

# exp3

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Computational Neuroscience Experiment 3

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MBA Tech. EXTC SEM 8

Part 1: Calculating Characteristics

```
[ ]: import pandas as pd
import numpy as np
import glob
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
```

```
[ ]: def shannon_entropy(x):
    prob_energy = pow(x,2)/np.sum(pow(x,2))
    shannon = - np.sum(prob_energy*np.log2(prob_energy))
    return shannon

def lee(x):
    prob_energy = pow(x,2)/np.sum(pow(x,2))
    log_energy = np.sum(prob_energy*np.log(prob_energy))
    return log_energy
```

```
[ ]: def individual_chars(path):
    data = pd.read_csv(path)
    for col in ['FZ -A2 ', 'CZ -A1 ', 'PZ -A2 ', 'BP2-REF']:
        data.pop(col)
    print(f"Individual Data for {path}\n")
    data_chars = pd.DataFrame({"Mean":data.mean(), "Median":data.
↪median(), "Summation":data.sum(), "Variance":data.var(), "Standard Deviation":
↪data.std(), "Shannon Entropy":shannon_entropy(data), "Log Energy Entropy":
↪lee(data)})
    return data_chars
```

```
[ ]: def hemisphere_chars(path):
    print(f"Hemisphere Data for {path}")
```

```

data = pd.read_csv(path)
for col in ['FZ -A2 ', 'CZ -A1 ', 'PZ -A2 ', 'BP2-REF']:
    data.pop(col)
right_hemisphere = data[[i for i in data.columns if "A2" in i]].sum()
left_hemisphere = data[[i for i in data.columns if "A1" in i]].sum()
combined = pd.DataFrame({"Mean": [left_hemisphere.mean(), right_hemisphere.
↪mean()], "Median": [left_hemisphere.median(), right_hemisphere.
↪median()], "Summation": [left_hemisphere.sum(), right_hemisphere.
↪sum()], "Variance": [left_hemisphere.var(), right_hemisphere.var()], "Standard_
↪Deviation": [left_hemisphere.std(), right_hemisphere.std()], "Shannon Entropy":
↪[shannon_entropy(left_hemisphere), shannon_entropy(right_hemisphere)], "Log_
↪Energy Entropy": [lee(left_hemisphere), lee(right_hemisphere)]})
combined.index = ["Left", "Right"]
combined = combined.transpose()
return combined

```

```

[ ]: def ecg_chars(path):
    print(f"ECG Data for {path}")
    data = pd.read_csv(path)
    data = data[["BP1-REF", "BP2-REF"]]
    bp1 = data["BP1-REF"]
    bp2 = data["BP2-REF"]
    combined = pd.DataFrame({"Mean": [bp2.mean(), bp1.mean()], "Median": [bp2.
↪median(), bp1.median()], "Summation": [bp2.sum(), bp1.sum()], "Variance": [bp2.
↪var(), bp1.var()], "Standard Deviation": [bp2.std(), bp1.std()], "Shannon_
↪Entropy": [shannon_entropy(bp2), shannon_entropy(bp1)], "Log Energy Entropy":
↪[lee(bp2), lee(bp1)]})
    combined.index = ["BP1", "BP2"]
    combined = combined.transpose()
    return combined

```

```

[ ]: for file in list(glob.glob("*"))[:2]:
    print(individual_chars(file), "\n\n")

for file in list(glob.glob("*"))[:2]:
    print(hemisphere_chars(file), "\n\n")

for file in list(glob.glob("*"))[:2]:
    print(ecg_chars(file), "\n\n")

```

Individual Data for 5\_filtered.csv

	Mean	Median	Summation	Variance	Standard Deviation	\
FP2-A2	0.037109	-1.0	95.0	826.277247	28.745039	
F8 -A2	0.013281	0.0	34.0	374.283528	19.346409	
T4 -A2	0.019922	-1.0	51.0	629.104331	25.081952	
T6 -A2	-0.014453	0.0	-37.0	361.839963	19.022091	
F4 -A2	0.015234	-0.5	39.0	367.993906	19.183167	

C4 -A2	0.032813	-1.0	84.0	399.743354	19.993583
P4 -A2	0.003125	0.0	8.0	576.565836	24.011785
O2 -A2	0.007422	0.0	19.0	325.106236	18.030703
FP1-A1	-0.016016	1.0	-41.0	1015.930576	31.873666
F7 -A1	-0.012109	0.0	-31.0	682.249560	26.119907
T3 -A1	-0.008594	0.0	-22.0	475.606804	21.808411
T5 -A1	-0.006250	0.0	-16.0	524.801837	22.908554
F3 -A1	0.001953	0.0	5.0	542.681903	23.295534
C3 -A1	-0.013672	0.0	-35.0	510.626230	22.597040
P3 -A1	-0.024219	1.0	-62.0	521.439820	22.835057
O1 -A1	-0.017188	2.0	-44.0	1264.352186	35.557730
BP1-REF	0.066016	-37.0	169.0	125332.497790	354.023301

	Shannon Entropy	Log Energy Entropy
FP2-A2	9.991692	-6.925713
F8 -A2	9.989861	-6.924444
T4 -A2	9.702919	-6.725551
T6 -A2	9.910432	-6.869388
F4 -A2	10.286965	-7.130381
C4 -A2	9.960380	-6.904010
P4 -A2	10.069736	-6.979809
O2 -A2	10.051941	-6.967474
FP1-A1	9.845378	-6.824296
F7 -A1	10.197430	-7.068320
T3 -A1	10.057748	-6.971500
T5 -A1	10.117729	-7.013075
F3 -A1	9.989024	-6.923864
C3 -A1	10.054928	-6.969545
P3 -A1	10.057916	-6.971616
O1 -A1	9.956933	-6.901620
BP1-REF	9.165193	-6.352828

Individual Data for 5\_unfiltered.csv

	Mean	Median	Summation	Variance	Standard Deviation \
FP2-A2	-532.105078	-538.0	-1362189.0	34918.605992	186.865208
F8 -A2	-912.291406	-919.5	-2335466.0	50633.135057	225.018077
T4 -A2	-683.771094	-685.0	-1750454.0	36277.901470	190.467586
T6 -A2	-291.118359	-296.5	-745263.0	40198.961757	200.496787
F4 -A2	-1018.371875	-1019.0	-2607032.0	10643.294246	103.166343
C4 -A2	-370.224609	-378.0	-947775.0	37360.661137	193.289061
P4 -A2	-56.543750	-62.5	-144752.0	49656.828097	222.838121
O2 -A2	-748.427344	-750.5	-1915974.0	30537.644582	174.750235
FP1-A1	89.635156	89.0	229466.0	8488.654645	92.133895
F7 -A1	182.542188	181.5	467308.0	9088.840736	95.335412
T3 -A1	-268.410937	-268.0	-687132.0	9389.642710	96.900169
T5 -A1	-174.315234	-176.0	-446247.0	8472.975618	92.048768

F3 -A1	-5.176953	-6.0	-13253.0	11575.022603	107.587279
C3 -A1	-280.126562	-280.0	-717124.0	5242.481826	72.404985
P3 -A1	49.713672	49.0	127267.0	9956.438109	99.781953
O1 -A1	349.441797	348.0	894571.0	12698.221699	112.686386
BP1-REF	-908.325000	-936.0	-2325312.0	162914.573505	403.626775

	Shannon Entropy	Log Energy Entropy
FP2-A2	11.020299	-7.638689
F8 -A2	11.160303	-7.735733
T4 -A2	11.120128	-7.707885
T6 -A2	10.563676	-7.322182
F4 -A2	11.292742	-7.827532
C4 -A2	10.767502	-7.463464
P4 -A2	10.774761	-7.468495
O2 -A2	11.176100	-7.746682
FP1-A1	10.433953	-7.232265
F7 -A1	10.770833	-7.465772
T3 -A1	11.006232	-7.628939
T5 -A1	10.765435	-7.462031
F3 -A1	10.779446	-7.471743
C3 -A1	11.146279	-7.726012
P3 -A1	10.560701	-7.320120
O1 -A1	11.066104	-7.670439
BP1-REF	11.107876	-7.699393

#### Hemisphere Data for 5\_filtered.csv

	Left	Right
Mean	-30.750000	36.625000
Median	-33.000000	36.500000
Summation	-246.000000	293.000000
Variance	406.785714	1774.553571
Standard Deviation	20.168929	42.125450
Shannon Entropy	2.445421	2.238853
Log Energy Entropy	-1.695037	-1.551855

#### Hemisphere Data for 5\_unfiltered.csv

	Left	Right
Mean	-1.814300e+04	-1.476113e+06
Median	5.700700e+04	-1.556322e+06
Summation	-1.451440e+05	-1.180890e+07
Variance	3.243688e+11	6.978557e+11
Standard Deviation	5.695338e+05	8.353776e+05
Shannon Entropy	2.296372e+00	2.467787e+00
Log Energy Entropy	-1.591724e+00	-1.710539e+00

ECG Data for 5\_filtered.csv

	BP1	BP2
Mean	-0.024219	0.066016
Median	0.000000	-37.000000
Summation	-62.000000	169.000000
Variance	246.054122	125332.497790
Standard Deviation	15.686112	354.023301
Shannon Entropy	9.299781	9.165193
Log Energy Entropy	-6.446117	-6.352828

ECG Data for 5\_unfiltered.csv

	BP1	BP2
Mean	-6.062273e+02	-9.083250e+02
Median	-6.050000e+02	-9.360000e+02
Summation	-1.551942e+06	-2.325312e+06
Variance	5.767807e+02	1.629146e+05
Standard Deviation	2.401626e+01	4.036268e+02
Shannon Entropy	1.131739e+01	1.110788e+01
Log Energy Entropy	-7.844618e+00	-7.699393e+00

## Part 2: Unfiltered Data Graphs

```
[ ]: N = 2
ind = np.arange(N)
width = 0.1

fig = plt.figure(figsize=(30,20))
ax = fig.add_subplot(2,3,1)

mean = [-98214.66,-144752.0]
rects1 = ax.bar(ind, mean, width, color='r')

median = [-13253.0,-1200034.5]
rects2 = ax.bar(ind+width, median, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Mean', 'Median'],loc="best")
def autolabel(rects):
    for rect in rects:
        h = rect.get_height()
        ax.text(rect.get_x()+rect.get_width()/2., 1.05*h, '%0.
↳2f'%int(h),ha='center', va='bottom')
```

```

autolabel(rects1)
autolabel(rects2)

ax = fig.add_subplot(2,3,2)
var = [341525930172.0,609048699770.4888]
rects1 = ax.bar(ind+width, var, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Variance'],loc="center")
autolabel(rects1)

ax = fig.add_subplot(2,3,3)
std = [584402.19,780415.72]
rects1 = ax.bar(ind+width, std, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Standard Deviation'],loc="center")
autolabel(rects1)

ax = fig.add_subplot(2,3,4)
kurt = [-1.0395,-0.833]
rects2 = ax.bar(ind+width, kurt, width, color='g')

skew = [0.370,-0.219]
rects1 = ax.bar(ind+width, skew, width, color='r')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Kurtosis','Skewness'],loc="best")
autolabel(rects1)
autolabel(rects2)

ax = fig.add_subplot(2,3,5)
rms = [559664.833,1547272.762]
rects1 = ax.bar(ind+width, rms, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Root Mean Square'],loc="center")
autolabel(rects1)

