          7552
                                  10.5167
                                                     NaN
         0
             C. A. /SOTON 34068
                                                                 S
883
                                  10.5000
                                                     NaN
                                                                 S
         0
              SOTON/OQ 392076
                                  7.0500
884
                                                     NaN
885
                        382652
                                  29.1250
                                                     NaN
                                                                 Q
         5
         0
                                                                 S
886
                        211536
                                  13.0000
                                                     NaN
                                                                 S
887
         0
                        112053
                                  30.0000
                                                     B42
                                                                C
889
         0
                        111369
                                  30.0000
                                                   C148
890
         0
                                                                 Q
                        370376
                                   7.7500
                                                     NaN
```

In [63]:

```
row_index_83_age = titanic_surivival.loc[83, "Age"]
row_index_100_pclass = titanic_surivival.loc[766, "Pclass"]
print (row_index_83_age)
print (row_index_100_pclass)
```

28.0

1

In [64]:

```
new_titanic_survival = titanic_survival.sort_values("Age", ascending=False)
print (new_titanic_survival[0:10])
titanic_reindexed = new_titanic_survival.reset_index(drop=True)
print ("-----")
print (titanic_reindexed.loc[0:10])
```

630 851 493 96 116 672 745	Passe	631 852 494 97 117 673 746		ved Po 1 0 0 0 0 0	class 1 3 1 1 3 2	Barkwo	Gold Mitche	Sver stagave dschmie Conne	non Henr nsson, M eytia, M dt, Mr. ors, Mr. r. Henry Edward	y Wils r. Joh r. Ram George Patri Micha	nan non e B .ck nel	
33		34		0	2		•	Whead	on, Mr.	Edward	l H	
54 280		55 281		0	1 3		Ostby, Mi	_	elhart C Duane, M			
	Sex	Age	SibSp	Parch	,	Ticket	Fare	Cabin	Embarke	d		
630	male	80.0	0	0		27042	30.0000	A23		S		
851	male	74.0	0	0	;	347060	7.7750	NaN		S		
493	male	71.0	0	0	PC	17609	49.5042	NaN		С		
96	male	71.0	0	0	PC	17754	34.6542	A5		С		
116	male	70.5	0	0	;	370369	7.7500	NaN		Q		
672	male	70.0	0	0	C. A.	24580	10. 5000	NaN		S		

In [65]:

```
def hundreth_row(column):
    hundredth_item = column.loc[99]
    return hundredth_item

hundreth_row = titanic_surivival.apply(hundreth_row)
print (hundreth_row)
```

PassengerId		100
Survived		0
Pclass		2
Name	Kantor,	Mr. Sinai
Sex		male
Age		34
SibSp		1
Parch		0
Ticket		244367
Fare		26
Cabin		NaN
Embarked		S
dtype: object		

```
In [67]:
```

```
def is minor(row):
    if row["Age"] < 18:
        return True
    else:
        return False
def generate_age_label(row):
    age = row["Age"]
    if pd. isnull(age):
        return "unknowa"
    elif age < 18:
        return "minor"
    else:
        return "adult"
age_labels = titanic_surivival.apply(generate_age_label, axis=1)
print (age_labels)
0
         adult
         adult
1
2
         adult
3
         adult
4
         adult
5
       unknowa
6
         adult
7
         minor
         adult
8
9
         minor
10
         minor
11
         adult
12
         adult
13
         adult
14
         minor
15
         adult
16
         minor
17
       unknowa
18
         adult
    [68]:
In
titanic surivival['age labels'] = age labels
age_group_survival = titanic_surivival.pivot_table(index="age_labels", values="Survived")
print (age_group_survival)
            Survived
age\_labels
adult
            0.381032
```

minor 0.539823 unknowa 0.293785

```
In [78]:
```

```
import pandas as pd
fandango = pd. read_csv('fandango_score_comparison.csv')
serier film = fandango['FILM']
print (type(serier film))
print (serier film[0:5])
series_rt = fandango['RottenTomatoes']
print (series rt[0:5])
<class 'pandas. core. series. Series'>
     Avengers: Age of Ultron (2015)
0
1
                  Cinderella (2015)
2
                     Ant-Man (2015)
3
             Do You Believe? (2015)
      Hot Tub Time Machine 2 (2015)
4
Name: FILM, dtype: object
0
     74
     85
1
2
     80
3
     18
4
     14
Name: RottenTomatoes, dtype: int64
In [85]:
from pandas import Series
film_names = serier_film.values
print (type(film_names))
#print (film names)
rt scores = series_rt.values
#orint (rt scores)
series_custom = Series(rt_scores , index=film_names)
series_custom[['Minions (2015)', 'Leviathan (2014)']]
<class 'numpy.ndarray'>
Out[85]:
Minions (2015)
                    54
Leviathan (2014)
                    99
dtype: int64
In [86]:
series_custom = Series(rt_scores , index=film_names)
series_custom[['Minions (2015)', 'Leviathan (2014)']]
fiveten = series custom[5:10]
print (fiveten)
The Water Diviner (2015)
                                 63
Irrational Man (2015)
                                 42
Top Five (2014)
                                 86
Shaun the Sheep Movie (2015)
                                 99
Love & Mercy (2015)
                                 89
dtype: int64
```

In [87]:

```
original index = series custom.index.tolist()
sorted_index = sorted(original_index)
sorted_by_index = series_custom.reindex(sorted_index)
print (sorted_by_index)
                                                      97
'71 (2015)
5 Flights Up (2015)
                                                      52
A Little Chaos (2015)
                                                      40
A Most Violent Year (2014)
                                                      90
About Elly (2015)
                                                      97
Aloha (2015)
                                                      19
American Sniper (2015)
                                                      72
American Ultra (2015)
                                                      46
Amy (2015)
                                                      97
Annie (2014)
                                                      27
                                                      80
Ant-Man (2015)
Avengers: Age of Ultron (2015)
                                                      74
Big Eyes (2014)
                                                      72
Birdman (2014)
                                                      92
Black Sea (2015)
                                                      82
Black or White (2015)
                                                      39
Blackhat (2015)
                                                      34
Cake (2015)
                                                      49
Chappie (2015)
                                                      30
01:11 44 (0015)
```

In [88]:

```
sc2 = series_custom.sort_index()
sc3 = series_custom.sort_values()
print (sc3[0:10])
```

```
Paul Blart: Mall Cop 2 (2015)
                                   5
                                   7
Hitman: Agent 47 (2015)
Hot Pursuit (2015)
                                   8
Fantastic Four (2015)
                                   9
Taken 3 (2015)
                                   9
The Boy Next Door (2015)
                                  10
The Loft (2015)
                                  11
Unfinished Business (2015)
                                  11
                                  12
Mortdecai (2015)
Seventh Son (2015)
                                  12
dtype: int64
```

In [128]:

```
import numpy as np
print (np. add(series_custom, series_custom))
np. sin (series custom)
np. max(series custom)
Avengers: Age of Ultron (2015)
                                                    148
Cinderella (2015)
                                                    170
Ant-Man (2015)
                                                     160
Do You Believe? (2015)
                                                     36
Hot Tub Time Machine 2 (2015)
                                                     28
The Water Diviner (2015)
                                                    126
Irrational Man (2015)
                                                     84
Top Five (2014)
                                                    172
Shaun the Sheep Movie (2015)
                                                    198
Love & Mercy (2015)
                                                    178
Far From The Madding Crowd (2015)
                                                    168
Black Sea (2015)
                                                    164
Leviathan (2014)
                                                    198
Unbroken (2014)
                                                    102
The Imitation Game (2014)
                                                    180
Taken 3 (2015)
                                                     18
Ted 2 (2015)
                                                     92
Southpaw (2015)
                                                    118
Night at the Museum: Secret of the Tomb (2014)
                                                    100
```

In [92]:

series_custom > 50

Out[92]:

```
Avengers: Age of Ultron (2015)
                                                     True
Cinderella (2015)
                                                     True
Ant-Man (2015)
                                                     True
Do You Believe? (2015)
                                                    False
Hot Tub Time Machine 2 (2015)
                                                    False
The Water Diviner (2015)
                                                     True
Irrational Man (2015)
                                                    False
Top Five (2014)
                                                     True
Shaun the Sheep Movie (2015)
                                                     True
Love & Mercy (2015)
                                                     True
Far From The Madding Crowd (2015)
                                                     True
Black Sea (2015)
                                                     True
Leviathan (2014)
                                                     True
Unbroken (2014)
                                                     True
The Imitation Game (2014)
                                                     True
Taken 3 (2015)
                                                    False
Ted 2 (2015)
                                                    False
Southpaw (2015)
                                                     True
```

In [94]:

In [95]:

```
import numpy as np
types = fandango_films.dtypes
print (types)
float_columns = types[types.values == 'float64'].index
float_df = fandango_films[float_columns]
print (float_df)
deviations = float_df.apply(lambda x: np.std(x))
```

```
FILM
                                object
RottenTomatoes
                                 int64
RottenTomatoes User
                                 int64
Metacritic
                                 int64
Metacritic_User
                               float64
                               float64
IMDB
Fandango Stars
                               float64
Fandango_Ratingvalue
                               float64
RT norm
                               float64
RT user norm
                               float64
Metacritic norm
                               float64
Metacritic user nom
                               float64
                               float64
IMDB norm
RT norm round
                               float64
RT user norm round
                               float64
Metacritic norm round
                               float64
Metacritic_user_norm_round
                               float64
IMDB norm round
                               float64
Metacritic user vote count
                                 int64
```

In [96]:

```
fandango_films["Avengers: Age of Ultron (2015)": "Hot Tub Time Machine 2 (2015)"]
fandango_films.loc["Avengers: Age of Ultron (2015)": "Hot Tub Time Machine 2 (2015)"]
fandango_films.loc['Kingsman: The Secret Service (2015)']
movies = ['Kingsman: The Secret Service (2015)', 'Do You Believe? (2015)', 'The Water Diviner (2015)
fandango_films.loc[movies]
```

Out[96]:

	FILM	RottenTomatoes	RottenTomatoes_User	Metacritic	Metacritic_User	IMDB
FILM						
Kingsman: The Secret Service (2015)	Kingsman: The Secret Service (2015)	75	84	58	7.9	7.8
Do You Believe? (2015)	Do You Believe? (2015)	18	84	22	4.7	5.4
The Water Diviner (2015)	The Water Diviner (2015)	63	62	50	6.8	7.2
3 rows × 22 columns						

In [98]:

```
fandango = pd.read_csv('fandango_score_comparison.csv')
print (type(fandango))
fandango_films = fandango.set_index('FILM', drop=False)
print (fandango_films.index)
```

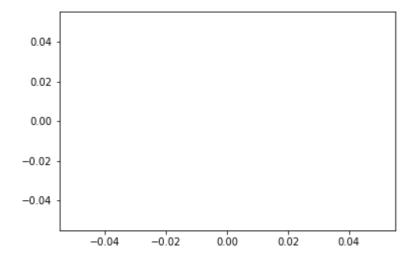
In [100]:

```
import pandas as pd
unrate = pd.read_csv('unrate.csv')
unrate['DATE'] = pd.to_datetime(unrate['DATE'])
print (unrate.head())
```

```
DATE VALUE
0 1948-01-01 3.4
1 1948-02-01 3.8
2 1948-03-01 4.0
3 1948-04-01 3.9
4 1948-05-01 3.5
```

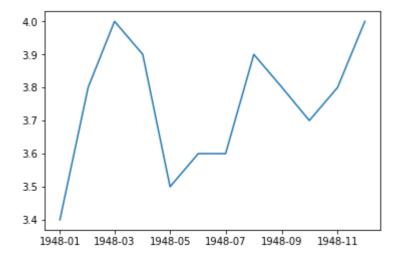
In [102]:

```
import matplotlib.pyplot as plt
plt.plot()
plt.show()
```



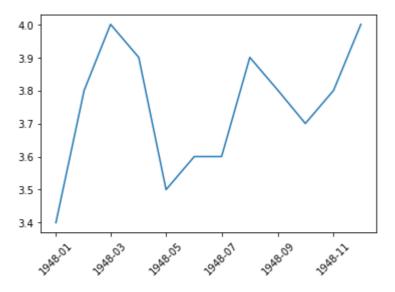
In [103]:

```
first_twelve = unrate[0:12]
plt.plot(first_twelve['DATE'], first_twelve['VALUE'])
plt.show()
```



In [104]:

```
plt.plot(first_twelve['DATE'], first_twelve['VALUE'])
plt.xticks(rotation=45)
plt.show()
```



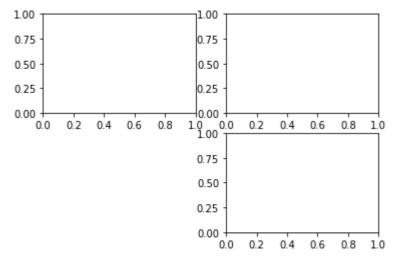
In [106]:

```
plt.plot(first_twelve['DATE'], first_twelve['VALUE'])
plt.xticks(rotation=45)
plt.xlabel('Month')
plt.ylabel('Unemployment Rate')
plt.title('Monthly Unemployment Trends, 1948')
plt.show()
```



In [108]:

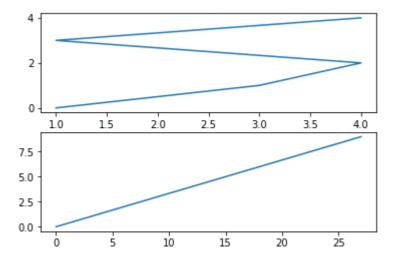
```
import matplotlib.pylab as plt
fig = plt.figure()
ax1 = fig.add_subplot(2, 2, 1)
ax2 = fig.add_subplot(2, 2, 2)
ax2 = fig.add_subplot(2, 2, 4)
plt.show()
```



In [110]:

```
import numpy as np
fig = plt.figure()
ax1 = fig.add_subplot(2,1,1)
ax2 = fig.add_subplot(2,1,2)

ax1.plot(np.random.randint(1,5,5), np.arange(5))
ax2.plot(np.arange(10)*3, np.arange(10))
plt.show()
```



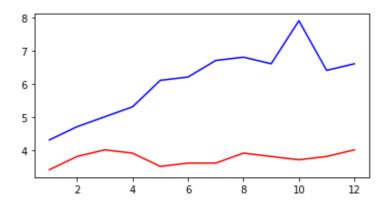
In [112]:

```
unrate['MONTH'] = unrate['DATE'].dt.month
unrate['MONTH'] = unrate['DATE'].dt.month
fig = plt.figure(figsize=(6,3))

plt.plot(unrate[0:12]['MONTH'], unrate[0:12]['VALUE'], c='red')
plt.plot(unrate[12:24]['MONTH'], unrate[12:24]['VALUE'], c='blue')
```

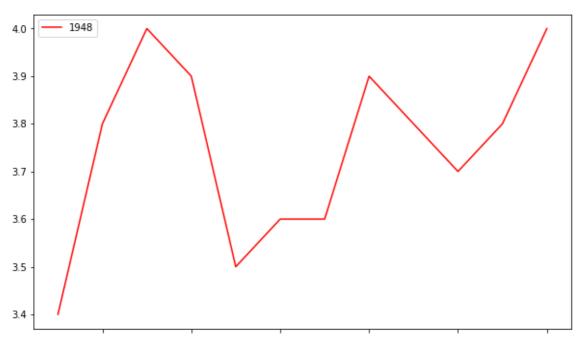
Out[112]:

[<matplotlib.lines.Line2D at 0x21c1bee4c18>]



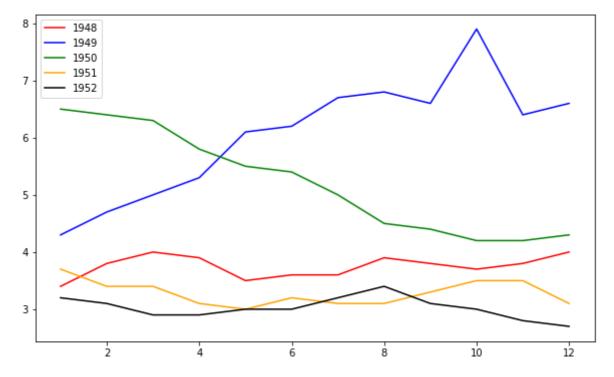
In [119]:

```
fig = plt.figure(figsize=(10,6))
colors = ['red', 'blue', 'green', 'orange', 'black']
for i in range(5):
    start_index = i*12
    end_index = (i+1)*12
    subset = unrate[start_index:end_index]
    label = str(1948 + i)
    plt.plot(subset['MONTH'], subset['VALUE'], c=colors[i], label=label)
    plt.legend(loc='best')
    plt.show()
```



In [116]:

```
fig = plt.figure(figsize=(10,6))
colors = ['red', 'blue', 'green', 'orange', 'black']
for i in range(5):
    start_index = i*12
    end_index = (i+1)*12
    subset = unrate[start_index:end_index]
    label = str(1948 + i)
    plt.plot(subset['MONTH'], subset['VALUE'], c=colors[i], label=label)
plt.legend(loc='upper left')
plt.show()
```



In [120]:

```
fig = plt.figure(figsize=(10,6))
colors = ['red', 'blue', 'green', 'orange', 'black']
for i in range(5):
    start_index = i*12
    end_index = (i+1)*12
    subset = unrate[start_index:end_index]
    label = str(1948 + i)
    plt.plot(subset['MONTH'], subset['VALUE'], c=colors[i], label=label)
plt.legend(loc='upper left')
plt.xlabel('Month')
plt.ylabel('Unemployment Rate')
plt.title('Monthly Unemployment Trends, 1948')
plt.show()
```

Monthly Unemployment Trends, 1948 1948 1949 1950 1951 1952 4 3 4 Month

In [149]:

```
import pandas as pd
reviews = pd.read_csv('fandango_score_comparison.csv')
cols = ['FILM', 'RT_user_norm', 'Metacritic', 'IMDB_norm', 'Fandango_Ratingvalue', 'Fandango_Stars']
norm_reviews = reviews[cols]
print(norm_reviews[:1])
```

```
FILM RT_user_norm Metacritic IMDB_norm \
0 Avengers: Age of Ultron (2015) 4.3 66 3.9

Fandango_Ratingvalue Fandango_Stars
0 4.5 5.0
```

```
In [147]:
```

```
import matplotlib.pyplot as plt
from numpy import arange
num_cols = ['FILM', 'RT_user_norm', 'Metacritic', 'IMDB_norm', 'Fandango_Ratingvalue', 'Fandango_Sta
bar_heights = norm_reviews.ix[0, num_cols].values
print (bar heights)
bar_positions = arange(5) + 0.75
print (bar_positions)
fig, ax = plt. subplots()
#ax. bar(bar_heights, bar_positions, 0.5)
plt.show()
['Avengers: Age of Ultron (2015)' 4.3 66 3.9 4.5 5.0]
[0.75 1.75 2.75 3.75 4.75]
D:\Anaconda\ancaonda\lib\site-packages\ipykernel_launcher.py:4: DeprecationWarnin
g:
.ix is deprecated. Please use
.loc for label based indexing or
.iloc for positional indexing
See the documentation here:
http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecated
(http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecate
  after removing the cwd from sys. path.
1.0
 0.8
```

In [150]:

```
import matplotlib.pyplot as plt
from numpy import arange
num_cols = ['RT_user_norm', 'Metacritic_user_nom', 'IMDB_norm', 'Fandango_Ratingvalue', 'Fandango_St
bar_widths = norm_reviews.ix[0, num_cols].values
bar_positions = arange(5) + 0.75
tick_positions = range(1,6)
fig, ax = plt.subplots()
ax.barh(bar_positions, bar_widths, 0.5)

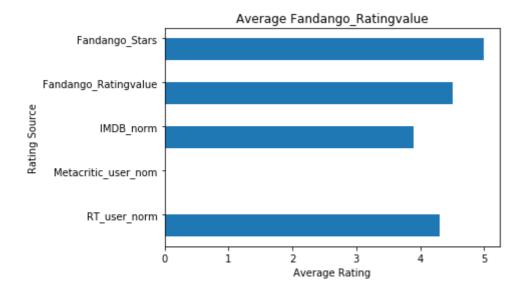
ax.set_yticks(tick_positions)
ax.set_yticklabels(num_cols)
ax.set_ytlabel('Rating_Source')
ax.set_vlabel('Average_Rating')
ax.set_title('Average_Fandango_Ratingvalue')
plt.show()
```

D:\Anaconda\ancaonda\lib\site-packages\ipykernel_launcher.py:5: DeprecationWarning:

- .ix is deprecated. Please use
- .loc for label based indexing or
- .iloc for positional indexing

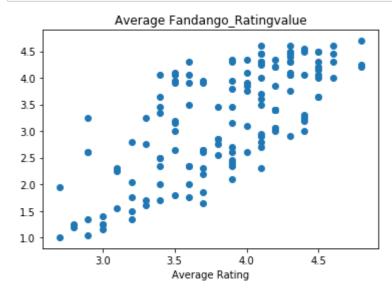
See the documentation here:

 $http://pandas.\ pydata.\ org/pandas-docs/stable/indexing.\ html\#ix-indexer-is-deprecated \ (http://pandas.\ pydata.\ org/pandas-docs/stable/indexing.\ html#ix-indexer-is-deprecated)$



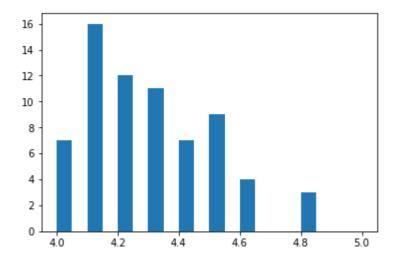
In [151]:

```
fix, ax = plt.subplots()
ax.scatter(norm_reviews['Fandango_Ratingvalue'], norm_reviews['RT_user_norm'])
ax.set_xlabel('Average Rating')
ax.set_title('Average Fandango_Ratingvalue')
plt.show()
```



In [152]:

```
fig, ax = plt.subplots()
ax.hist(norm_reviews['Fandango_Ratingvalue'], range=(4, 5), bins=20)
plt.show()
```



In [153]:

```
Cinderella (2015)
                                               4.0
                                                             67
                                                                       3.55
1
2
                    Ant-Man (2015)
                                               4.5
                                                             64
                                                                       3.90
                                                             22
3
           Do You Believe? (2015)
                                               4.2
                                                                       2.70
    Hot Tub Time Machine 2 (2015)
                                                             29
4
                                               1.4
                                                                       2.55
```

In [156]:

```
fandango_distribution = norm_reviews['Fandango_Ratingvalue'].value_counts()
fandango_distribution = fandango_distribution.sort_index()

imdb_distribution = norm_reviews['IMDB_norm'].value_counts()
imdb_distribution = imdb_distribution.sort_index()

print(fandango_distribution)
print(imdb_distribution)
3.8 5
```

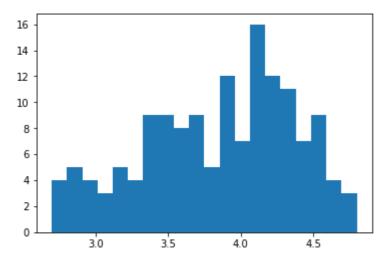
```
3.9
       12
4.0
        7
       16
4.1
4.2
       12
4.3
       11
4.4
        7
        9
4.5
4.6
        4
4.8
Name: Fandango Ratingvalue, dtype: int64
2.00
         1
```

2

2.60

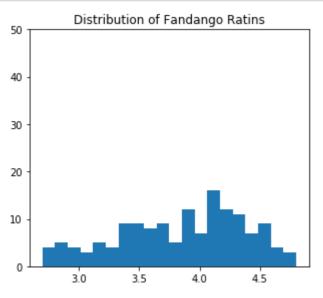
In [202]:

```
fig, ax = plt.subplots()
#ax. hist (norm_reviews['Fandango_Ratingvalue'])
#ax. hist (norm_reviews['Fandango_Ratingvalue'], bins=20)
ax. hist (norm_reviews['Fandango_Ratingvalue'] , bins=20)
plt. show()
```



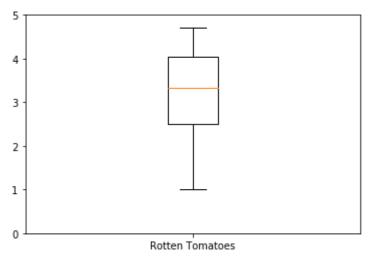
In [185]:

```
fig = plt.figure(figsize=(5, 20))
ax1 = fig.add_subplot(4, 1, 1)
#ax2 = fig.add_subplot(4, 1, 2)
#ax3 = fig.add_subplot(4, 1, 3)
#ax4 = fig.add_subplot(4, 1, 4)
ax1.hist(norm_reviews['Fandango_Ratingvalue'], bins=20)
ax1.set_title('Distribution of Fandango Ratins')
ax1.set_ylim(0, 50)
plt.show()
```



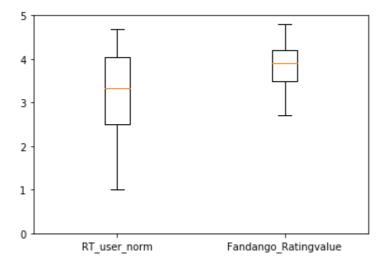
In [187]:

```
fig, ax = plt.subplots()
ax.boxplot(norm_reviews['RT_user_norm'])
ax.set_xticklabels(['Rotten Tomatoes'])
ax.set_ylim(0,5)
plt.show()
```



In [201]:

```
num_cols = ['RT_user_norm', 'Fandango_Ratingvalue']
fig, ax = plt.subplots()
ax.boxplot(norm_reviews[num_cols].values)
ax.set_xticklabels(num_cols, rotation=0)
ax.set_ylim(0,5)
plt.show()
```



In [206]:

```
import pandas as pd
import matplotlib.pyplot as plt
women_degrees = pd.read_csv('percent-bachelors-degrees-women-usa.csv')
major_cats = ['Biology', 'Computer Science', 'Engineering', 'Math and Statistics']
cb_dark_blue = (0/255, 107/255, 164/255)
cb_orange = (255/255, 128/255, 14/255)
fig = plt.figure(figsize=(12, 12))
for sp in range (0, 4):
    ax = fig. add\_subplot(2, 2, sp+1)
    ax.plot(women_degrees['Year'], women_degrees[major_cats[sp]], c=cb_dark_blue, label='Women')
    ax.plot(women_degrees['Year'], 100-women_degrees[major_cats[sp]], c=cb_orange, label='Men')
    for key, spine in ax. spines. items():
        spine. set visible (False)
    ax.set_xlim(1968, 2011)
    ax. set_ylim(0, 100)
    ax.set_title(major_cats[sp])
    #ax. tick_params(bottom="off", top="off", left="off", ringht="off")
plt.legend(loc='upper right')
plt.show()
```

