Clustering NBA Teams

```
In [68]: import pandas as pd
    import numpy as np
    import plotly.express as px
    import pickle
    from sklearn.preprocessing import MinMaxScaler
    from sklearn.decomposition import PCA
    from sklearn.cluster import KMeans
    from plotly.offline import init_notebook_mode
    init_notebook_mode(connected = True)
    import plotly.io as pio
    pio.renderers.default = "svg"
```

Importing Dataset

In [72]: team_stats

Out[72]:

	WIN%	PTS	FGM	FGA	FG%	3РМ	3PA	3P%	FTM	FTA	 AST\nRATIO	OREB%	DRE
0	0.341	94.3	36.6	83.0	44.1	3.1	9.9	31.7	18.0	24.2	 14.6	33.4	
1	0.427	99.3	37.2	83.9	44.4	5.1	15.4	33.1	19.8	26.5	 15.9	32.7	
2	0.598	98.4	35.8	79.7	44.9	4.1	12.2	33.9	22.7	30.0	 18.4	28.3	
3	0.207	84.8	31.3	75.4	41.5	4.1	12.6	32.9	18.1	25.5	 15.8	33.1	
4	0.390	97.0	36.3	82.1	44.2	4.2	11.2	37.3	20.2	26.9	 17.3	30.8	
620	0.431	110.1	40.9	88.4	46.2	12.7	34.9	36.4	15.7	20.3	 17.4	25.8	
621	0.451	114.1	42.2	89.4	47.2	10.7	28.5	37.6	19.0	23.4	 17.9	23.3	
622	0.736	112.8	40.2	87.9	45.8	13.8	37.0	37.4	18.5	23.2	 18.1	25.6	
623	0.611	111.3	40.1	85.1	47.1	13.4	35.2	38.0	17.8	22.8	 16.8	26.1	
624	0.347	114.4	41.5	90.9	45.7	12.0	32.6	36.8	19.4	24.6	 17.6	26.2	

625 rows × 37 columns

Normalizing Team Stats Data

```
In [73]: scaler = MinMaxScaler()
    scaler.fit(team_stats)
    normalized_team_stats=scaler.transform(team_stats)
    normalized_team_stats=pd.DataFrame(normalized_team_stats)
    normalized_team_stats.columns=team_stats.columns
    normalized_team_stats
```

Out[73]:

	WIN%	PTS	FGM	FGA	FG%	3РМ	3PA	3P%	FTM	
0	0.299745	0.292754	0.417323	0.441624	0.343750	0.044118	0.055851	0.282609	0.487395	0.5
1	0.409439	0.437681	0.464567	0.487310	0.375000	0.191176	0.202128	0.384058	0.638655	0.6
2	0.627551	0.411594	0.354331	0.274112	0.427083	0.117647	0.117021	0.442029	0.882353	8.0
3	0.128827	0.017391	0.000000	0.055838	0.072917	0.117647	0.127660	0.369565	0.495798	0.5
4	0.362245	0.371014	0.393701	0.395939	0.354167	0.125000	0.090426	0.688406	0.672269	0.6
620	0.414541	0.750725	0.755906	0.715736	0.562500	0.750000	0.720745	0.623188	0.294118	0.2
621	0.440051	0.866667	0.858268	0.766497	0.666667	0.602941	0.550532	0.710145	0.571429	0.4
622	0.803571	0.828986	0.700787	0.690355	0.520833	0.830882	0.776596	0.695652	0.529412	0.4
623	0.644133	0.785507	0.692913	0.548223	0.656250	0.801471	0.728723	0.739130	0.470588	0.4
624	0.307398	0.875362	0.803150	0.842640	0.510417	0.698529	0.659574	0.652174	0.605042	0.5

