

Appendix 11: Hypothesis 3 Test

In [29]:

```
import pandas as pd
import numpy as np
import seaborn as sns
from sklearn.preprocessing import MinMaxScaler
%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt
import statsmodels.api as sm
from scipy import stats

from statsmodels.formula.api import ols
```

In [30]:

```
df=pd.read_csv('data_final.csv')
df.head()
```

Out[30]:

	Unnamed: 0	track	artist	uri	danceability	energy
0	0	Wild Things	Alessia Cara	spotify:track:2ZyuwVvV6Z3XJaXIFbspeE	0.741	0.626
1	1	Love Someone	Lukas Graham	spotify:track:2JqnpxIO9dmvjUMCaLCLJ	0.550	0.415
2	2	Here's To Never Growing Up	Avril Lavigne	spotify:track:0qwcGscxUHGZTgq0zcaqk1	0.482	0.875
3	3	Crawling Back To You	Daughtry	spotify:track:6BDtTzjbJ5kKKSWcJT8MIX	0.438	0.915
4	4	Faster	Matt Nathanson	spotify:track:6pIKFdrBnKF0y3CRuceTDh	0.742	0.855

5 rows x 32 columns



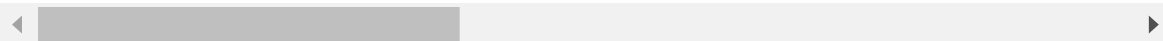
In [31]:

```
df=df.iloc[:,1:]
df.head()
```

Out[31]:

	track	artist	uri	danceability	energy	key	loudness
0	Wild Things	Alessia Cara	spotify:track:2ZyuwVvV6Z3XJaXIFbspeE	0.741	0.626	1	-1.2
1	Love Someone	Lukas Graham	spotify:track:2JqnpexlO9dmvjUMCaLCLJ	0.550	0.415	9	-1.2
2	Here's To Never Growing Up	Avril Lavigne	spotify:track:0qwcGscxUHGZTgq0zcaqk1	0.482	0.873	0	-1.2
3	Crawling Back To You	Daughtry	spotify:track:6BDtTzjbJ5kKKSWCJT8MIX	0.438	0.919	0	-1.2
4	Faster	Matt Nathanson	spotify:track:6pIKFdrBnKF0y3CRuceTDh	0.742	0.853	9	-1.2

5 rows × 31 columns



In [32]:

```
genres=df['genres'].unique()
decades=df['Decade'].unique()
print(genres)
print(decades)
```

```
['Pop' 'Easy listening' 'Hip hop' 'Metal' 'Country' 'Electronic' 'Rock'
 'Folk' 'Latin' 'Classical' 'Jazz' 'R&B' 'Caribbean' 'Blues']
['10s' '00s' '90s' '80s' '70s' '60s']
```

In [33]:

```
tb = [[0 for x in range(6)] for y in range(len(df))]
len(tb)
```

Out[33]:

33354

In [34]:

```
df_1=df
y=df_1.iloc[:,[25,26]]
x=df_1.iloc[:,[3,4,10,12]]
norm = MinMaxScaler().fit(x)
x_norm = norm.transform(x)
x=pd.DataFrame(data=x_norm,columns=x.columns)
x.head()
data=pd.concat([x,y],axis=1)
data.head()
```

Out[34]:

	danceability	energy	instrumentalness	valence	genres	Decade
0	0.734180	0.625906	0.000000	0.710977	Pop	10s
1	0.528627	0.414853	0.000000	0.275932	Pop	10s
2	0.455446	0.872968	0.000000	0.742195	Pop	10s
3	0.408093	0.918980	0.000000	0.196375	Pop	10s
4	0.735256	0.852963	0.000005	0.956697	Pop	10s

In [35]:

```
for index,row in data.iterrows():
    tb[index][0]=row['genres']
    tb[index][1]=row['Decade']
    tb[index][2]=row['danceability']
    tb[index][3]=row['energy']
    tb[index][4]=row['instrumentalness']
    tb[index][5]=row['valence']
```

In [36]:

```
tb[:5]
```

Out[36]:

```
[['Pop',
  '10s',
  0.7341799397331037,
  0.6259061024317104,
  0.0,
  0.7109768378650553],
 ['Pop',
  '10s',
  0.5286267757210504,
  0.41485312813516195,
  0.0,
  0.27593152064451154],
 ['Pop',
  '10s',
  0.4554455445544555,
  0.8729681149968643,
  0.0,
  0.7421953675730111],
 ['Pop',
  '10s',
  0.40809298321136456,
  0.9189796638956379,
  0.0,
  0.19637462235649547],
 ['Pop',
  '10s',
  0.735256134309083,
  0.8529630937365279,
  4.79e-06,
  0.9566968781470291]]
```

In [38]:

```
x=['Genre', 'Decade', 'Danceability', 'Energy', 'Instrumentalness', 'Valence']
df_tb=pd.DataFrame(data=tb,columns=x)
df_tb.head()
```

Out[38]:

	Genre	Decade	Danceability	Energy	Instrumentalness	Valence
0	Pop	10s	0.734180	0.625906	0.000000	0.710977
1	Pop	10s	0.528627	0.414853	0.000000	0.275932
2	Pop	10s	0.455446	0.872968	0.000000	0.742195
3	Pop	10s	0.408093	0.918980	0.000000	0.196375
4	Pop	10s	0.735256	0.852963	0.000005	0.956697

In [39]:

```
d_melt=df_tb.iloc[:, :3]
d_melt.head()
```

Out[39]:

	Genre	Decade	Danceability
0	Pop	10s	0.734180
1	Pop	10s	0.528627
2	Pop	10s	0.455446
3	Pop	10s	0.408093
4	Pop	10s	0.735256

In [41]:

```
model = ols('Danceability ~ C(Genre) + C(Decade) + C(Genre):C(Decade)', data=d_melt).fit()
anova_table = sm.stats.anova_lm(model, typ=2)
anova_table
```

Out[41]:

	sum_sq	df	F	PR(>F)
C(Genre)	335.326773	13.0	1080.019062	0.000000e+00
C(Decade)	27.072615	5.0	226.707950	8.732465e-239
C(Genre):C(Decade)	32.805159	65.0	21.131743	7.966763e-239
Residual	794.595785	33270.0	NaN	NaN

In [43]:

```
d_melt=df_tb.iloc[:, [0,1,3]]
model = ols('Energy ~ C(Genre) + C(Decade) + C(Genre):C(Decade)', data=d_melt).fit()
anova_table = sm.stats.anova_lm(model, typ=2)
anova_table
```

Out[43]:

	sum_sq	df	F	PR(>F)
C(Genre)	488.358079	13.0	999.265824	0.0
C(Decade)	101.612683	5.0	540.584918	0.0
C(Genre):C(Decade)	72.445538	65.0	29.647242	0.0
Residual	1250.739285	33270.0	NaN	NaN

In [44]:

```
d_melt=df_tb.iloc[:,[0,1,4]]
model = ols('Instrumentalness ~ C(Genre) + C(Decade) + C(Genre):C(Decade)', data=d_melt
).fit()
anova_table = sm.stats.anova_lm(model, typ=2)
anova_table
```

Out[44]:

	sum_sq	df	F	PR(>F)
C(Genre)	815.699946	13.0	1054.430106	0.000000e+00
C(Decade)	15.742111	5.0	52.908285	6.935869e-55
C(Genre):C(Decade)	98.902500	65.0	25.569641	3.059811e-296
Residual	1979.803487	33270.0	NaN	NaN

In [45]:

```
d_melt=df_tb.iloc[:,[0,1,5]]
model = ols('Valence ~ C(Genre) + C(Decade) + C(Genre):C(Decade)', data=d_melt).fit()
anova_table = sm.stats.anova_lm(model, typ=2)
anova_table
```

Out[45]:

	sum_sq	df	F	PR(>F)
C(Genre)	417.757155	13.0	587.067008	0.000000e+00
C(Decade)	51.910606	5.0	189.667631	6.678632e-200
C(Genre):C(Decade)	42.290686	65.0	11.886076	5.125215e-120
Residual	1821.149802	33270.0	NaN	NaN

In []: