

```
In [1234]: import pandas as pd
import numpy as np

hc = pd.read_csv('Headcount.csv', parse_dates=[0], index_col =0)
nh = pd.read_csv('New_Hire.csv', parse_dates=[0], index_col =0)
```

Headcount

```
In [1235]: #remove white space
hc.rename(columns=lambda x: x.replace(" ", "_"), inplace=True)
nh.rename(columns=lambda x: x.replace(" ", "_"), inplace=True)
```

```
In [1236]: hc.head()
```

Out[1236]:

	Manager_ID	Geography	Region	Country_ID	Work_Location_Name	Sex_Code	Hire_Date	Salary_I
Month ID								
2020-01-01	Management	AP	AU/NZ	Australia	NORWICH	Female	May 1, 2003	
2020-01-01	Non-Management	AP	AU/NZ	Australia	Not Available	Male	Nov 16, 2016	
2020-01-01	Non-Management	AP	AU/NZ	Australia	SOUTHGATE	Male	Jul 1, 2013	
2020-01-01	Non-Management	AP	AU/NZ	Australia	NORWICH	Male	Jun 15, 1981	
2020-01-01	Management	AP	ASEAN	Malaysia	SELANGOR	Male	May 1, 2007	

```
In [1237]: hc.dtypes
```

```
Out[1237]: Manager_ID      object
Geography      object
Region         object
Country_ID     object
Work_Location_Name  object
Sex_Code       object
Hire_Date      object
Salary_Band_Code  object
Salary_Band_Date  object
Local_Position_Date  object
Service_Group   object
Segment_Name    object
Sec_Job_Category_Name  object
Pri_Job_Category_Name  object
Sec_Job_Category_Code  object
Pri_Job_Category_Code  object
Primary_Job_Role  object
Role_/_Specialty  object
Specialty       object
Is_Distinguished_Engineer  object
Is_Fellow       object
Tech_ID         object
Is_Tech_Job_Role  object
URM             object
Diversity_(grouping)  object
Diversity       object
Level_4_Name    object
Level_5_Name    object
Level_6_Name    object
Level_7_Name    object
Level_8_Name    object
Level_9_Name    object
Level_10_Name   object
Hire_Type       object
Is_CIC          object
Salary_Band_Movement_Up  float64
Years_In_Band   float64
Years_In_Previous_Band  float64
Empl_Count      int64
Grouping_Code_1  object
Grouping_Code_2  object
Hire_Year       int64
Salary_Band_Year  object
dtype: object
```

```
In [1238]: nh.head()
```

Out[1238]:

	Country	Geography	Region	Work_Location_Code	Work_Location_Name	Month_ID	BP_CNUM	Status
Row								
1	USA	NaN	US	???	Not Available	2020-January	5G9305897	A
4	USA	NaN	US	BNT	AUSTIN	2020-January	3J2886897	A
16	USA	NaN	US	CF5	IBM CLOUD-DAL11 (USSL1783)	2020-January	3J3641897	A
17	USA	NaN	US	CF5	IBM CLOUD-DAL11 (USSL1783)	2020-January	4J6270897	A
33	USA	NaN	US	HK9	EMERYVILLE-ASPERA	2020-January	4J5897897	A

```
In [1239]: nh.dtypes
```

Out[1239]:

Country	object
Geography	object
Region	object
Work_Location_Code	object
Work_Location_Name	object
Month_ID	object
BP_CNUM	object
Status	object
URM	object
Sex_Short_ID	object
Is_Tech_Job_Role	object
Sec_Job_Category_Code	object
Primary_Job_Role	object
Pri_Job_Category_Name	object
Sec_Job_Category_Name	object
Salary_Band_Code	int64
Movement_Group_1_ID	object
Movement_Group_2_ID	object
Grouping_Code_1	object
Grouping_Code_2	object
dtype:	object

```
In [1240]: nh['Status'].unique().tolist()
```

Out[1240]: ['A']

```
In [1241]: nh = nh.drop(['Status'], axis=1)
```

```
In [1242]: nh.dtypes
```

```
Out[1242]: Country          object
Geography          object
Region             object
Work_Location_Code object
Work_Location_Name object
Month_ID           object
BP_CNUM            object
URM                object
Sex_Short_ID       object
Is_Tech_Job_Role   object
Sec_Job_Category_Code object
Primary_Job_Role   object
Pri_Job_Category_Name object
Sec_Job_Category_Name object
Salary_Band_Code   int64
Movement_Group_1_ID object
Movement_Group_2_ID object
Grouping_Code_1    object
Grouping_Code_2    object
dtype: object
```

```
In [1243]: hc.describe(include='all')
```

Out[1243]:

	Manager_ID	Geography	Region	Country_ID	Work_Location_Name	Sex_Code	Hire_Date	Salary_Band
count	8126	3608	8126	8126	8126	8126	8126	
unique	2	6	20	55	303	2	2840	
top	Non-Management	Europe	US	USA	AUSTIN	Male	Jan 1, 2014	
freq	7464	1981	3975	3975	842	6422	190	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
min	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
25%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
50%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
75%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
max	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

## Data Cleaning

### Join tables

```
In [1244]: df = pd.concat([hc,nh], ignore_index=True, sort=False)
```

```
In [1245]: #another way to combine Country and Country_ID
#df['Country'] = np.where(df['Country'].isna(), df['Country_ID'], df['Country'])
```

```
In [1246]: df.head()
```

Out[1246]:

	Manager_ID	Geography	Region	Country_ID	Work_Location_Name	Sex_Code	Hire_Date	Salary_Band_Cod
0	Management	AP	AU/NZ	Australia	NORWICH	Female	May 1, 2003	1
1	Non-Management	AP	AU/NZ	Australia	Not Available	Male	Nov 16, 2016	1
2	Non-Management	AP	AU/NZ	Australia	SOUTHGATE	Male	Jul 1, 2013	1
3	Non-Management	AP	AU/NZ	Australia	NORWICH	Male	Jun 15, 1981	1
4	Management	AP	ASEAN	Malaysia	SELANGOR	Male	May 1, 2007	1

```
In [1247]: df = df.rename(columns={"Diversity_(grouping)": "Diversity_Grouping"})
```

```
In [1248]: df.head()
```

Out[1248]:

	Manager_ID	Geography	Region	Country_ID	Work_Location_Name	Sex_Code	Hire_Date	Salary_Band_Cod
0	Management	AP	AU/NZ	Australia	NORWICH	Female	May 1, 2003	1
1	Non-Management	AP	AU/NZ	Australia	Not Available	Male	Nov 16, 2016	1
2	Non-Management	AP	AU/NZ	Australia	SOUTHGATE	Male	Jul 1, 2013	1
3	Non-Management	AP	AU/NZ	Australia	NORWICH	Male	Jun 15, 1981	1
4	Management	AP	ASEAN	Malaysia	SELANGOR	Male	May 1, 2007	1

```
In [1249]: df['Country']=df.Country.combine_first(df.Country_ID)
```

```
In [1250]: #drop Country_ID
df = df.drop(['Country_ID'], axis=1)
```

```
In [1251]: #drop Region, since it is redundant from COUNTRY
df = df.drop(['Region'], axis=1)
```

```
In [1252]: #combine into SEX column
df['Sex']=df.Sex_Code.combine_first(df.Sex_Short_ID)
```

```
In [1253]: #drop old Sex columns
df = df.drop(['Sex_Code'], axis=1)
df = df.drop(['Sex_Short_ID'], axis=1)

In [1254]: #convert 'Hire Date' to month/year
df['Hire_Date'] = pd.to_datetime(df['Hire_Date']).dt.strftime('%m/%Y')

In [1255]: #combine 'Hire Date' and 'Month_ID'
df['Hire_Date']=df.Hire_Date.combine_first(df.Month_ID)

In [1256]: #drop Month_ID
df = df.drop(['Month_ID'], axis=1)

In [1257]: #rename 'Movement Group 2 ID' to Hire_Type
df['Hire_Type']=df.Hire_Type.combine_first(df.Movement_Group_2_ID)

In [1258]: #drop old Movement Group 2 ID column
df = df.drop(['Movement_Group_2_ID'], axis=1)

In [1259]: #remove 'Salary_Band_Movement_Up' column due to value only being 1 or NaN
df = df.drop(['Salary_Band_Movement_Up'], axis=1)

In [1260]: #drop Diversity_grouping column since it is represented by Sex and URM
df = df.drop(['Diversity_Grouping'], axis=1)

In [1261]: df.head()
```

Out[1261]:

	Manager_ID	Geography	Work_Location_Name	Hire_Date	Salary_Band_Code	Salary_Band_Date	Local_Pos
0	Management	AP	NORWICH	05/2003	10	Jul 1, 2009	O
1	Non-Management	AP	Not Available	11/2016	10	Nov 16, 2016	Au
2	Non-Management	AP	SOUTHGATE	07/2013	10	Jun 1, 2018	Se
3	Non-Management	AP	NORWICH	06/1981	10	May 1, 2000	O
4	Management	AP	SELANGOR	05/2007	10	Nov 1, 2014	M

```

In [1262]: #clean null values
def assess_NA(data):
    """
    Returns a pandas dataframe denoting the total number of NA values and the per
    centage of NA values in each column.
    The column names are noted on the index.

    Parameters
    -----
    data: dataframe
    """
    # pandas series denoting features and the sum of their null values
    null_sum = data.isnull().sum()# instantiate columns for missing data
    total = null_sum.sort_values(ascending=False)
    percent = ( (null_sum / len(data.index))*100).round(2) ).sort_values(ascendi
ng=False)

    # concatenate along the columns to create the complete dataframe
    df_NA = pd.concat([total, percent], axis=1, keys=['Number of NA', 'Percent NA
'])

    # drop rows that don't have any missing data; omit if you want to keep all ro
ws
    df_NA = df_NA[ (df_NA.T != 0).any() ]

    return df_NA

```

```
In [1263]: #show frequency and percentage of nulls
df_NA = assess_NA(df)
df_NA
```

Out[1263]:

	Number of NA	Percent NA
Grouping_Code_2	8189	99.94
BP_CNUM	8126	99.17
Work_Location_Code	8126	99.17
Movement_Group_1_ID	8126	99.17
Grouping_Code_1	4857	59.28
Geography	4545	55.47
Years_In_Previous_Band	2298	28.04
Hire_Type	922	11.25
Specialty	243	2.97
Salary_Band_Date	71	0.87
Years_In_Band	71	0.87
Local_Position_Date	70	0.85
Role_/_Specialty	68	0.83
Is_Fellow	68	0.83
Is_Distinguished_Engineer	68	0.83
Diversity	68	0.83
Pri_Job_Category_Code	68	0.83
Segment_Name	68	0.83
Service_Group	68	0.83
Tech_ID	68	0.83
Manager_ID	68	0.83
Level_4_Name	68	0.83
Salary_Band_Year	68	0.83
Level_5_Name	68	0.83
Level_6_Name	68	0.83
Level_7_Name	68	0.83
Level_8_Name	68	0.83
Level_9_Name	68	0.83
Level_10_Name	68	0.83
Is_CIC	68	0.83
Empl_Count	68	0.83
Hire_Year	68	0.83
Primary_Job_Role	8	0.10



```
In [1264]: #should drop columns: 'Grouping_Code_2', 'Movement_Group_1_ID', 'Status', 'BP_CNUM', 'Work_Location_Code'
df = df.drop(['Grouping_Code_2', 'Movement_Group_1_ID', 'BP_CNUM', 'Work_Location_Code'], axis=1)
```

```
In [1265]: df.head()
```

Out[1265]:

	Manager_ID	Geography	Work_Location_Name	Hire_Date	Salary_Band_Code	Salary_Band_Date	Local_Pos
0	Management	AP	NORWICH	05/2003	10	Jul 1, 2009	O
1	Non-Management	AP	Not Available	11/2016	10	Nov 16, 2016	Au
2	Non-Management	AP	SOUTHGATE	07/2013	10	Jun 1, 2018	Se
3	Non-Management	AP	NORWICH	06/1981	10	May 1, 2000	O
4	Management	AP	SELANGOR	05/2007	10	Nov 1, 2014	M

```
In [1266]: df_NA = assess_NA(df)
df_NA
```

Out[1266]:

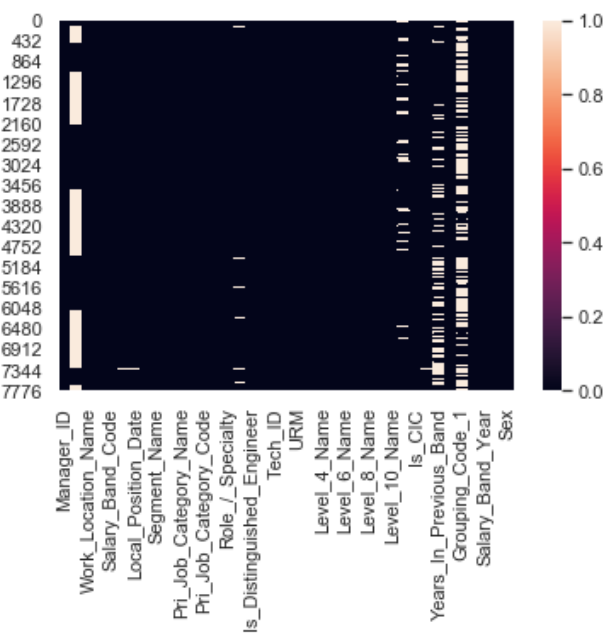
	Number of NA	Percent NA
Grouping_Code_1	4857	59.28
Geography	4545	55.47
Years_In_Previous_Band	2298	28.04
Hire_Type	922	11.25
Specialty	243	2.97
Salary_Band_Date	71	0.87
Years_In_Band	71	0.87
Local_Position_Date	70	0.85
Manager_ID	68	0.83
Segment_Name	68	0.83
Pri_Job_Category_Code	68	0.83
Role_/_Specialty	68	0.83
Is_Distinguished_Engineer	68	0.83
Is_Fellow	68	0.83
Tech_ID	68	0.83
Diversity	68	0.83
Level_4_Name	68	0.83
Level_5_Name	68	0.83
Level_6_Name	68	0.83
Level_7_Name	68	0.83
Level_8_Name	68	0.83
Level_9_Name	68	0.83
Level_10_Name	68	0.83
Is_CIC	68	0.83
Empl_Count	68	0.83
Hire_Year	68	0.83
Salary_Band_Year	68	0.83
Service_Group	68	0.83
Primary_Job_Role	8	0.10

Data Exploration

```
In [1267]: import matplotlib.pyplot as plt
plt.rc("font", size=14)
import seaborn as sns
sns.set(style="white")
sns.set(style="whitegrid", color_codes=True)
```

```
In [1268]: sns.heatmap(df.isnull())
```

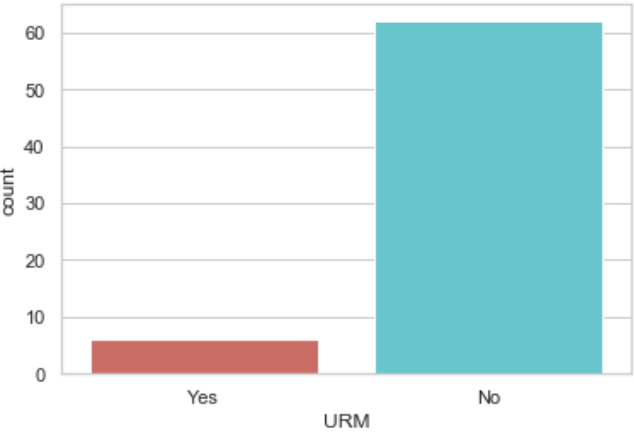
```
Out[1268]: <matplotlib.axes._subplots.AxesSubplot at 0x7ff9f702c510>
```



URM

New Hire

```
In [1269]: sns.countplot(x='URM', data=nh, palette='hls')
plt.show()
```



```
In [1270]: nh.URM.value_counts()
```

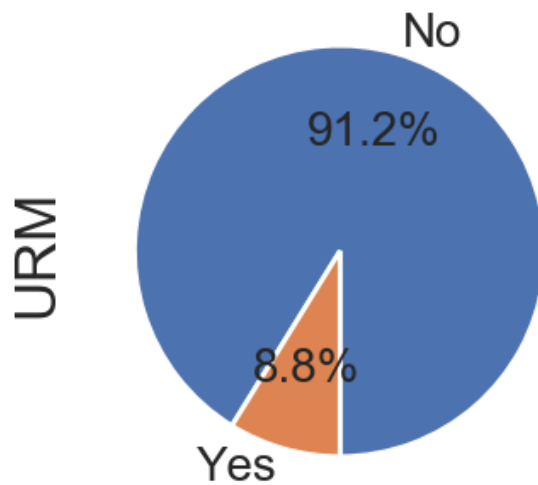
```
Out[1270]: No      62
           Yes      6
           Name: URM, dtype: int64
```

```
In [1271]: fig = plt.figure(figsize=(2,2), dpi=200)
           ax = plt.subplot(111)

           nh.URM.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270, fo
           ntsize=10)
           plt.title('Number of URM in New Hires')
```

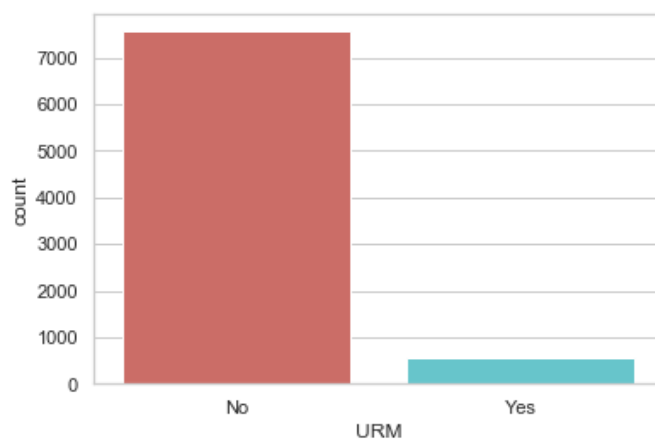
```
Out[1271]: Text(0.5, 1.0, 'Number of URM in New Hires')
```

## Number of URM in New Hires



## Head Count

```
In [1272]: sns.countplot(x='URM', data=hc, palette='hls')
           plt.show()
```



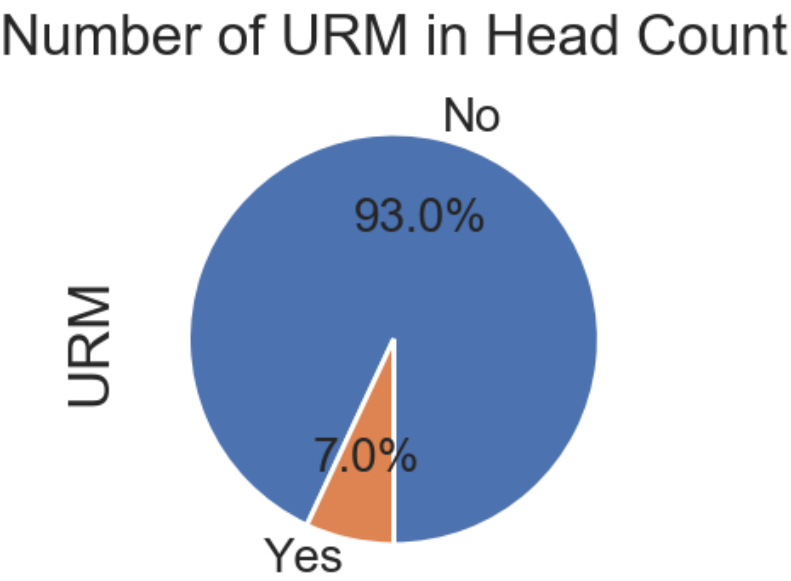
```
In [1273]: hc.URM.value_counts()
```

```
Out[1273]: No      7561
           Yes      565
           Name: URM, dtype: int64
```

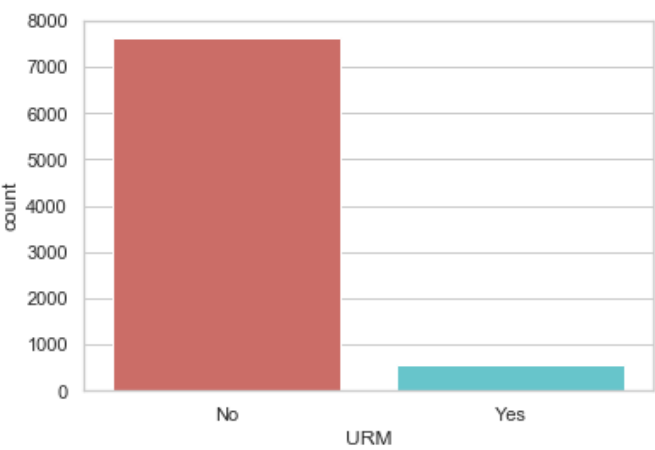
```
In [1274]: fig = plt.figure(figsize=(2,2), dpi=200)
           ax = plt.subplot(111)

           hc.URM.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270, fo
           ntsize=10)
           plt.title('Number of URM in Head Count')
```

Out[1274]: Text(0.5, 1.0, 'Number of URM in Head Count')



```
In [1275]: sns.countplot(x='URM', data=df, palette='hls')
           plt.show()
```



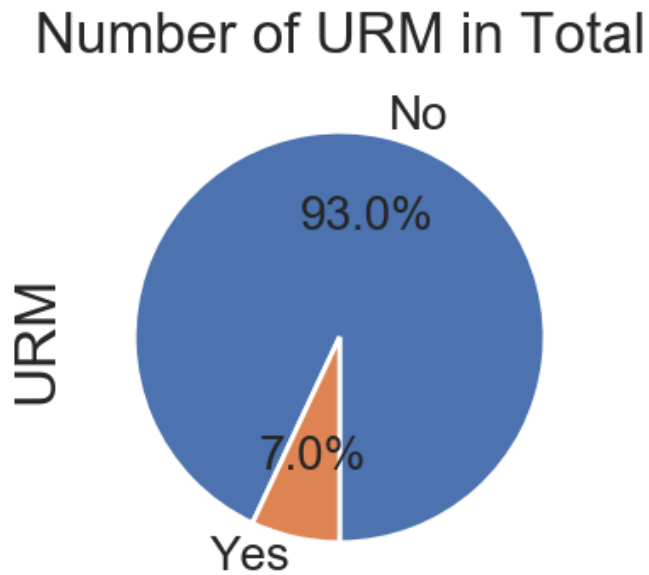
```
In [1276]: df.URM.value_counts()
```

Out[1276]: No 7623  
 Yes 571  
 Name: URM, dtype: int64

```
In [1277]: fig = plt.figure(figsize=(2,2), dpi=200)
ax = plt.subplot(111)

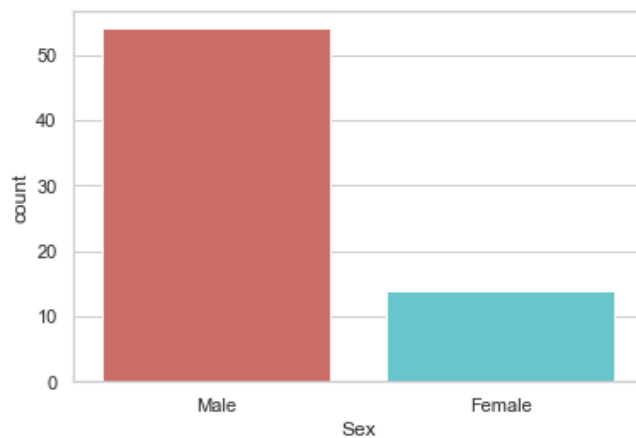
df.URM.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270, fontsize=10)
plt.title('Number of URM in Total')
```

```
Out[1277]: Text(0.5, 1.0, 'Number of URM in Total')
```



## Sex

```
In [1278]: sns.countplot(x='Sex_Short_ID', data=nh, palette='hls')
plt.xlabel('Sex')
plt.show()
```



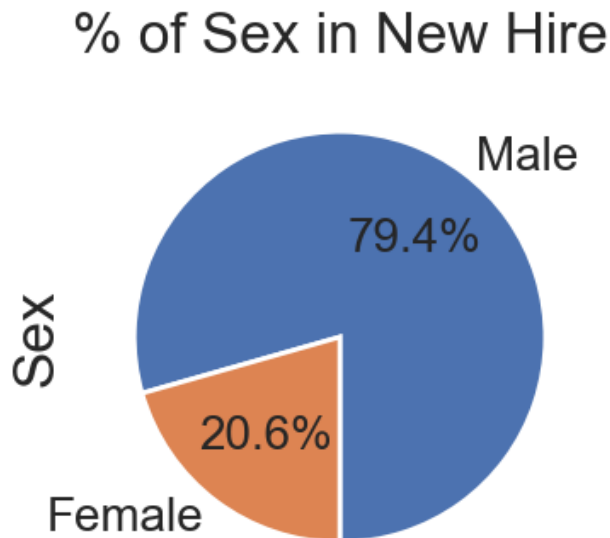
```
In [1279]: nh.Sex_Short_ID.value_counts()
```

```
Out[1279]: Male      54
Female    14
Name: Sex_Short_ID, dtype: int64
```

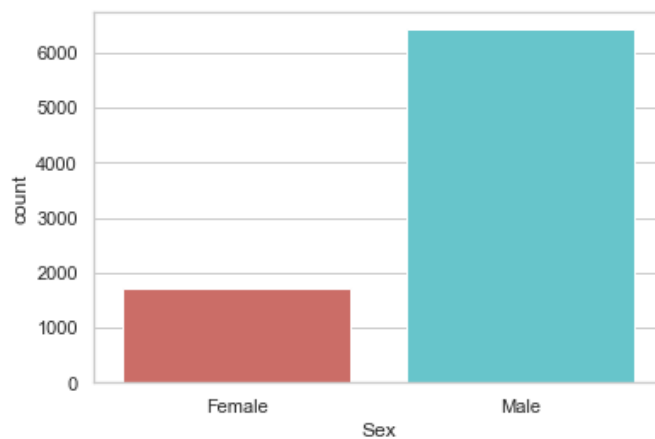
```
In [1280]: fig = plt.figure(figsize=(2,2), dpi=200)
ax = plt.subplot(111)

nh.Sex_Short_ID.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270, fontsize=10)
plt.title('% of Sex in New Hire')
plt.ylabel('Sex')
```

Out[1280]: Text(0, 0.5, 'Sex')



```
In [1281]: sns.countplot(x='Sex_Code', data=hc, palette='hls')
plt.xlabel('Sex')
plt.show()
```



```
In [1282]: hc.Sex_Code.value_counts()
```

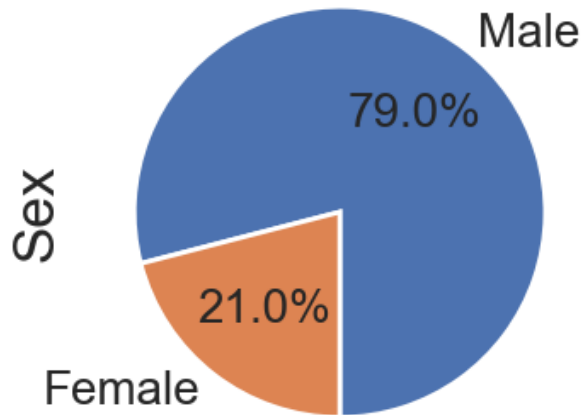
Out[1282]: Male 6422  
Female 1704  
Name: Sex\_Code, dtype: int64

```
In [1283]: fig = plt.figure(figsize=(2,2), dpi=200)
           ax = plt.subplot(111)

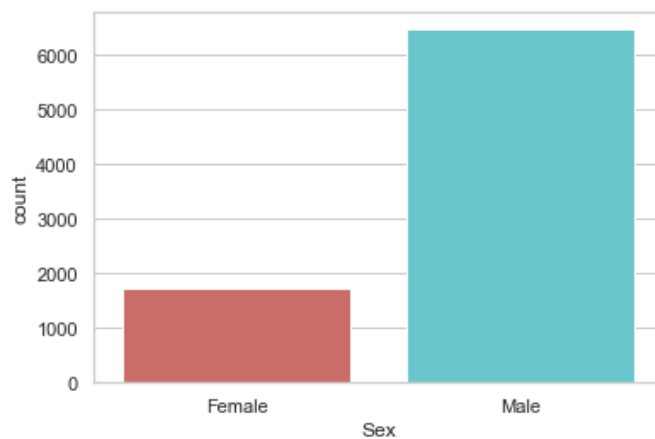
           hc.Sex_Code.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270,
           fontsize=10)
           plt.title('% of Sex in Head Count')
           plt.ylabel('Sex')
```

```
Out[1283]: Text(0, 0.5, 'Sex')
```

## % of Sex in Head Count



```
In [1284]: sns.countplot(x='Sex', data=df, palette='hls')
           plt.xlabel('Sex')
           plt.show()
```



```
In [1285]: df.Sex.value_counts()
```

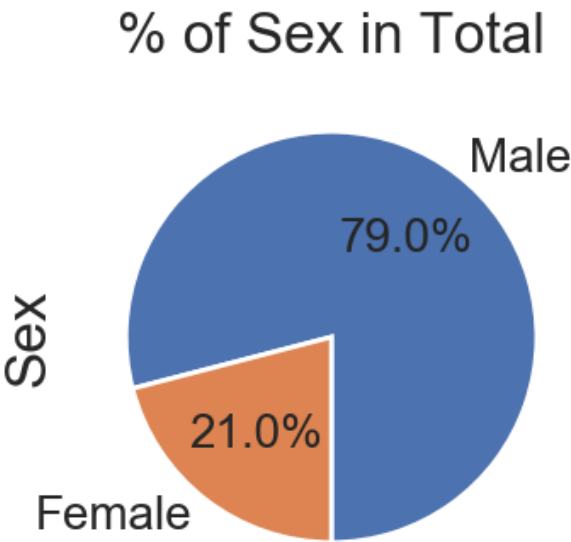
```
Out[1285]: Male      6476
           Female    1718
           Name: Sex, dtype: int64
```



```
In [1286]: fig = plt.figure(figsize=(2,2), dpi=200)
ax = plt.subplot(111)

df.Sex.value_counts().plot(kind='pie',ax=ax, autopct='%1.1f%%',startangle=270, fo
ntsize=10)
plt.title('% of Sex in Total')
plt.ylabel('Sex')
```

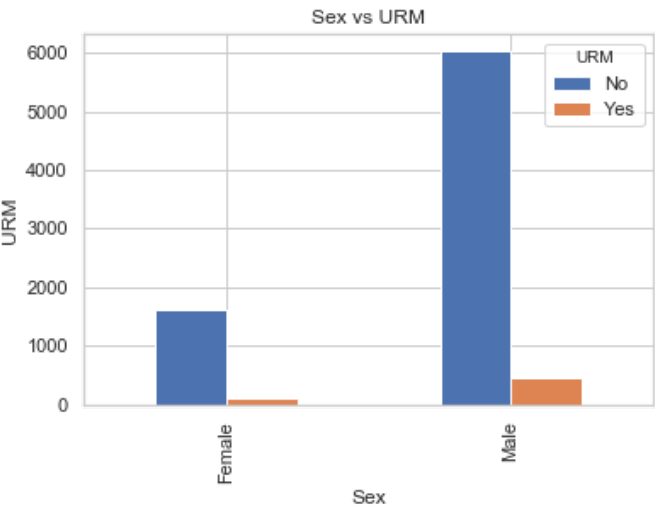
Out[1286]: Text(0, 0.5, 'Sex')



URM vs Sex

```
In [1287]: %matplotlib inline
pd.crosstab(df.Sex,df.URM).plot(kind='bar')
plt.title('Sex vs URM')
plt.xlabel('Sex')
plt.ylabel('URM')
```

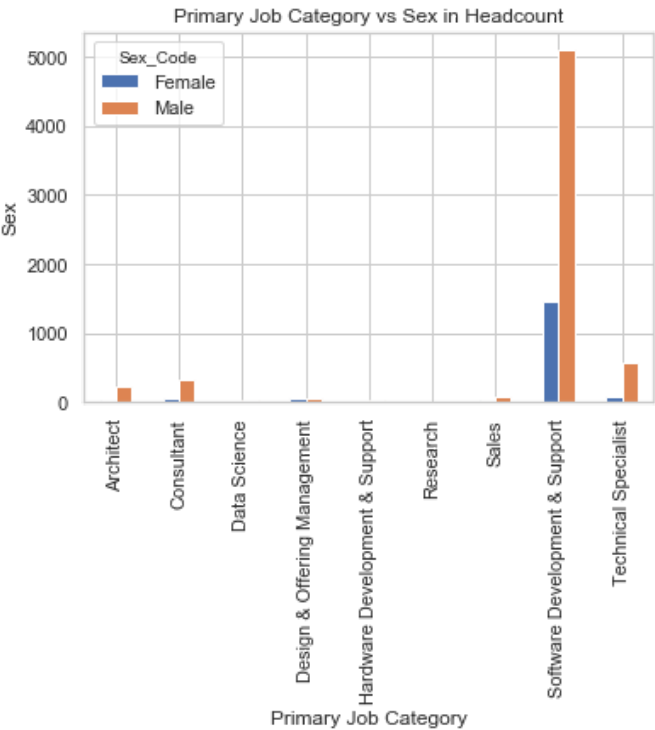
Out[1287]: Text(0, 0.5, 'URM')



Primary Job Category vs Sex

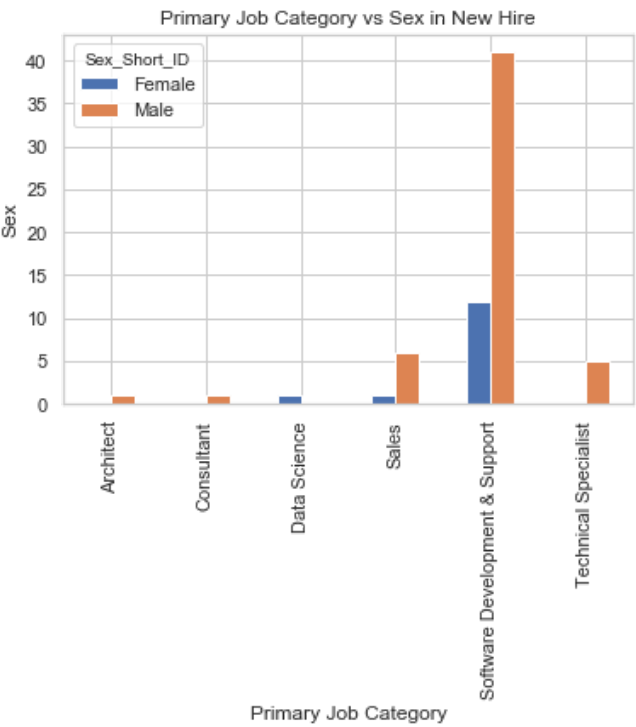
```
In [1288]: %matplotlib inline
pd.crosstab(hc.Pri_Job_Category_Name,hc.Sex_Code).plot(kind='bar')
plt.title('Primary Job Category vs Sex in Headcount')
plt.xlabel('Primary Job Category')
plt.ylabel('Sex')
```

Out[1288]: Text(0, 0.5, 'Sex')



```
In [1289]: %matplotlib inline
pd.crosstab(nh.Pri_Job_Category_Name,nh.Sex_Short_ID).plot(kind='bar')
plt.title('Primary Job Category vs Sex in New Hire')
plt.xlabel('Primary Job Category')
plt.ylabel('Sex')
```

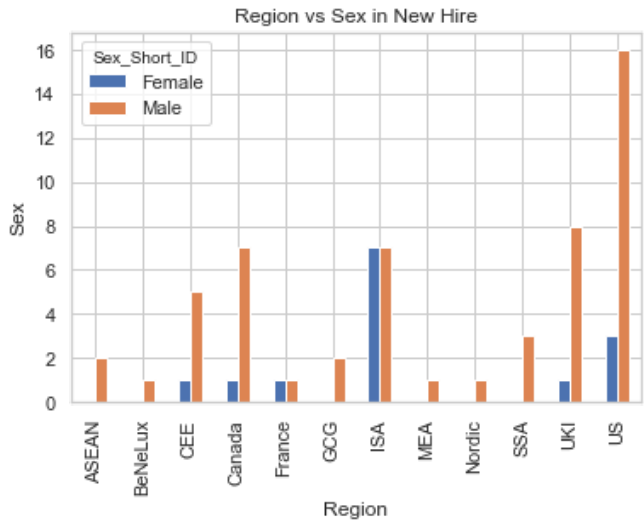
Out[1289]: Text(0, 0.5, 'Sex')



Region vs Sex

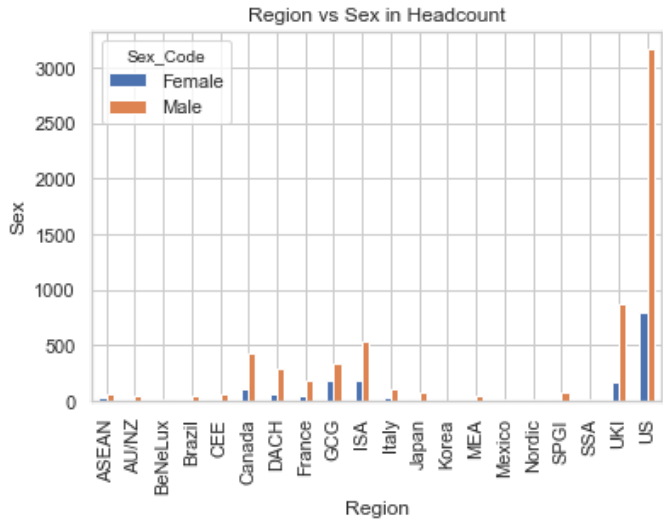
```
In [1290]: #new hire
%matplotlib inline
pd.crosstab(nh.Region,nh.Sex_Short_ID).plot(kind='bar')
plt.title('Region vs Sex in New Hire')
plt.xlabel('Region')
plt.ylabel('Sex')
```

Out[1290]: Text(0, 0.5, 'Sex')



```
In [1291]: #head count
%matplotlib inline
pd.crosstab(hc.Region,hc.Sex_Code).plot(kind='bar')
plt.title('Region vs Sex in Headcount')
plt.xlabel('Region')
plt.ylabel('Sex')
```

Out[1291]: Text(0, 0.5, 'Sex')



Hire Date vs Sex

```
In [1298]: #head count
%matplotlib inline
pd.crosstab(hc.Hire_Date, hc.Sex_Code).plot(kind='line', figsize=(10,10))
plt.title('Hire Date vs Sex in Headcount')
plt.xlabel('Hire Date')
plt.ylabel('Sex')
```

Out[1298]: Text(0, 0.5, 'Sex')