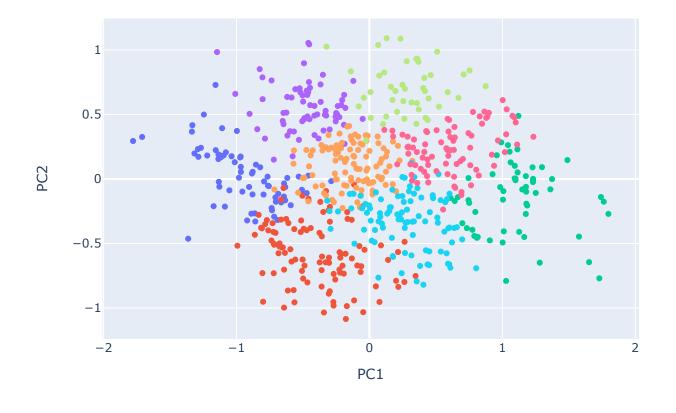
625 rows × 37 columns

Function to Visualize NBA Team Clusters

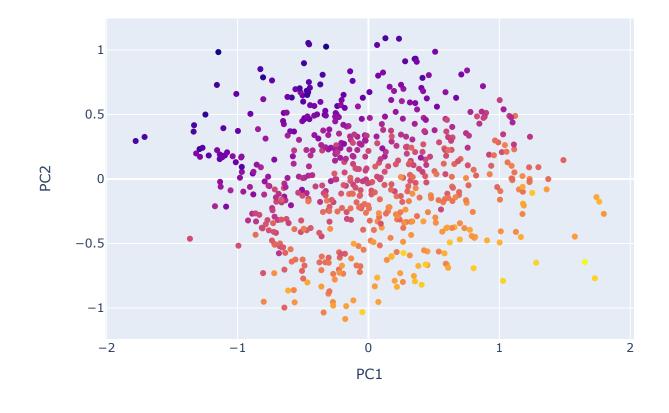
```
""" Visualize KMeans Clustering of NBA Teams
In [74]:
         Args:
             normalized df: normalized team stats df to cluster
             teams: teams df
             stats: team statistics df
             components: number of components to visualize 2 or 3
             color: what feature color scatter points by
             hover data: data to see on plotly scatterplot hover
         Returns:
             pca: dataframe with principal components
             knn: Kmeans object trained on data
         Raises:
             Exception(): Invalid input
         . . . .
         def visualize KMeans clustering(normalized df,teams,stats,k,components,colo
             if components!=2 and components!=3:
                 raise Exception("Invalid component: '{}' - Must select 2 or 3".form
             # 2 principal components
             if components==2:
                 # fitting clusters on normalized data
                 knn=KMeans(n_clusters=k,random_state=0)
                 df fit=normalized df.copy()
                 # adding more weight to winning percentage
                 df fit['WIN%']=df fit['WIN%']*2.5
                 knn.fit(df fit)
                 # dimensionality reduction w/ PCA
                 pca = PCA(n components=2)
                 components = pca.fit transform(normalized df)
                 pca=pd.DataFrame(data=components, columns=["PC1", "PC2"])
                 pca['TEAM']=teams['TEAM']+" "+teams['YEAR']
                 pca['Clusters']=knn.labels
                 pca['Clusters']=pca['Clusters'].astype('0')
                 pca=pd.concat([pca,stats],axis=1)
                 # visualizing w/ plotly
                 fig = px.scatter(pca, x='PC1', y='PC2',color=color,hover data=hover
                 fig.show(renderer="svg")
```

```
# returning df and trained object
    if output!=False:
        return pca, knn
else:
    # same as above but 3 principal components
    knn=KMeans(n clusters=k,random state=0)
    df_fit=normalized_df.copy()
    df_fit['WIN%']=df_fit['WIN%']*2
    knn.fit(df_fit)
    pca = PCA(n_components=3)
    components = pca.fit transform(normalized df)
    total_var = pca.explained_variance_ratio_.sum() * 100
    pca=pd.DataFrame(data=components, columns=["PC1", "PC2","PC3"])
    pca['TEAM']=teams['TEAM']+" "+teams['YEAR']
    pca['Clusters']=knn.labels_
    pca['Clusters']=pca['Clusters'].astype('0')
    pca=pd.concat([pca,stats],axis=1)
    fig = px.scatter 3d(pca, x='PC1', y='PC2', z='PC3',color=color,hove
    fig.show(renderer="svg")
    if output!=False:
        return pca, knn
```

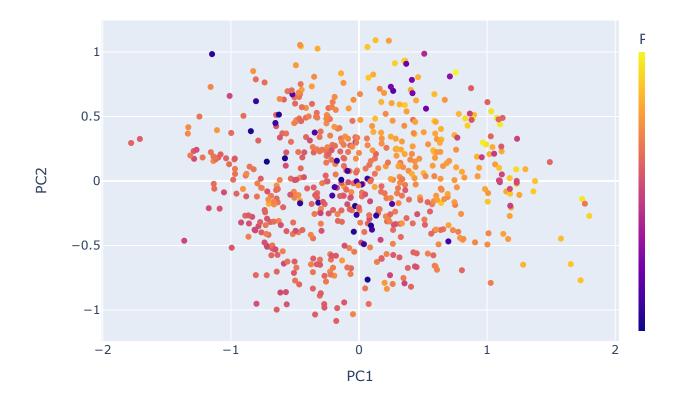
In [75]: df,knn=visualize_KMeans_clustering(normalized_team_stats,teams,team_stats,8



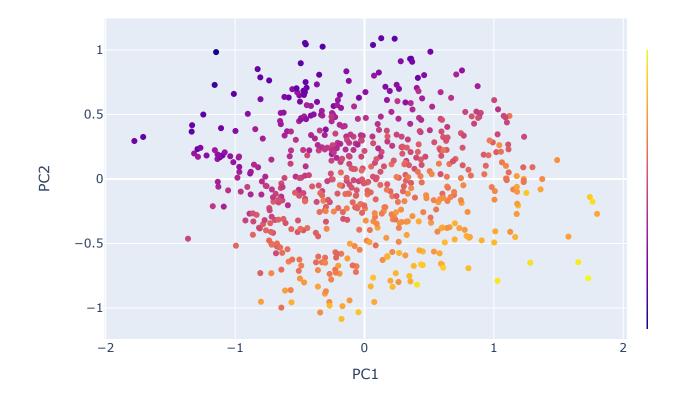
In [76]: visualize_KMeans_clustering(normalized_team_stats,teams,team_stats,8,2,colo



In [77]: visualize_KMeans_clustering(normalized_team_stats,teams,team_stats,8,2,colo

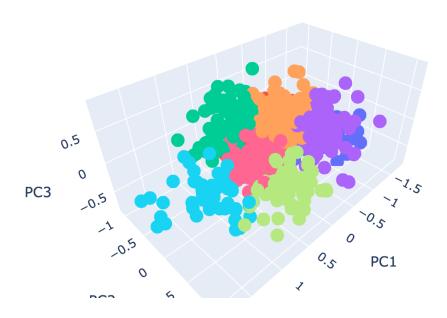


In [78]: visualize_KMeans_clustering(normalized_team_stats,teams,team_stats,8,2,colo



In [79]: visualize_KMeans_clustering(normalized_team_stats,teams,team_stats,8,3,colo

Total Explained Variance: 60.61%



Saving KMeans Object for Python Module

```
In [39]: pickle.dump(knn, open( "knn.pkl", "wb" ))
In [170]: # testing object
kmeans_object=pd.read_pickle('knn.pkl')
```

In [179]: # predicting cluster sample team data

```
normalized_team_stats.iloc[6]
Out[179]: WIN%
                          0.409439
           PTS
                          0.428986
           FGM
                          0.472441
           FGA
                          0.507614
           FG%
                          0.354167
           3PM
                          0.235294
           3PA
                          0.244681
           3P%
                          0.420290
           FTM
                          0.546218
           FTA
                          0.617450
           FT%
                          0.367816
           OREB
                          0.625000
           DREB
                          0.307190
           REB
                          0.527027
           AST
                          0.453846
           TOV
                          0.564103
           STL
                          0.270833
           BLK
                          0.879310
                          0.666667
           BLKA
           PF
                          0.695238
           PFD
                          0.097276
           +/-
                          0.462745
           OFFRTG
                          0.445344
           DEFRTG
                          0.481013
                          0.481061
           NETRTG
           AST%
                          0.600000
           AST/TO
                          0.432432
           AST\nRATIO
                          0.436620
           OREB%
                          0.668790
           DREB%
                          0.320312
           REB%
                          0.579545
           TOV%
                          0.471910
           EFG%
                          0.340426
           TS%
                          0.358209
           PACE
                          0.472406
           PIE
                          0.423529
           POSS
                          0.743194
           Name: 6, dtype: float64
In [180]: kmeans_object.predict(np.asarray(normalized_team_stats.iloc[6]).reshape(1,
Out[180]: array([5], dtype=int32)
```

```
localhost:8889/notebooks/Clustering.ipynb
```