```

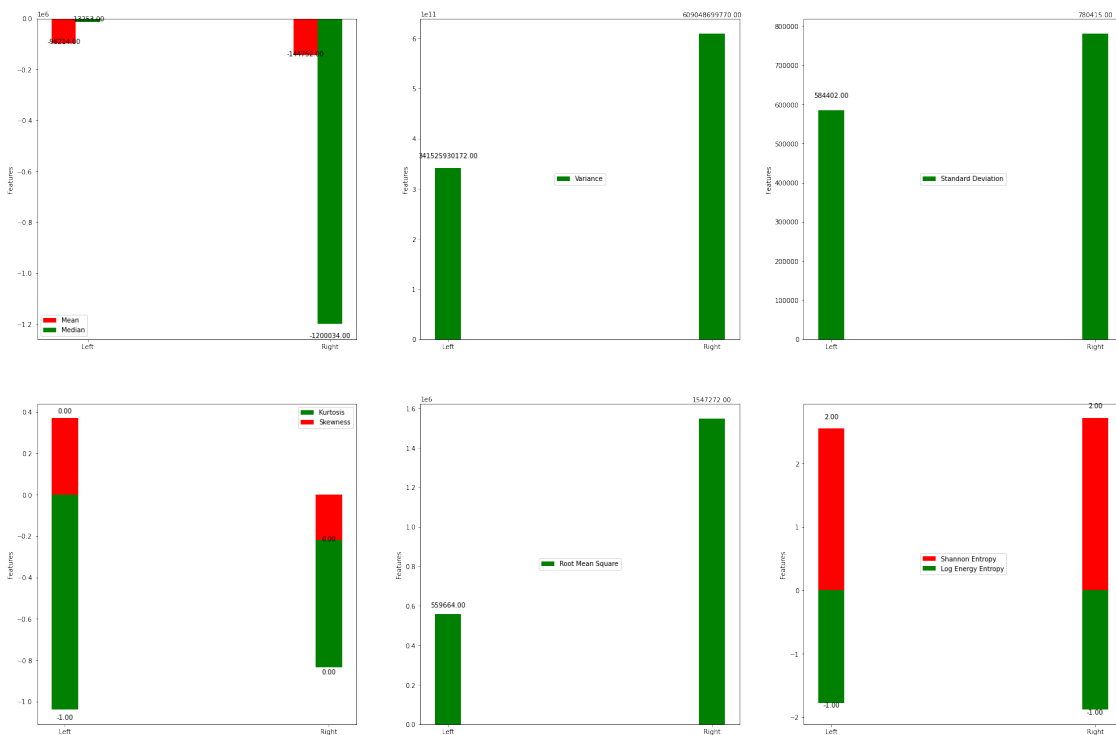
```

ax = fig.add_subplot(2,3,6)
se = [2.560,2.720]
rects1 = ax.bar(ind+width, se, width, color='r')

lee = [-1.774, -1.885]
rects2 = ax.bar(ind+width, lee, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Shannon Entropy', 'Log Energy Entropy'],loc="center")
autolabel(rects1)
autolabel(rects2)

```



### Part 3: Filtered Data Graphs

```

[ ]: N = 2
ind = np.arange(N)
width = 0.1

fig = plt.figure(figsize=(30,20))
ax = fig.add_subplot(2,3,1)

mean = [-29.555, 42.8]

```

```

rects1 = ax.bar(ind, mean, width, color='r')

median = [-31.0          , 36.5]
rects2 = ax.bar(ind+width, median, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Mean', 'Median'],loc="best")
def autolabel(rects):
    for rect in rects:
        h = rect.get_height()
        ax.text(rect.get_x()+rect.get_width()/2., 1.05*h, '%0.
↪2f'%int(h),ha='center', va='bottom')

autolabel(rects1)
autolabel(rects2)

ax = fig.add_subplot(2,3,2)
var = [368.777          ,2417.733]
rects1 = ax.bar(ind+width, var, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Variance'],loc="center")
autolabel(rects1)

ax = fig.add_subplot(2,3,3)
std = [19.203          , 49.170]
rects1 = ax.bar(ind+width, std, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Standard Deviation'],loc="center")
autolabel(rects1)

ax = fig.add_subplot(2,3,4)
kurt = [0.629, -0.138]
rects2 = ax.bar(ind+width, kurt, width, color='g')

skew = [0.141,0.300]
rects1 = ax.bar(ind+width, skew, width, color='r')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)

```



```

ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Kurtosis', 'Skewness'],loc="best")
autolabel(rects1)
autolabel(rects2)

ax = fig.add_subplot(2,3,5)
rms = [34.660      , 63.307]
rects1 = ax.bar(ind+width, rms, width, color='g')

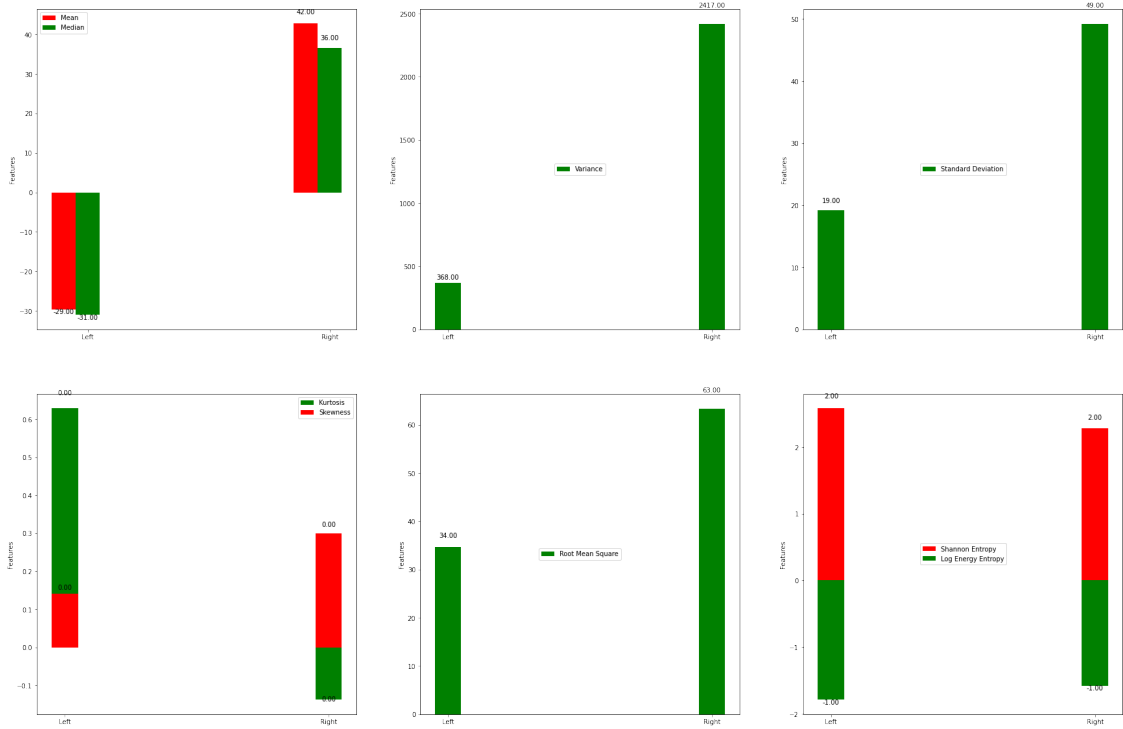
ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Root Mean Square'],loc="center")
autolabel(rects1)

ax = fig.add_subplot(2,3,6)
se = [2.583      , 2.282]
rects1 = ax.bar(ind+width, se, width, color='r')

lee_met = [-1.790, -1.582]
rects2 = ax.bar(ind+width, lee_met, width, color='g')

ax.set_ylabel('Features')
ax.set_xticks(ind+width)
ax.set_xticklabels( ('Left', 'Right') )
ax.legend(['Shannon Entropy', 'Log Energy Entropy'],loc="center")
autolabel(rects1)
autolabel(rects2)

```



[ ]: