PROGRAM-11: DATA VISUALIZATION

REQUIREMENT:

With the help of suitable data and plots of your choice discuss how the data visualization can lead to misleading information.

IMPORTING LIBRARIES

In [2]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
```

ABOUT DATASET

This dataset consists of data if an advertisement will be success or not.

IMPORTING DATASET AND DISPLAYING IT

```
In [3]:
```

```
media = pd.read_csv ('Media.csv')
actual = pd.read_csv ('Actual.csv')
```

In [4]:

media.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 6513 entries, 0 to 6512 Data columns (total 11 columns): 6513 non-null int64 6513 non-null object realtionship_status 6513 non-null object Industry 6513 non-null object genre Targeted Sex 6513 non-null object 6513 non-null int64 average_min_perweek Airtime 6513 non-null object airlocation 6513 non-null object ratings 6513 non-null float64 expensive 6513 non-null object 6513 non-null object money_back_guarantee dtypes: float64(1), int64(2), object(8) memory usage: 559.8+ KB

In [5]:

actual.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 26048 entries, 0 to 26047
Data columns (total 12 columns):
                        26048 non-null int64
                        26048 non-null object
realtionship_status
                        26048 non-null object
industry
genre
                        26048 non-null object
Targeted Sex
                        26048 non-null object
                        26048 non-null int64
average_min_perweek
                        26048 non-null object
Airtime
airlocation
                        26048 non-null object
ratings
                        26048 non-null float64
                        26048 non-null object
expensive
                        26048 non-null object
money_back_guarantee
                        26048 non-null bool
netgain
dtypes: bool(1), float64(1), int64(2), object(8)
memory usage: 2.2+ MB
```

In [6]:

```
media=media.drop(['money_back_guarantee'],axis = 1)
media.head(7)
```

Out[6]:

	id	realtionship_status	Industry	genre	Targeted Sex	average_min_perweek	Airtime
0	1	Widowed	Auto	Comedy	Male	10	Daytime
1	4	Married-civ-spouse	Pharma	Comedy	Female	40	Morning
2	5	Divorced	Entertainment	Comedy	Male	50	Morning
3	9	Married-civ-spouse	Pharma	Infomercial	Female	40	Primetime
4	10	Married-civ-spouse	Pharma	Comedy	Female	40	Primetime
5	20	Never-married	Entertainment	Comedy	Male	40	Primetime
6	28	Divorced	Auto	Comedy	Female	40	Primetime

In [7]:

```
actual=actual.drop(['money_back_guarantee','netgain'],axis = 1)
actual.head(7)
```

Out[7]:

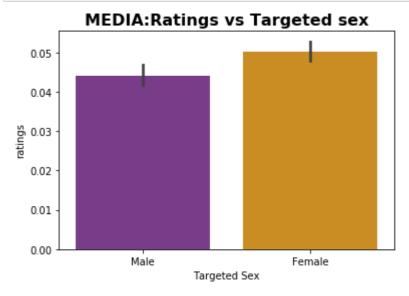
Airtin	average_min_perweek	Targeted Sex	genre	industry	realtionship_status	id	
Primetir	45	Male	Comedy	Auto	Married-spouse- absent	19717	0
Primetir	45	Male	Comedy	Pharma	Married-civ-spouse	31593	1
Primetir	45	Female	Comedy	Entertainment	Divorced	5681	2
Primetin	40	Female	Infomercial	Political	Separated	15491	3
Primetir	48	Male	Comedy	Pharma	Married-civ-spouse	23587	4
Primetir	40	Female	Comedy	Auto	Divorced	28523	5
Morni	50	Male	Infomercial	Pharma	Married-civ-spouse	12290	6
>							4

VISUALIZATION

1. RATINGS VS TARGET SEX

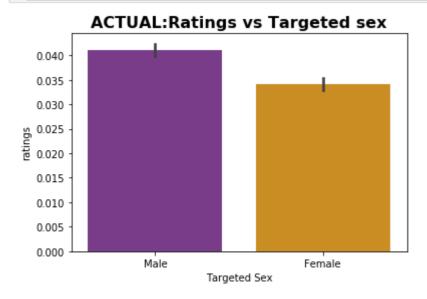
In [8]:

```
ax = sns.barplot(x="Targeted Sex", y="ratings", palette="CMRmap",data=media).set_title('MED
```



In [9]:

```
ax = sns.barplot(x="Targeted Sex", y="ratings", palette="CMRmap",data=actual).set_title('AC
```



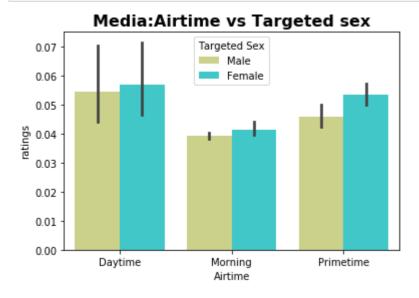
INFERENCE:

Compairing Media and Actual Dataset its clear that media has maniplucated the dataset. In Actual, the target sex is Male but when we compare with Media its clear that target sex is female.

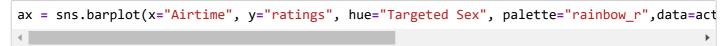
2. RATINGS vs AIRTIME

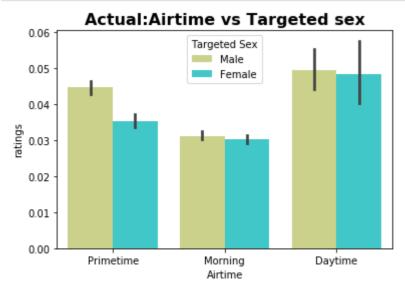
In [10]:

```
ax = sns.barplot(x="Airtime", y="ratings", hue="Targeted Sex",palette="rainbow_r", data=med
```



In [11]:





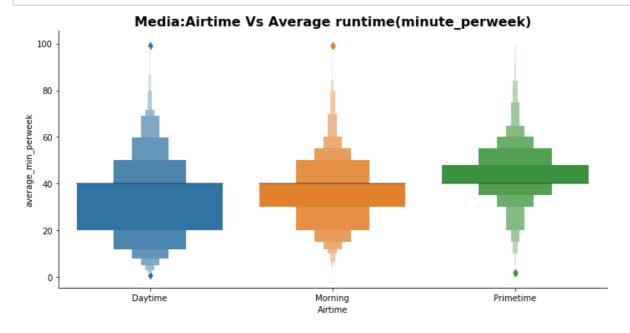
INFERENCE:

Compairing Media and Actual Dataset its clear that media has maniplucated the dataset. In Actual, the target sex and airtime anlysis is direct opposite from Media.

3. AIRTIME VS AVERAGE RUNTIME(minute_perweek)

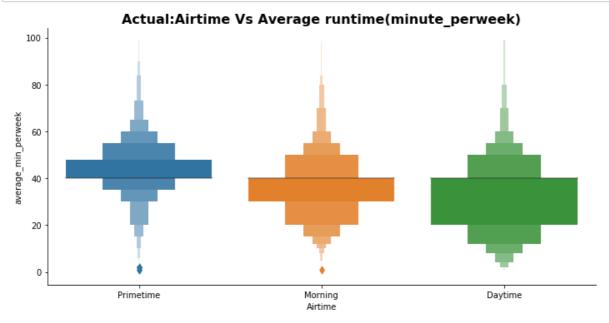
In [12]:

```
sns.catplot(x='Airtime', y='average_min_perweek', data=media, kind='boxen', aspect=2)
plt.title('Media:Airtime Vs Average runtime(minute_perweek)', weight='bold', fontsize=16)
plt.show()
```



In [13]:

```
sns.catplot(x='Airtime', y='average_min_perweek', data=actual, kind='boxen', aspect=2)
plt.title('Actual:Airtime Vs Average runtime(minute_perweek)', weight='bold', fontsize=16)
plt.show()
```

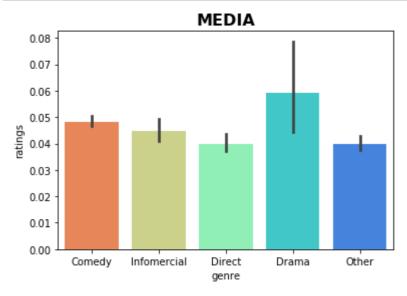


INFERENCE:

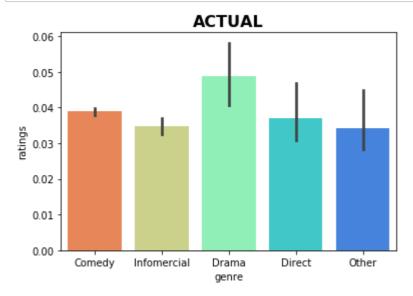
Compairing Media and Actual Dataset there is no manipulcation done.

4. GENRE VS RATINGS

In [14]:



In [15]:

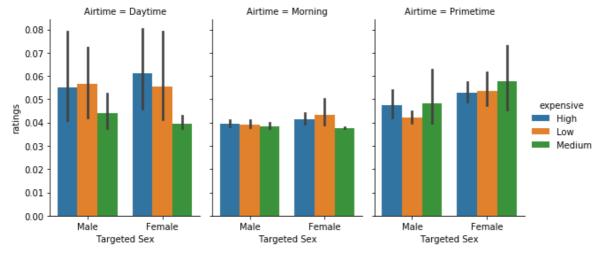


INFERENCE:

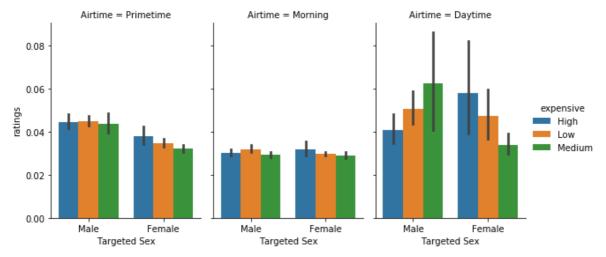
Compairing Media and Actual Dataset its clear that media has maniplucated the dataset. The visualization shows slight diffence in genre between Actual and Media.

5. TARGETED SEX VS EXPENSE

In [16]:



In [17]:



INFERENCE:

Compairing Media and Actual Dataset its clear that media has maniplucated the dataset. There is a drastic change in Expense vs targeted sex when compared between Actual and media.

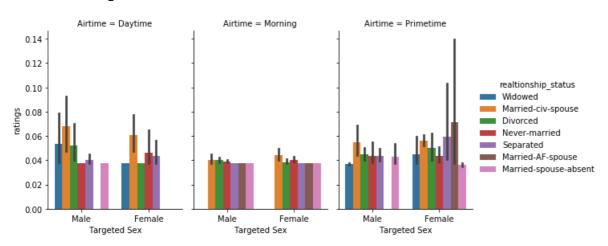
6. TARGETED SEX VS RELATIONSHIP STATUS

In [18]:

MEDIA

Out[18]:

<seaborn.axisgrid.FacetGrid at 0x21946c43c08>

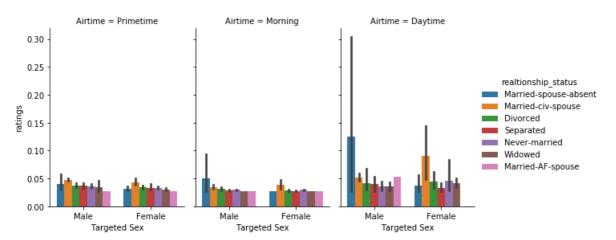


In [19]:

ACTUAL

Out[19]:

<seaborn.axisgrid.FacetGrid at 0x21947326288>



INFERENCE:

Compairing Media and Actual Dataset its clear that media has maniplucated the dataset. There is a drastic change in Relationship status vs targeted sex when compared between Actual and media.