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In [3]: ### Written by vapticuno, May 8, 2023
### Viability ranking analysis of 2023 ADV OU (Spring) metagame
### VR post https://www.smogon.com/forums/threads/adv-ou-viability-ranking.3503019/page-30

import numpy as np
import matplotlib.pyplot as plt
import matplotlib.transforms as transforms
import matplotlib.patches as patches
import matplotlib.cm as cm
import matplotlib.colors as colors
%matplotlib inline
import pandas as pd
import itertools
from timeit import default_timer as timer
from scipy import stats
from scipy import cluster
import pandas as pd
import copy
```

```
In [45]: ### Initialization

# csv input with header=voters and row labels=Pokemon
fnIn = '2023_ADV_rankings_outliers_in.csv'
fnOut = '2023_ADV_rankings_outliers_out.csv'
fnPrevIn = '2022_R2_ADV_rankings_outliers_in.csv'
fnPrevOut = '2022_R2_ADV_rankings_outliers_out.csv'
cutoff = 37 # cut off analysis at this number of Pokemon
rankFracThreshold = 1-6.5/24 # throw out Pokemon that are not ranked by over this fraction of voters
gen = 'ADV OU'
year = '2023'

rank = pd.read_csv(fnIn,header=0,index_col=0)
rankval = rank.values
N = np.shape(rankval)
```

```
In [46]: ### Outlier compensation
rankvalNoOutliers = copy.deepcopy(rankval)
Q1 = np.quantile(rankval,0.16,axis=1) # set reference range
Q3 = np.quantile(rankval,0.84,axis=1) # set reference range
IQR = Q3 - Q1
limit = 1 # set outlier boundary
for ii in range(0,rankval.shape[0]):
    lb = Q1[ii] - limit*IQR[ii]
    ub = Q3[ii] + limit*IQR[ii]
    for jj in range(0,rankval.shape[1]):
        if rankval[ii,jj] == np.nan:
            rankvalNoOutliers[ii,jj] = np.nan
        elif rankval[ii,jj] < lb:
            rankvalNoOutliers[ii,jj] = lb
        elif rankval[ii,jj] > ub:
            rankvalNoOutliers[ii,jj] = ub
```

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In [47]: rankNoOutliers = pd.DataFrame(rankvalNoOutliers)
rankNoOutliers.to_csv(fnOut, header=False, index=False)
```

```
In [48]: ### Sort data by outlier-removed ranks
rankval = rank.values
rankavg = np.nanmean(rankvalNoOutliers,axis=1)
rankstd = np.nanstd(rankvalNoOutliers,axis=1)
orderNoOutliers = np.argsort(rankavg)
rankavg = rankavg[orderNoOutliers]
rankstd = rankstd[orderNoOutliers]
monList = rank.index.values
monList = monList[orderNoOutliers]
rankval = rankval[orderNoOutliers,:]
# Ranking Difference
rankMeanSubtracted = rankval - np.repeat(np.array([rankavg]).transpose(),rankv
al.shape[1],axis=1)
```

```
In [49]: ### Final VR order  
for n in range(0,N[0]):  
    print(str(n+1).zfill(2) + ' ' + monList[n])
```

01 Tyranitar
02 Skarmory
03 Metagross
04 Swampert
05 Blissey
06 Zapdos
07 Gengar
08 Salamence
09 Jirachi
10 Celebi
11 Aerodactyl
12 Suicune
13 Claydol
14 Dugtrio
15 Starmie
16 Moltres
17 Charizard
18 Snorlax
19 Magnetron
20 Flygon
21 Milotic
22 Breloom
23 Forretress
24 Heracross
25 Hariyama
26 Gyarados
27 Jolteon
28 Cloyster
29 Regice
30 Smeargle
31 Venusaur
32 Kingdra
33 Jynx
34 Medicham
35 Porygon2
36 Raikou
37 Vaporeon
38 Machop
39 Registeel
40 Ludicolo
41 Weezing
42 Misdreavus
43 Blaziken
44 Steelix
45 Regirock
46 Camerupt
47 Armaldo
48 Glalie
49 Marowak
50 Articuno
51 Houndoom
52 Dusclops
53 Haunter
54 Donphan
55 Alakazam
56 Quagsire
57 Banette

58 Poliwrath
59 Lapras
60 Roselia
61 Hypno
62 Gardevoir
63 Umbreon
64 Dragonite
65 Golduck
66 Slaking
67 Sceptile
68 Exeggutor
69 Ninjask
70 Pinsir
71 Rhydon
72 Politoed
73 Rapidash
74 Scizor
75 Arcanine
76 Lanturn
77 Slowbro
78 Blastoise
79 Dewgong
80 Ursaring
81 Walrein
82 Lunatone
83 Mantine
84 Solrock
85 Jumpluff
86 Miltank
87 Sableye
88 Entei
89 Nidoqueen
90 Omastar
91 Aggron
92 Swellow
93 Nidoking

```
In [50]: ### Final VR order after removing frequently unranked Pokemon
orderCount = 0
monCount = 1
orderReduced = []
for n in range(0,N[0]):
    orderCount += 1
    if sum(np.isnan(rankval[n,:])) > rankFracThreshold*N[1]:
        continue
    print(str(monCount).zfill(2) + ' ' + monList[n])
    orderReduced += [orderCount-1]
    monCount += 1
```

01 Tyranitar
02 Skarmory
03 Metagross
04 Swampert
05 Blissey
06 Zapdos
07 Gengar
08 Salamence
09 Jirachi
10 Celebi
11 Aerodactyl
12 Suicune
13 Claydol
14 Dugtrio
15 Starmie
16 Moltres
17 Charizard
18 Snorlax
19 Magnetron
20 Flygon
21 Milotic
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23 Forretress
24 Heracross
25 Hariyama
26 Gyarados
27 Jolteon
28 Cloyster
29 Regice
30 Smeargle
31 Venusaur
32 Kingdra
33 Jynx
34 Medicham
35 Porygon2
36 Raikou
37 Vaporeon
38 Machop
39 Registeel
40 Ludicolo
41 Weezing
42 Misdreavus
43 Blaziken
44 Steelix
45 Regirock
46 Camerupt
47 Armaldo
48 Glalie
49 Marowak
50 Houndoom
51 Roselia
52 Sceptile

```

In [51]: ### Analyze previous rankings
rankOld = pd.read_csv(fnPrevIn,header=0,index_col=0)
rankvalOld = rankOld.values
Nold = np.shape(rankvalOld)

rankNoOutliersOld = pd.DataFrame(rankvalNoOutliers)
rankNoOutliersOld.to_csv(fnPrevOut, header=False, index=False)

rankvalNoOutliersOld = copy.deepcopy(rankvalOld)
Q1old = np.quantile(rankvalOld,0.16,axis=1)
Q3old = np.quantile(rankvalOld,0.84,axis=1)
IQRold = Q3old - Q1old
limitOld = 1
for ii in range(0,rankvalOld.shape[0]):
    lb = Q1old[ii] - limitOld*IQRold[ii]
    ub = Q3old[ii] + limitOld*IQRold[ii]
    for jj in range(0,rankvalOld.shape[1]):
        if rankvalOld[ii,jj] == np.nan:
            rankvalNoOutliersOld[ii,jj] = np.nan
        elif rankvalOld[ii,jj] < lb:
            rankvalNoOutliersOld[ii,jj] = lb
        elif rankvalOld[ii,jj] > ub:
            rankvalNoOutliersOld[ii,jj] = ub

rankavgOld = np.nanmean(rankvalNoOutliersOld,axis=1)
rankstdOld = np.nanstd(rankvalNoOutliersOld,axis=1)
orderNoOutliersOld = np.argsort(rankavgOld)
rankavgOld = rankavgOld[orderNoOutliersOld]
rankstdOld = rankstdOld[orderNoOutliersOld]
monListOld = rankOld.index.values
monListOld = monListOld[orderNoOutliersOld]

indexOldIntoNew = [list(monListOld).index(mon) for mon in monList[:cutoff]]
rankOldIntoNew = rankavgOld[indexOldIntoNew]
rankDiff = rankavg[:cutoff] - rankOldIntoNew

```



```

In [52]: ### Compare current and previous rankings
seismic = cm.get_cmap('coolwarm', 32)
rankDiffStd = np.sqrt(rankstdOld[:cutoff]**2/Nold[1]+rankstd[:cutoff]**2/N[1])
zChange = -rankDiff / (rankDiffStd+1e-7)

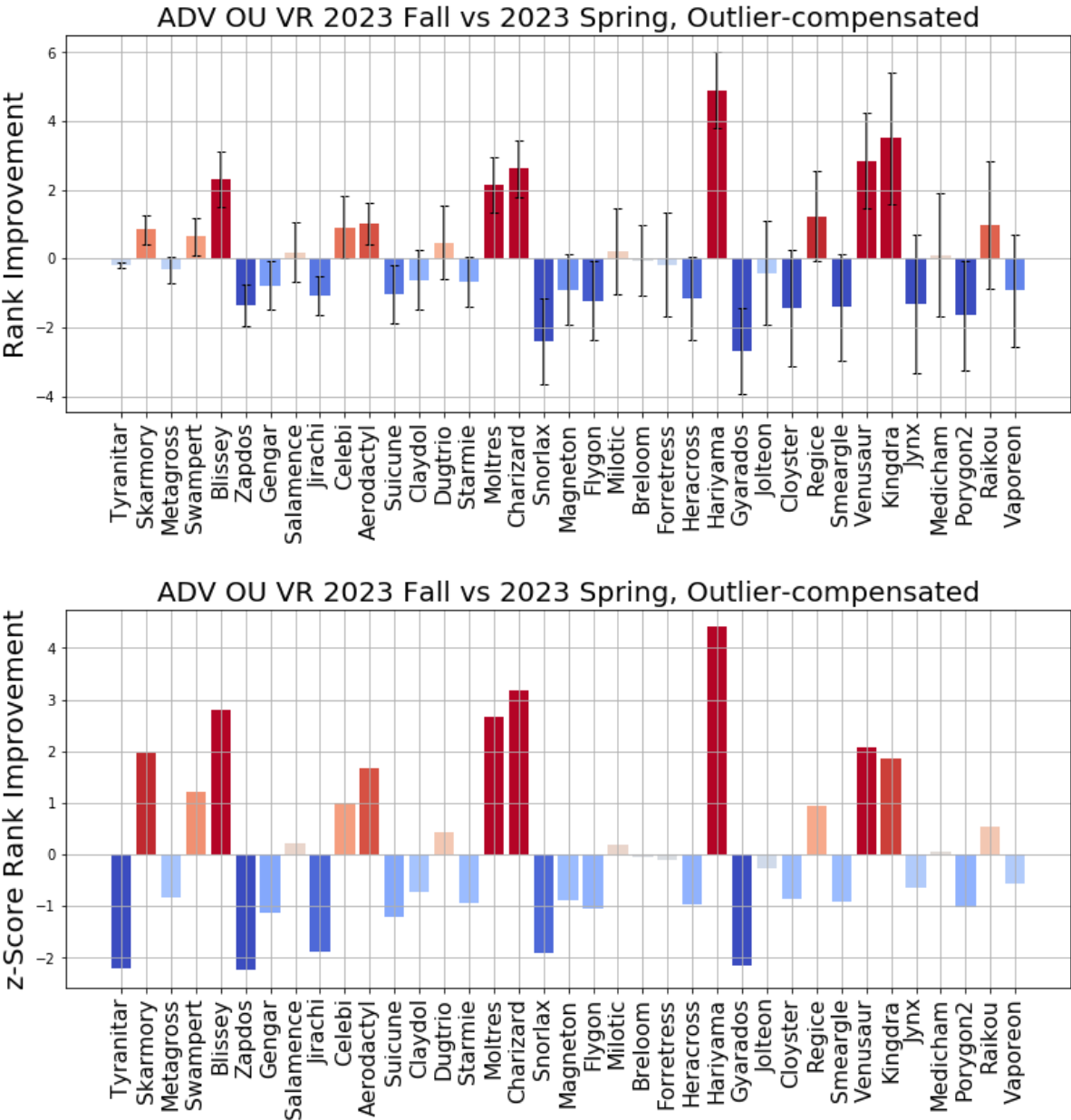
fig = plt.figure(figsize=(13,5))
ax = plt.axes()
ax.grid(zorder=0)
ax.bar(list(monList[:cutoff]),list(-rankDiff),zorder=1,color=seismic(0.5-rankD
iff/max(rankDiff)),yerr=rankDiffStd,ecolor='k',capsize=3)
plt.xticks(rotation=90,fontsize=580/cutoff)
plt.ylabel('Rank Improvement',fontsize=20)
#plt.title(gen + ' VR ' + year + ' vs ' + str(int(year)-1) + ', Outlier-compen
sated',fontsize=20)
plt.title(gen + ' VR ' + year + ' Fall vs ' + year + ' Spring, Outlier-compens
ated',fontsize=20)

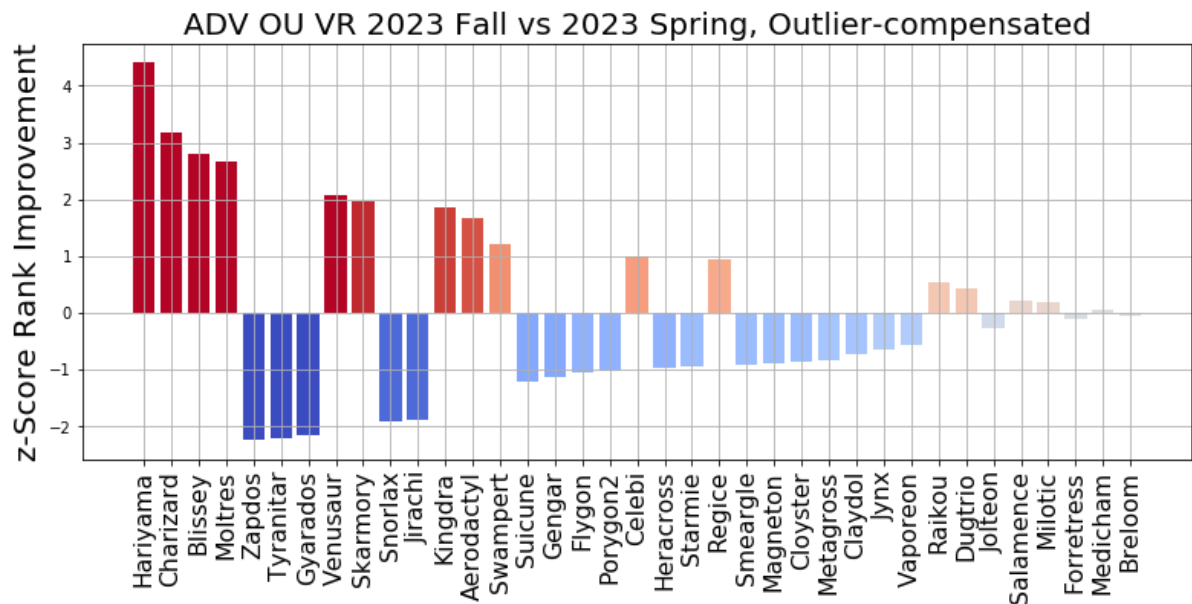
fig = plt.figure(figsize=(13,5))
ax = plt.axes()
ax.grid(zorder=0)
ax.bar(list(monList[:cutoff]),list(zChange),zorder=1,color=seismic(0.5+zChang
e/max(np.abs(zChange))))
plt.xticks(rotation=90,fontsize=580/cutoff)
plt.ylabel('z-Score Rank Improvement',fontsize=20)
#plt.title(gen + ' VR ' + year + ' vs ' + str(int(year)-1) + ', Outlier-compen
sated',fontsize=20)
plt.title(gen + ' VR ' + year + ' Fall vs ' + year + ' Spring, Outlier-compens
ated',fontsize=20)

orderChange = sorted(range(len(monList[:cutoff])),reverse=True,key=lambda x:n
p.abs(zChange[x]))
fig = plt.figure(figsize=(13,5))
ax = plt.axes()
ax.grid(zorder=0)
ax.bar(list(monList[:cutoff][orderChange]),list(zChange[orderChange]),zorder=
1,color=seismic(0.5+zChange[orderChange]/max(np.abs(zChange))))
plt.xticks(rotation=90,fontsize=580/cutoff)
plt.ylabel('z-Score Rank Improvement',fontsize=20)
#plt.title(gen + ' VR ' + year + ' vs ' + str(int(year)-1) + ', Outlier-compen
sated',fontsize=20)
plt.title(gen + ' VR ' + year + ' Fall vs ' + year + ' Spring, Outlier-compens
ated',fontsize=20)

```

Out[52]: Text(0.5, 1.0, 'ADV OU VR 2023 Fall vs 2023 Spring, Outlier-compensated')





```
In [53]: ### Substitute non-ranked Pokemon according to missingRankMethod
# If using mean rank, missingRankMethod = 0; if substituting by rank 999, missingRankMethod = 1
# Used for dendrogram, dissimilarity matrix and camps analysis
missingRankMethod = 0

rankvalSub = copy.deepcopy(rankval)
for m in np.arange(0,cutoff):
    for n in np.arange(0,N[1]):
        if np.isnan(rankval[m,n]):
            if missingRankMethod == 0:
                rankvalSub[m,n] = rankavg[m]
            elif missingRankMethod == 1:
                rankvalSub[m,n] = 999
```

```

In [54]: ### Handle non-ranked Pokemon and eliminating correlation in case of equal ranks

maxDist = 3.5
dissimMons = np.zeros((cutoff,cutoff))
def compareArr(arr1,arr2):
    assert len(arr1) == len(arr2)
    N = len(arr1)
    arr1cp = copy.deepcopy(arr1)
    arr2cp = copy.deepcopy(arr2)
    logicalArr = np.zeros(N)
    for n in range(N):
        if arr1cp[n] == arr2cp[n]:
            logicalArr[n] = 0.5
        else:
            logicalArr[n] = arr1cp[n] > arr2cp[n]
    return logicalArr

### Perform rank rate comparison across Pokemon pairs and perform Logit transform
for m1 in range(cutoff):
    for m2 in range(cutoff):
        if m1 == m2:
            continue
        rate = sum(compareArr(rankvalSub[m1,:],rankvalSub[m2,:])) / N[1]
        dissimMons[m1,m2] = min(abs(np.log(rate/(1-rate))),maxDist)

```

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:24: RuntimeWarning: divide by zero encountered in log
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:24: RuntimeWarning: divide by zero encountered in double_scalars

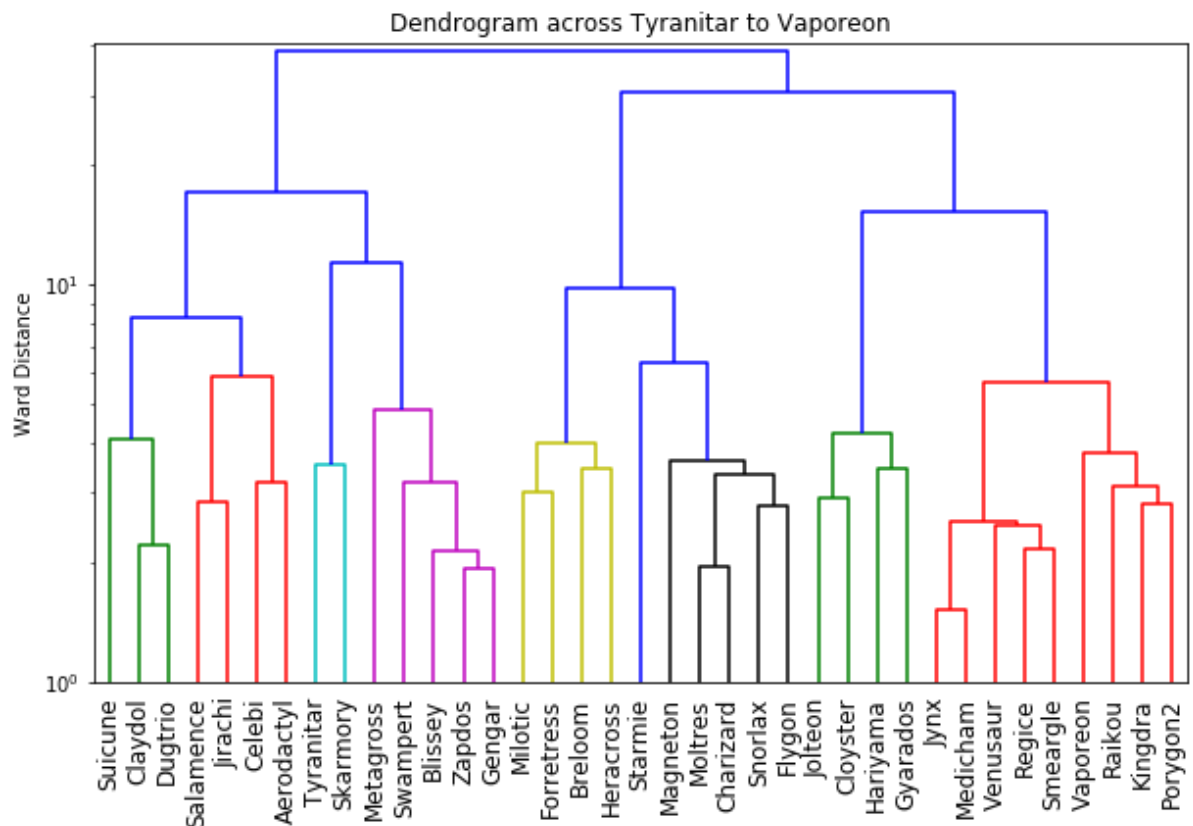
```
In [55]: ### Perform hierarchical clustering with Ward Linkage
# Adjust the threshold to get the appropriate tiers
tierThreshold = 6 # threshold
Hmons = cluster.hierarchy.ward(dissimMons)
fig = plt.figure(figsize=(10,6))
ax = plt.gca()
ax.set_yscale('log')
ax.set_ylim([1,50])
ax.set_ylabel('Ward Distance')
Dmons = cluster.hierarchy.dendrogram(Hmons,labels=[monList[i] for i in range(cutoff)],leaf_rotation=90,color_threshold=tierThreshold,leaf_font_size=12)
clustersMons = cluster.hierarchy.fcluster(Hmons, tierThreshold, criterion='distance')
plt.title('Dendrogram across ' + monList[0] + ' to ' + monList[cutoff-1]);
```

C:\Users\lambj\anaconda3\lib\site-packages\scipy\cluster\hierarchy.py:830: ClusterWarning: scipy.cluster: The symmetric non-negative hollow observation matrix looks suspiciously like an uncondensed distance matrix

return linkage(y, method='ward', metric='euclidean')

C:\Users\lambj\anaconda3\lib\site-packages\scipy\cluster\hierarchy.py:2834: UserWarning: Attempted to set non-positive bottom ylim on a log-scaled axis. Invalid limit will be ignored.

ax.set_ylim([0, dvw])



```
In [56]: # Check consistency of clustering with rankings
# If inconsistent, Pokemon of the same cluster number will be separated.
print('Cluster Labels: ' + ",".join([str(m) for m in clustersMons]))
uniqueLabels = [clustersMons[0]]
for x in clustersMons:
    if x != uniqueLabels[-1]:
        uniqueLabels.append(x)
crossovers = []
for ii in range(len(uniqueLabels)-3):
    if uniqueLabels[ii] == uniqueLabels[ii+2] and uniqueLabels[ii+1] == uniqueLabels[ii+3]:
        crossovers.append([uniqueLabels[ii], uniqueLabels[ii+1]])
for p in crossovers:
    idx10 = list(clustersMons).index(p[1])
    idx00 = list(clustersMons).index(p[0], idx10)
    idx11 = list(clustersMons).index(p[1], idx00)
    if idx00 - idx10 == 1 and idx11 - idx00 > 1:
        print(monList[idx10] + ' is miscategorized / misordered')
    elif idx00 - idx10 > 1 and idx11 - idx00 == 1:
        print(monList[idx00] + ' is miscategorized / misordered')
    else:
        print(", ".join(monList[idx10:idx11]) + ' are miscategorized / misordered')
```

Cluster Labels: 3,3,4,4,4,4,4,2,2,2,2,1,1,1,7,6,6,6,6,6,5,5,5,5,8,8,8,8,9,9,9,9,9,9,9

```
In [57]: # Tier Modification if inconsistent above, as well as any visual changes from dendrogram
#clustersMons[list(monList).index('Porygon2')] = clustersMons[list(monList).index('Regice')]
#clustersMons[list(monList).index('Dugtrio')] = clustersMons[list(monList).index('Aerodactyl')]

### Obtain clusters
clusterDividers = [0] + [clustersMons[i+1]-clustersMons[i] != 0 for i in range(len(clustersMons)-1)] + [1]
clusterIndices = np.insert(np.nonzero(clusterDividers)[0], 0, 0)

### Modify cluster indices (insert lower tier cutoffs before len(monList), check graph below)
clusterIndicesMod = np.append(clusterIndices, [49, len(monList)])
clusterIndicesMod.sort()

### Tier names (see dendrogram to determine numerical subtiers)
tierNames = ['S', 'A', 'B1', 'B2', 'B3', 'C1', 'C2', 'D1', 'D2', 'E1', 'E2']
```

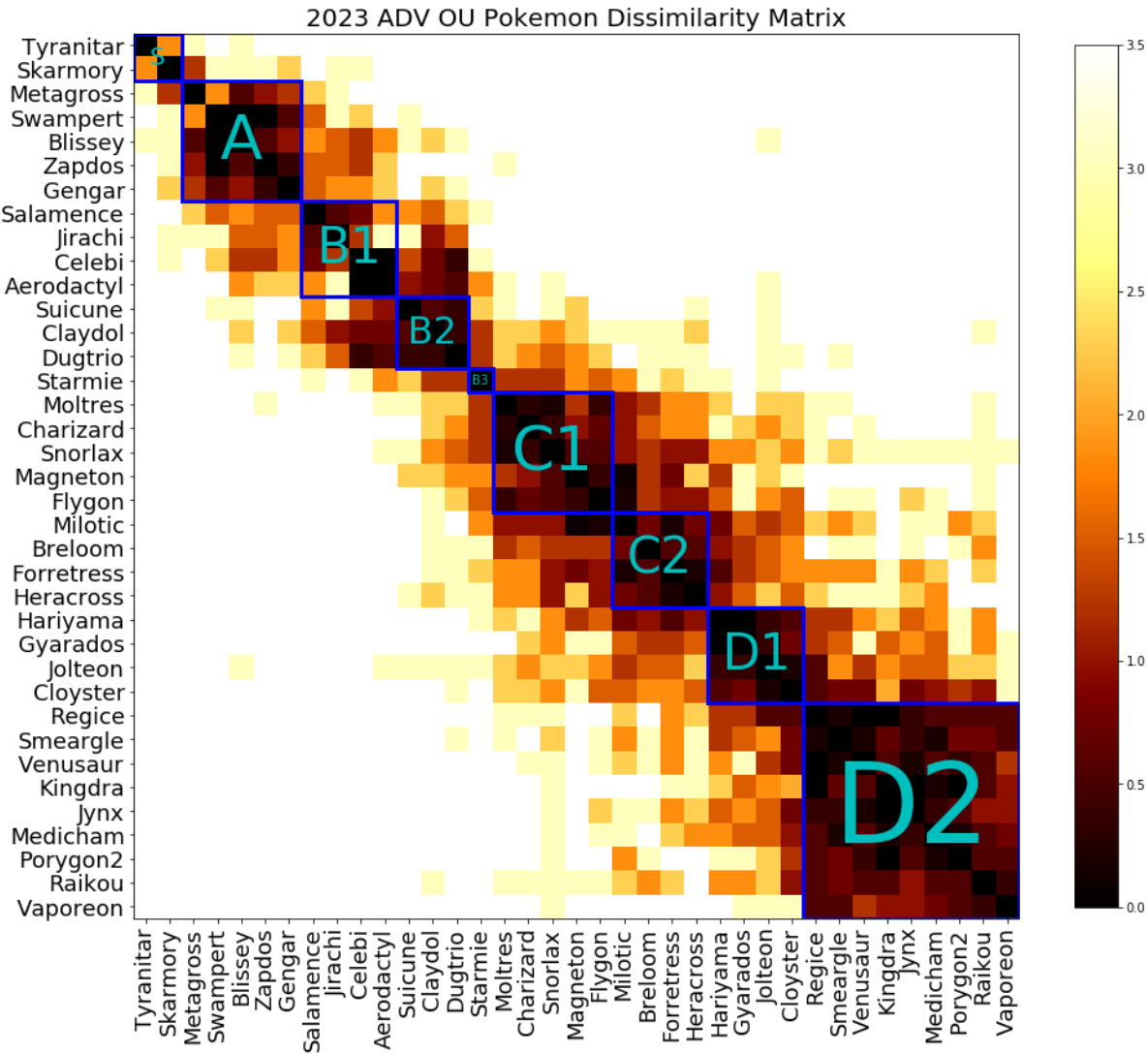
```
In [58]: ### Print VR tier writeup to be used in Smogon forums
textVR = ''
for n in range(len(clusterIndicesMod)-1):
    textVR += tierNames[n] + ': '
    for m in range(clusterIndicesMod[n],clusterIndicesMod[n+1]):
        if sum(np.isnan(rankval[m,:])) > rankFracThreshold*N[1]:
            continue
        textVR += ':' + monList[m] + ':'
    textVR += '\n'
print(textVR)
```

```
S: :Tyranitar::Skarmory:
A: :Metagross::Swampert::Blissey::Zapdos::Gengar:
B1: :Salamence::Jirachi::Celebi::Aerodactyl:
B2: :Suicune::Claydol::Dugtrio:
B3: :Starmie:
C1: :Moltres::Charizard::Snorlax::Magnetron::Flygon:
C2: :Milotic::Breloom::Forretress::Heracross:
D1: :Hariyama::Gyarados::Jolteon::Cloyster:
D2: :Regice::Smeagle::Venusaur::Kingdra::Jynx::Medicham::Porygon2::Raikou::V
apoleon:
E1: :Machop::Registeel::Ludicolo::Weezing::Misdreavus::Blaziken::Steelix::Re
girock::Camerupt::Armaldo::Glalie::Marowak:
E2: :Houndoom::Roselia::Sceptile:
```

```

In [59]: ### Draw dissimilarity matrix from rank rates across pairs of Pokemon
fig = plt.figure(figsize=(16,16))
ax = plt.gca()
h = ax.imshow(dissimMons, interpolation='none', cmap='afmhot', vmin=0, vmax=maxDis
t)
for c in range(len(clusterIndicesMod)-1):
    ax.plot([clusterIndicesMod[c]-0.5, clusterIndicesMod[c+1]-0.5], [clusterIndi
cesMod[c]-0.5, clusterIndicesMod[c]-0.5], color='b', linewidth=3)
    ax.plot([clusterIndicesMod[c]-0.5, clusterIndicesMod[c+1]-0.5], [clusterIndi
cesMod[c+1]-0.5, clusterIndicesMod[c+1]-0.5], color='b', linewidth=3)
    ax.plot([clusterIndicesMod[c]-0.5, clusterIndicesMod[c]-0.5], [clusterIndice
sMod[c]-0.5, clusterIndicesMod[c+1]-0.5], color='b', linewidth=3)
    ax.plot([clusterIndicesMod[c+1]-0.5, clusterIndicesMod[c+1]-0.5], [clusterIn
dicesMod[c]-0.5, clusterIndicesMod[c+1]-0.5], color='b', linewidth=3)
fig.colorbar(h, ax=ax, shrink=0.8)
# ax.set_title('Dissimilarity Across S/A/B' + ' Camps\n Red = ' + namesA[camps
[0][0]] + ' Camp Favors Y; ' + namesA[camps[1][0]] + ' Camp Favors X\n Blue = '
+ namesA[camps[1][0]] + ' Camp Favors Y; ' + namesA[camps[0][0]] + ' Camp Fa
vors X', fontsize=18)
ax.set_xticks(range(0, dissimMons.shape[1]))
ax.set_yticks(range(0, dissimMons.shape[0]))
ax.set_xticklabels([monList[i] for i in range(cutoff)], fontsize=18)
ax.set_yticklabels([monList[i] for i in range(cutoff)], fontsize=18)
ax.set_xlim([-1/2, cutoff-1/2])
ax.set_ylim([cutoff-1/2, -1/2])
ax.set_title(year + ' ' + gen + ' Pokemon Dissimilarity Matrix', fontsize=20)
plt.setp(ax.get_xticklabels(), rotation=90, ha="right", va="center", rotation_mo
de="anchor");
for ii in range(len(clusterIndicesMod)-1):
    pos = (clusterIndicesMod[ii]+clusterIndicesMod[ii+1]-1)/2
    if pos > cutoff:
        continue
    tierFontSize = (clusterIndicesMod[ii+1]-clusterIndicesMod[ii])*10
    text = ax.text(pos, pos, tierNames[ii], ha="center", va="center", color="c", f
ontsize=tierFontSize)

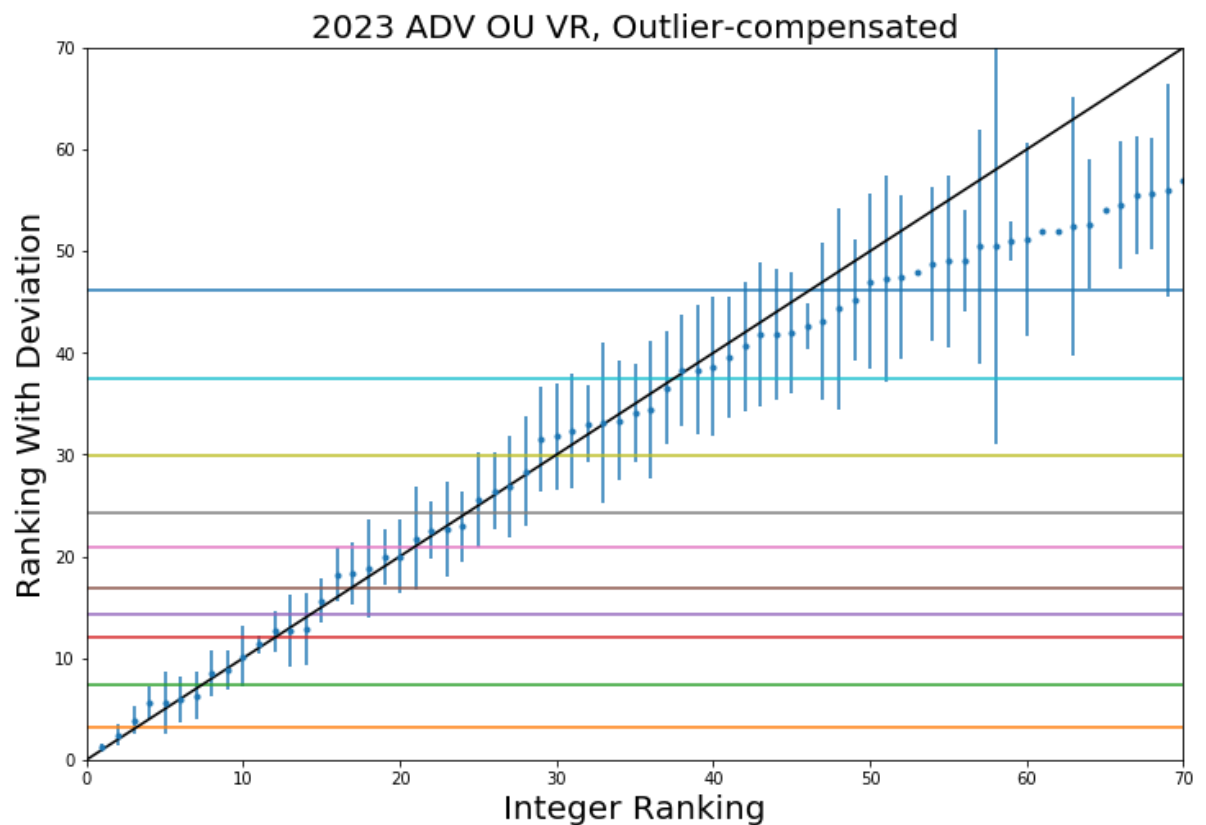
```

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In [60]: ### Draw VR plot
monsLim = 70 # decide on a reasonable limit
# monsLim = len(monList)
fig = plt.figure(figsize=(12,8))
ax = plt.axes()
plt.errorbar(np.arange(1,N[0]+1),rankavg,yerr=rankstd,marker='.',linestyle='none')
for c in clusterIndicesMod[1:-1]:
    plt.plot([0,monsLim],[(rankavg[c-1]+rankavg[c])/2]*2)
plt.plot([0,monsLim],[0,monsLim],color='k')
plt.xlim([0,monsLim])
plt.ylim([0,monsLim])
plt.xlabel('Integer Ranking',fontsize=20)
plt.ylabel('Ranking With Deviation',fontsize=20)
plt.title(year + ' ' + gen + ' VR, Outlier-compensated',fontsize=20);

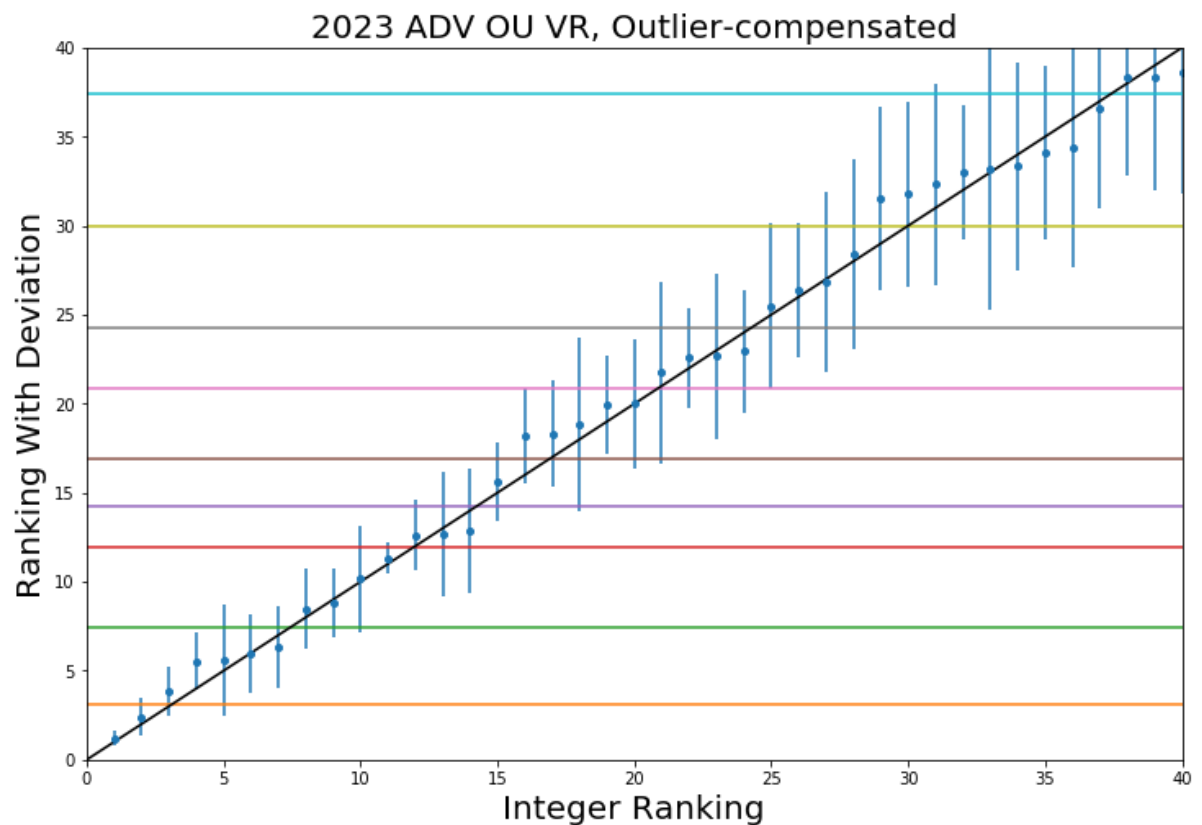
```



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In [61]: ### Draw zoomed in VR plot
monsLimZoom = 40 # zoom to this number
fig = plt.figure(figsize=(12,8))
ax = plt.axes()
plt.errorbar(np.arange(1,N[0]+1),rankavg,yerr=rankstd,marker='.',linestyle='none',markersize=8)
for c in clusterIndicesMod[1:-1]:
    plt.plot([0,monsLim],[0,monsLim],[(rankavg[c-1]+rankavg[c])/2]*2)
plt.plot([0,monsLim],[0,monsLim],color='k')
plt.xlim([0,monsLimZoom])
plt.ylim([0,monsLimZoom])
plt.xlabel('Integer Ranking',fontsize=20)
plt.ylabel('Ranking With Deviation',fontsize=20)
plt.title(year + ' ' + gen + ' VR, Outlier-compensated',fontsize=20);

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In [62]: ### Calculate ranking correlations for dissimilarity matrix across voters
C = len(clusterIndicesMod) - 1
totalCountArr = np.zeros((N[1],N[1],C,C))
cumulativeCountArr = np.zeros((N[1],N[1],C,C))
corrArr = np.zeros((N[1],N[1],C,C))
tArr = np.zeros((N[1],N[1],C,C))
for n1 in np.arange(0,N[1]):
    start = timer();
    for n2 in np.arange(0,N[1]):
        if n1 > n2:
            corrArr[n1,n2,:,:] = corrArr[n2,n1,:,:]
            totalCountArr[n1,n2,:,:] = totalCountArr[n2,n1,:,:]
        elif n1 == n2:
            continue
        else:
            for c1 in range(C):
                for c2 in range(c1+1):
                    for m1 in range(clusterIndicesMod[c1],clusterIndicesMod[c1
+1]):
                        for m2 in range(clusterIndicesMod[c2],clusterIndicesMo
d[c2+1]):
                            if m2 < m1:
                                corr = np.sign(rankvalSub[m1,n1]-rankvalSub[m
2,n1])*np.sign(rankvalSub[m1,n2]-rankvalSub[m2,n2])
                                corrArr[n1,n2,c1,c2] += corr
                                if corr != 0:
                                    totalCountArr[n1,n2,c1,c2] += 1

            for c1 in range(C):
                for c2 in range(c1+1):
                    tArr[:, :, c1, c2] = np.sum(corrArr[:, :, c2:c1, c2:c1], axis=(2,3))
                    cumulativeCountArr[:, :, c1, c2] = np.sum(totalCountArr[:, :, c2:c1, c2:c1],
axis=(2,3))

tArr = tArr / cumulativeCountArr
for n1 in range(N[1]):
    tArr[n1,n1,:,:] = 1

```

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:21: RuntimeWarning: invalid value encountered in sign

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:31: RuntimeWarning: invalid value encountered in true_divide

```

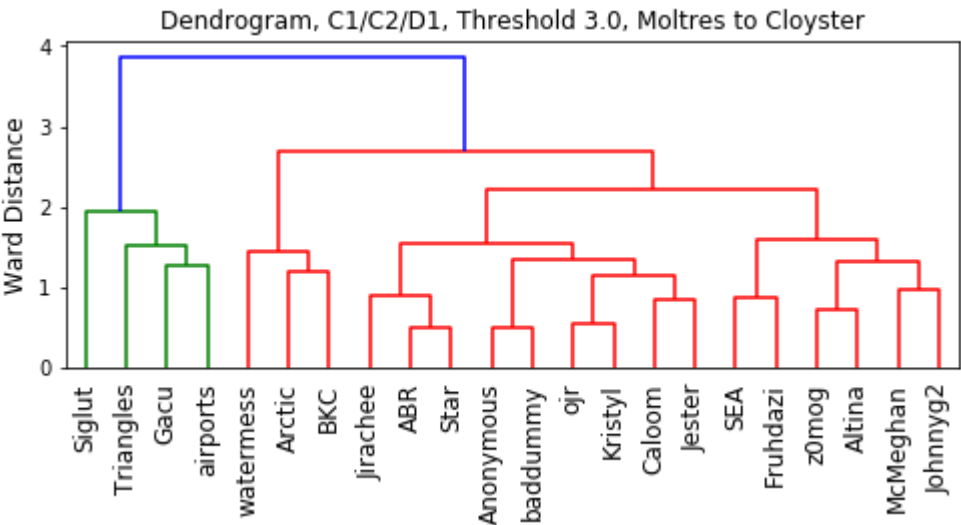
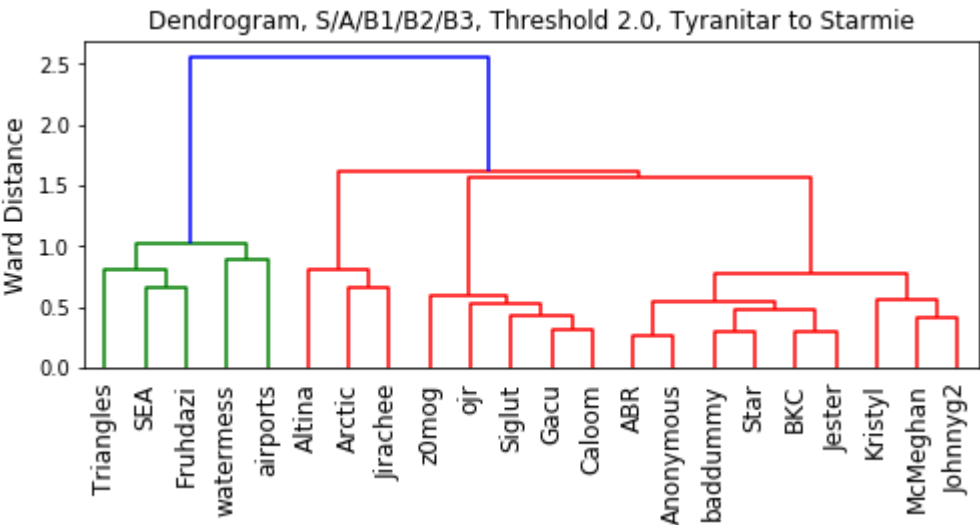
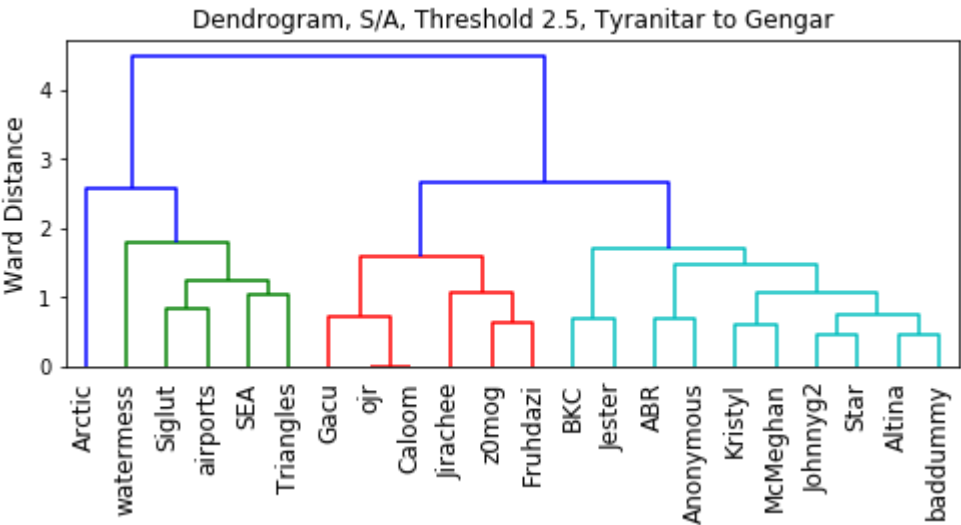
In [63]: ### Find dendrogram for interested tiers
# Interested tier spans are written in a list of lists called segments
# Each entry of segments captures the span from the tier represented by the first to second index
# The indices correspond to tier divisions defined by the clusterIndicesMod
segments = [[0,2],[0,5],[5,8],[0,9]]
names = list(rank.columns)
Hlist = []
Dlist = []
thresholdSegments = [2.55,2,3,1]

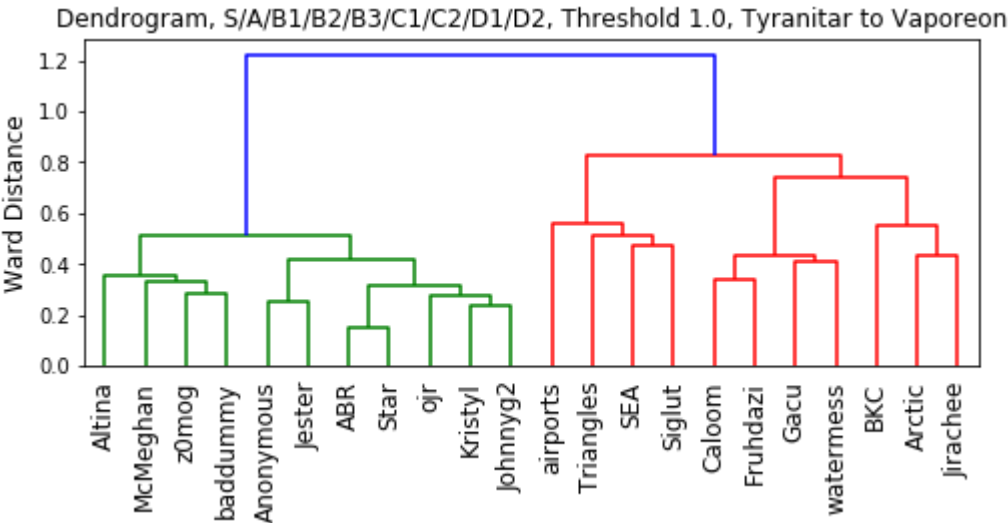
for i in range(len(segments)):
    pair = segments[i]
    fig = plt.figure(figsize=(8,3))
    ax = plt.axes()
    ax.set_ylabel('Ward Distance',fontsize=12)
    title = 'Dendrogram, ' + tierNames[pair[0]]
    for n in range(pair[0]+1,pair[1]):
        title += '/' + tierNames[n]
    title += ', Threshold ' + '{:.1f}'.format(thresholdSegments[i])
    title += ', ' + monList[clusterIndicesMod[pair[0]]] + ' to ' + monList[clusterIndicesMod[pair[1]]-1]

    ax.set_title(title,fontsize=12)
    Hlist += [cluster.hierarchy.ward(1-tArr[:, :, pair[1], pair[0]])]
    Dlist += [cluster.hierarchy.dendrogram(Hlist[-1],labels=names,color_threshold=thresholdSegments[i],leaf_rotation=90,leaf_font_size=12)]

```

```
C:\Users\lambj\anaconda3\lib\site-packages\scipy\cluster\hierarchy.py:830: Cl
usterWarning: scipy.cluster: The symmetric non-negative hollow observation ma
trix looks suspiciously like an uncondensed distance matrix
    return linkage(y, method='ward', metric='euclidean')
```





```

In [64]: ### Generates all statistical plots based on tier spans defined above in list segments

for n in range(len(segments)):
    pair = segments[n]
    order = Dlist[n]['leaves']
    rankMeanSubtractedCutoff = rankMeanSubtracted[:cutoff,order]
    namesSegment = [names[i] for i in order]
    # marker = itertools.cycle
    (( 'v', '^', '<', '>', '*', 'd', 'o', 'p', 'P', 'h', 'H', 'X', 'D'))
    marker = itertools.cycle(('v', '^', '<', '>', '*', 'd', 'X'))
    clustersSegmentUnordered = cluster.hierarchy.fcluster(Hlist[n], thresholdS
egments[n], criterion='distance')
    clustersSegment = clustersSegmentUnordered[order]
    clustersSegmentDividers = [0] + [clustersSegment[i+1]-clustersSegment[i] !
= 0 for i in range(len(clustersSegment)-1)] + [1]
    clustersSegmentIndices = np.insert(np.nonzero(clustersSegmentDividers)[0],
0,0)
    segmentLen = np.diff(clustersSegmentIndices)

    ### Plot Relative Ranks
    cArr = ['r'] + ['g'] + ['b'] + ['m'] + ['c']
    cArr.insert(np.argmax(segmentLen), 'y')
    color = itertools.cycle(tuple(cArr))
    delta = 0.2
    fig = plt.figure(figsize=(12,5))
    ax = plt.axes()
    ax.grid(zorder=0)
    offset = lambda p: transforms.ScaledTranslation(p/72.,0, plt.gcf().dpi_sca
le_trans)
    rankMeanSubtractedSegmentCutoff = []
    rankStdSubtractedSegmentCutoff = []
    monListCutoff = list(monList[:cutoff])
    for p in range(0,len(segmentLen)):
        offsetNum = 5*((p-len(clustersSegmentIndices)+2)/(len(clustersSegmentI
ndices)-1)+0.5)
        rankMeanSubtractedSegment = rankMeanSubtractedCutoff[:,clustersSegment
Indices[p]:clustersSegmentIndices[p+1]]
        trans = plt.gca().transData
        # ax.scatter(list(monList[:Dcut]),list(-np.mean(rankMeanSubtractedADca
mp,axis=1)),err=list(np.std(rankMeanSubtractedADcamp,axis=1)),zorder=1,label=n
amesAD[p],marker = next(marker),s=200,alpha=0.5,color=cArrAD[p])
        rankMeanSubtractedSegmentCutoff += [-np.nanmean(rankMeanSubtractedSegm
ent,axis=1)]
        rankStdSubtractedSegmentCutoff += [np.nanstd(rankMeanSubtractedSegmen
t,axis=1)/np.sqrt(segmentLen[p]-1)]
        if segmentLen[p] < 4:
            continue
        ax.errorbar(monListCutoff,list(rankMeanSubtractedSegmentCutoff[p]),lis
t(rankStdSubtractedSegmentCutoff[p]),zorder=1,label=namesSegment[clustersSegme
ntIndices[p]],
                    linestyle='None',marker=next(marker),markersize=10,alpha=
0.5,color=next(color),transform=trans+offset(offsetNum))

    ax.set_ylim(ax.get_ylim())
    ax.set_xlim(ax.get_xlim())

```



```

    ax.plot([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.5],[-99,99],color='k')
    ax.fill([-99,-99,clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.5],[-99,99,99,-99],'k',alpha=0.1)
    ax.plot([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.5],[-99,99],color='k')
    ax.fill([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.5,cutoff*2,cutoff*2],[-99,99,99,-99],'k',alpha=0.1)
    plt.xticks(rotation=90,size=18)
    ax.legend(loc='upper center',bbox_to_anchor=(1.1,1),ncol=1)
    plt.ylabel('Relative Rank In Favor',fontsize=20)
    title = year + ' ' + gen + ' VR Relative Statistics: ' + tierNames[pair[0]]
    for m in range(pair[0]+1,pair[1]):
        title += '/' + tierNames[m]
    plt.title(title,fontsize=20)

    ### Plot Ranking Correlations
    tClustered = np.array([[tArr[i,j,pair[1],pair[0]] for j in order] for i in order])
    for ii in range(0,tClustered.shape[0]):
        tClustered[ii,ii] = np.nan
    fig = plt.figure(figsize=(12,12))
    ax = plt.gca()
    h = ax.imshow(tClustered,interpolation='none',cmap='viridis')
    fig.colorbar(h,ax=ax,shrink=0.8)
    title = 'Ranking Correlation'
    title = tierNames[pair[0]]
    for m in range(pair[0]+1,pair[1]):
        title += '/' + tierNames[m]
    title += ', ' + monList[clusterIndicesMod[pair[0]]] + ' to ' + monList[clusterIndicesMod[pair[1]]-1]

    ax.set_title(title,fontsize=18)
    ax.set_xticks(range(0,tClustered.shape[1]))
    ax.set_yticks(range(0,tClustered.shape[0]))
    ax.set_xticklabels([names[i] for i in order],fontsize=18)
    ax.set_yticklabels([names[i] for i in order],fontsize=18)
    plt.setp(ax.get_xticklabels(), rotation=90, ha="right",va="center",rotation_mode="anchor");
    for ii in range(0,tClustered.shape[0]):
        for jj in range(0,tClustered.shape[1]):
            if ii != jj:
                text = ax.text(ii, jj, int(100*tClustered[ii, jj]),
                               ha="center", va="center", color="w",fontsize=12)

    ### Plot Sorted Differences by Z Score
    rankMeanSubtractedCutoffOtherSegments = []
    rankStdSubtractedCutoffOtherSegments = []
    rankStdDifference = []
    orderSegment = []
    seismic = cm.get_cmap('coolwarm', 32)
    dataNormalizer = colors.Normalize()
    for p in range(len(segmentLen)):
        rankSubtractedCutoffOtherSegments = rankMeanSubtractedCutoff[:,np.array(np.concatenate((range(clustersSegmentIndices[p]),range(clustersSegmentIndices

```

```

s[p+1],len(order)))]),dtype=int)]
    rankMeanSubtractedCutoffOtherSegments += [-np.nanmean(rankSubtractedCutoffOtherSegments,axis=1)]
    rankStdSubtractedCutoffOtherSegments += [np.nanstd(rankSubtractedCutoffOtherSegments,axis=1)/np.sqrt(len(order)-segmentLen[p]-1)]
    rankStdDifference += [np.sqrt(rankStdSubtractedSegmentCutoff[p]**2 + rankStdSubtractedCutoffOtherSegments[p]**2)]
    zSegment = (rankMeanSubtractedSegmentCutoff[p] - rankMeanSubtractedCutoffOtherSegments[p]) / (rankStdDifference[p]+1e-7)

    orderSegment += [sorted(range(len(monListCutoff)),reverse=True,key=lambda x:np.abs(zSegment[x]))]
    if segmentLen[p] < 4:
        continue
    # order by z score
    fig = plt.figure(figsize=(12,5))
    ax = plt.axes()
    ax.grid(zorder=0)
    offset = lambda p: transforms.ScaledTranslation(p/72.,0, plt.gcf().dpi*_scale_trans)
    limit = min(max(abs(zSegment[abs(zSegment)<20]))*1.05,5)
    ax.bar(range(len(monListCutoff)),zSegment[orderSegment[p]],zorder=1,color=seismic(zSegment[orderSegment[p]]/(2)+0.5))
    ax.set_xticks(range(len(monListCutoff)))
    ax.set_xticklabels([monListCutoff[i] for i in orderSegment[p]],fontsize=360/cutoff)
    plt.xticks(rotation=90,size=16)
    ax.set_ylabel('Z Score of Relative Rank in Favor',fontsize=15)
    title = tierNames[pair[0]]
    for m in range(pair[0]+1,pair[1]):
        title += '/' + tierNames[m]
    title += ', Threshold ' + '{:.1f}'.format(thresholdSegments[i])
    title += ', ' + namesSegment[clustersSegmentIndices[p]] + ' camp'
    title += ', ' + str(segmentLen[p]) + '/' + str(len(order)) + ' voters'
    ax.set_title(title,fontsize=18)
    ax.set_ylim([-limit,limit])

    # order by rank
    fig = plt.figure(figsize=(12,5))
    ax = plt.axes()
    ax.grid(zorder=0)
    offset = lambda p: transforms.ScaledTranslation(p/72.,0, plt.gcf().dpi*_scale_trans)
    limit = min(max(abs(zSegment[abs(zSegment)<20]))*1.05,5)
    ax.bar(range(len(monListCutoff)),zSegment,zorder=1,color=seismic(zSegment/(2)+0.5))
    ax.set_xticks(range(len(monListCutoff)))
    ax.set_xticklabels(monListCutoff,fontsize=360/cutoff)
    plt.xticks(rotation=90,size=16)
    ax.set_ylabel('Z Score of Relative Rank in Favor',fontsize=15)
    title = tierNames[pair[0]]
    for m in range(pair[0]+1,pair[1]):
        title += '/' + tierNames[m]
    title += ', Threshold ' + '{:.1f}'.format(thresholdSegments[i])
    title += ', ' + namesSegment[clustersSegmentIndices[p]] + ' camp'
    title += ', ' + str(segmentLen[p]) + '/' + str(len(order)) + ' voters'
    ax.set_title(title,fontsize=18)

```

```

ax.set_ylim([-limit,limit])
ax.set_xlim(ax.get_xlim())
ax.plot([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.
5],[-99,99],color='k')
ax.fill([-99,-99,clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair
[0]]-0.5],[-99,99,99,-99],'k',alpha=0.1)
ax.plot([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.
5],[-99,99],color='k')
ax.fill([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.
5,cutoff*2,cutoff*2],[-99,99,99,-99],'k',alpha=0.1)

### Plot Dissimilarity Matrix
dissimSegment = np.zeros((cutoff,cutoff))
totalCountSegment = np.zeros((cutoff,cutoff))
for m1 in range(cutoff):
    totalCountSegment[m1,m1] = 1
for m1 in range(cutoff):
    for m2 in range(cutoff):
        if m1 == m2:
            continue

        for n1 in range(clustersSegmentIndices[p],clustersSegmentIndices[p+1]):
            for n2 in np.array(np.concatenate((range(clustersSegmentIndices[p]),range(clustersSegmentIndices[p+1],len(order)))),dtype=int):
                sign1 = np.sign(rankvalSub[m1,order[n1]]-rankvalSub[m
2,order[n1]])
                sign2 = np.sign(rankvalSub[m1,order[n2]]-rankvalSub[m
2,order[n2]])
                corr = sign1 * sign2
                if corr != 0:
                    dissimSegment[m1,m2] += - (1 - corr)/2 * sign1
                    totalCountSegment[m1,m2] += 1
dissimSegment = dissimSegment / totalCountSegment
for m1 in range(cutoff):
    dissimSegment[m1,m1] = 0
fig = plt.figure(figsize=(12,12))
ax = plt.gca()
h = ax.imshow(dissimSegment,interpolation='none',cmap='seismic',vmin=-
1,vmax=1)
ax.plot([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[1]]-0.
5],[clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.5],color='k')
ax.plot([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[1]]-0.
5],[clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.5],color='k')
ax.plot([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.
5],[clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[1]]-0.5],color='k')
ax.plot([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.
5],[clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[1]]-0.5],color='k')
ax.fill([-0.5,-0.5,clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.5],[-0.5,cutoff-0.5,cutoff-0.5,-0.5],'k',alpha=0.1)
ax.fill([clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.
5,cutoff-0.5,cutoff-0.5],[-0.5,cutoff-0.5,cutoff-0.5,-0.5],'k',alpha=0.1)
ax.fill([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.
5,clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.5],[clusterIndicesMod[pair[1]]-0.5,cutoff-0.5,cutoff-0.5,clusterIndicesMod[pair[1]]-0.5],color='k',alpha=0.1)
ax.fill([clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.

```

```

5,clusterIndicesMod[pair[1]]-0.5,clusterIndicesMod[pair[1]]-0.5],[-0.5,clusterIndicesMod[pair[0]]-0.5,clusterIndicesMod[pair[0]]-0.5,-0.5], 'k', alpha=0.1)

fig.colorbar(h,ax=ax,shrink=0.8)
title = 'Dissimilarity '
title += tierNames[pair[0]]
for m in range(pair[0]+1,pair[1]):
    title += '/' + tierNames[m]
title += ', Threshold ' + '{:.1f}'.format(thresholdSegments[i])
title += ', ' + str(segmentLen[p]) + '/' + str(len(order)) + ' voters
in '

title += namesSegment[clustersSegmentIndices[p]] + ' Camp'
title += ' \n Red = ' + namesSegment[clustersSegmentIndices[p]] + ' Ca
mp Favors Y; ' + ' Remainder Favors X\n Blue = ' + ' Remainder Favors Y; ' + n
amesSegment[clustersSegmentIndices[p]] + ' Camp Favors X'
ax.set_title(title,fontsize=18)
ax.set_xticks(range(0,dissimSegment.shape[1]))
ax.set_yticks(range(0,dissimSegment.shape[0]))
ax.set_xticklabels([monList[i] for i in range(cutoff)],fontsize=520/cu
toff)
ax.set_yticklabels([monList[i] for i in range(cutoff)],fontsize=520/cu
toff)
plt.setp(ax.get_xticklabels(), rotation=90, ha="right",va="center",rot
ation_mode="anchor");
for ii in range(0,dissimSegment.shape[0]):
    for jj in range(0,dissimSegment.shape[1]):
        if dissimSegment[ii,jj] != 0:
            text = ax.text(jj, ii, int(100*dissimSegment[ii, jj]),
                            ha="center", va="center", color="w",fontsiz
e=360/cutoff)

```

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:34: RuntimeWarning: invalid value encountered in true_divide

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:58: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:98: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:118: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:163: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:98: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:118: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:163: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:21: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:58: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:98: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

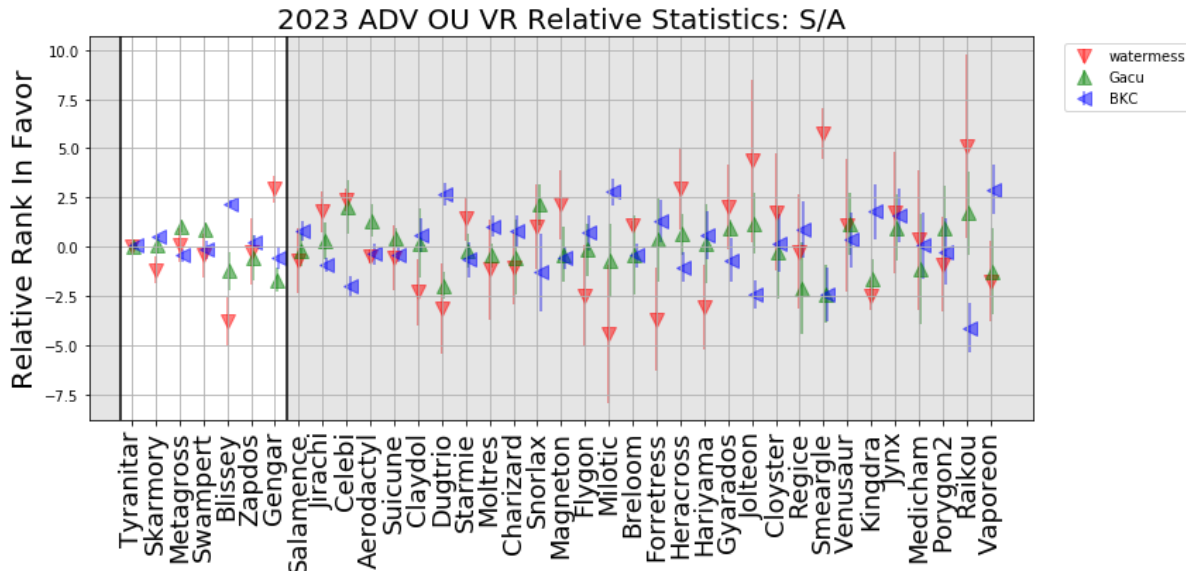
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:118: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParameter `figure.max_open_warning`).

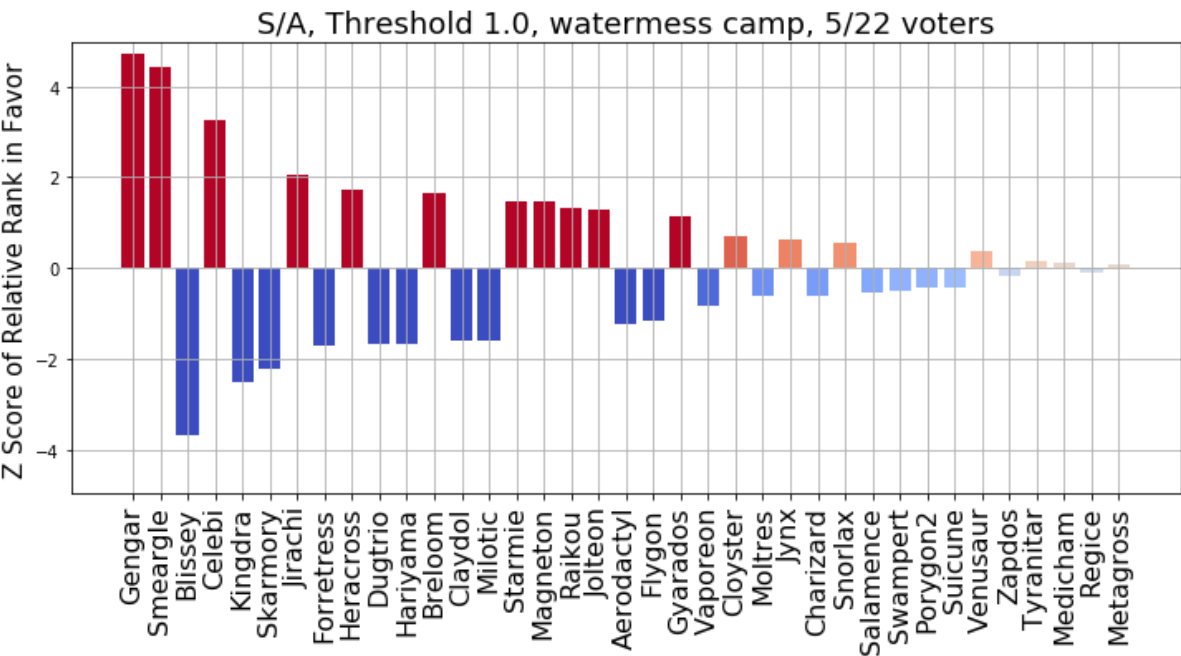
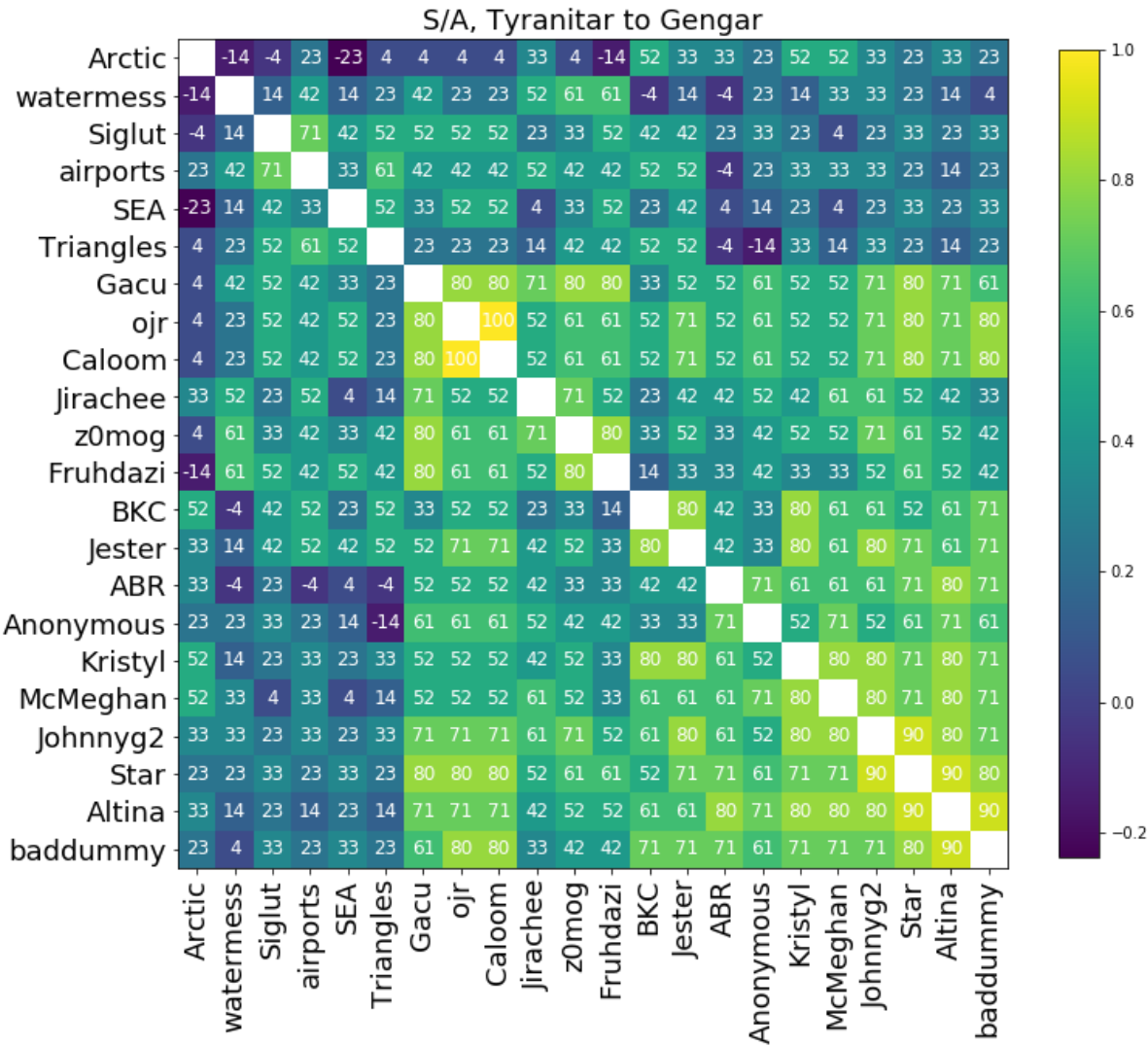
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:163: Runtime Warning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParam `figure.max_open_warning`).

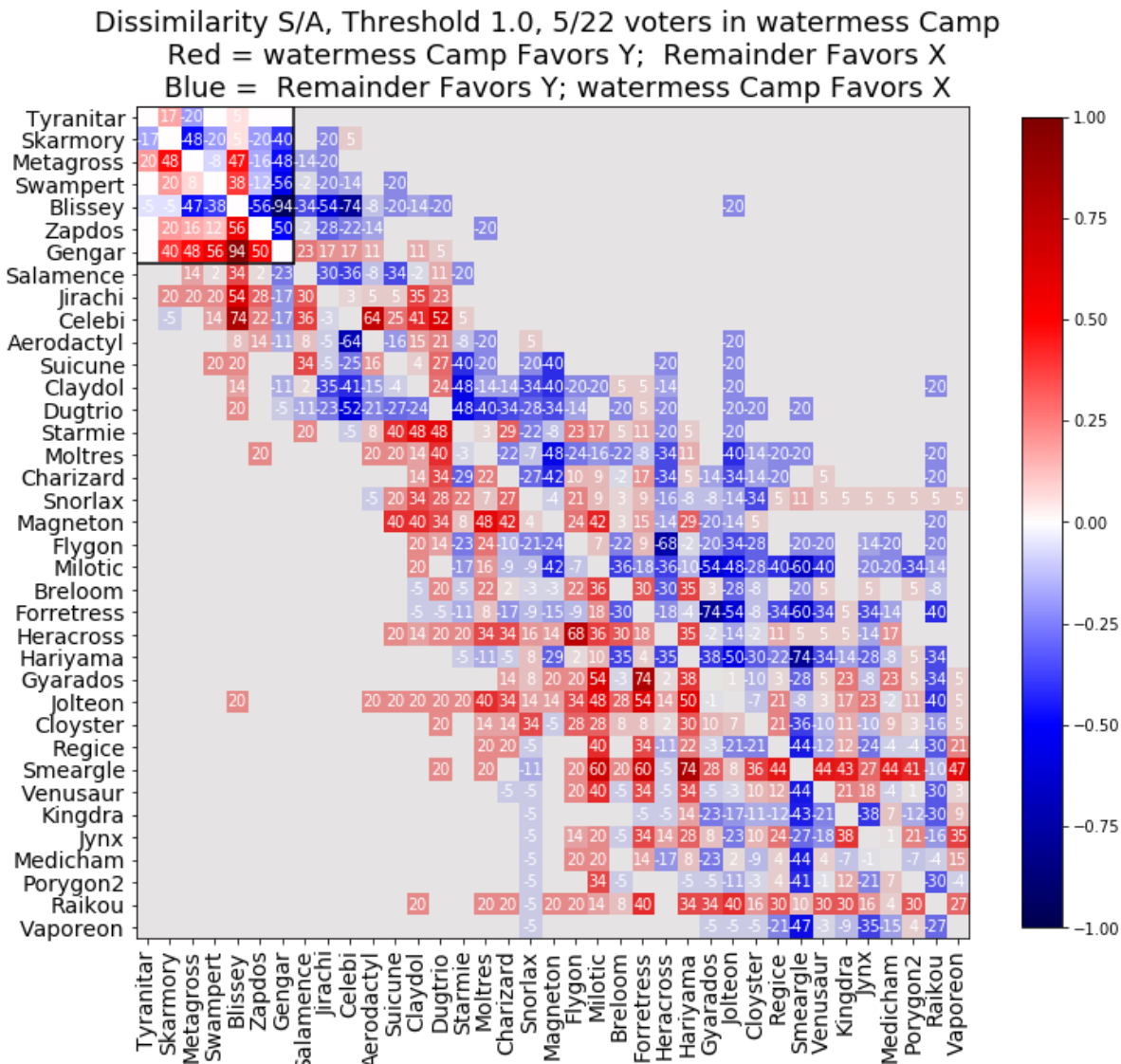
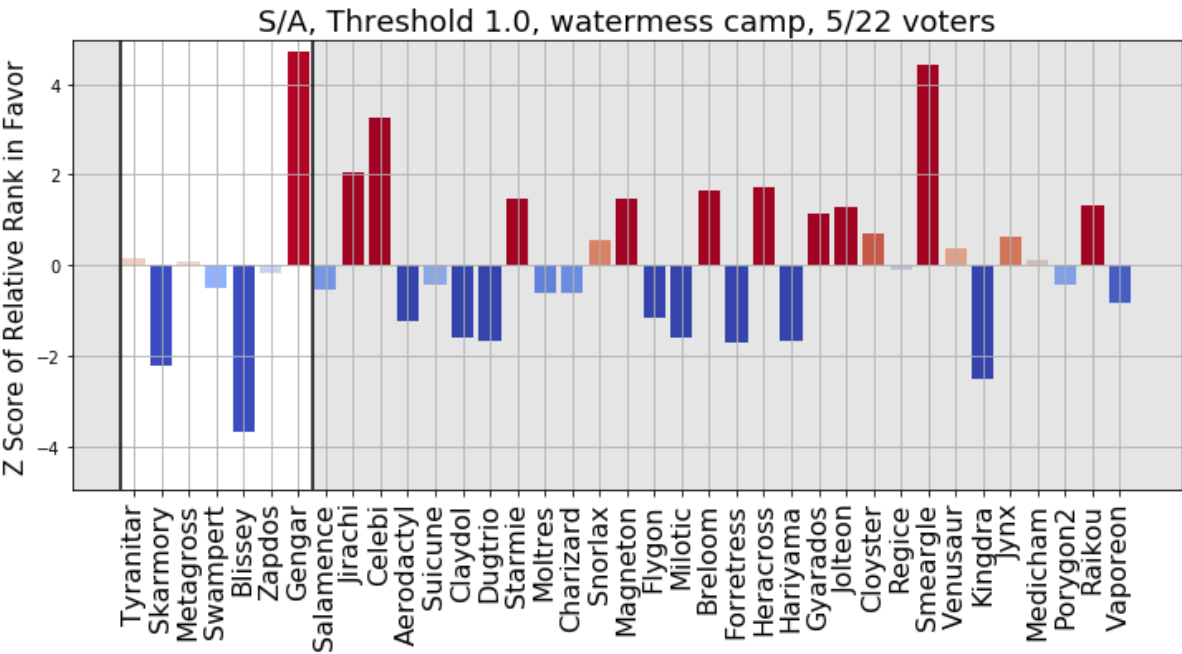
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:98: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParam `figure.max_open_warning`).

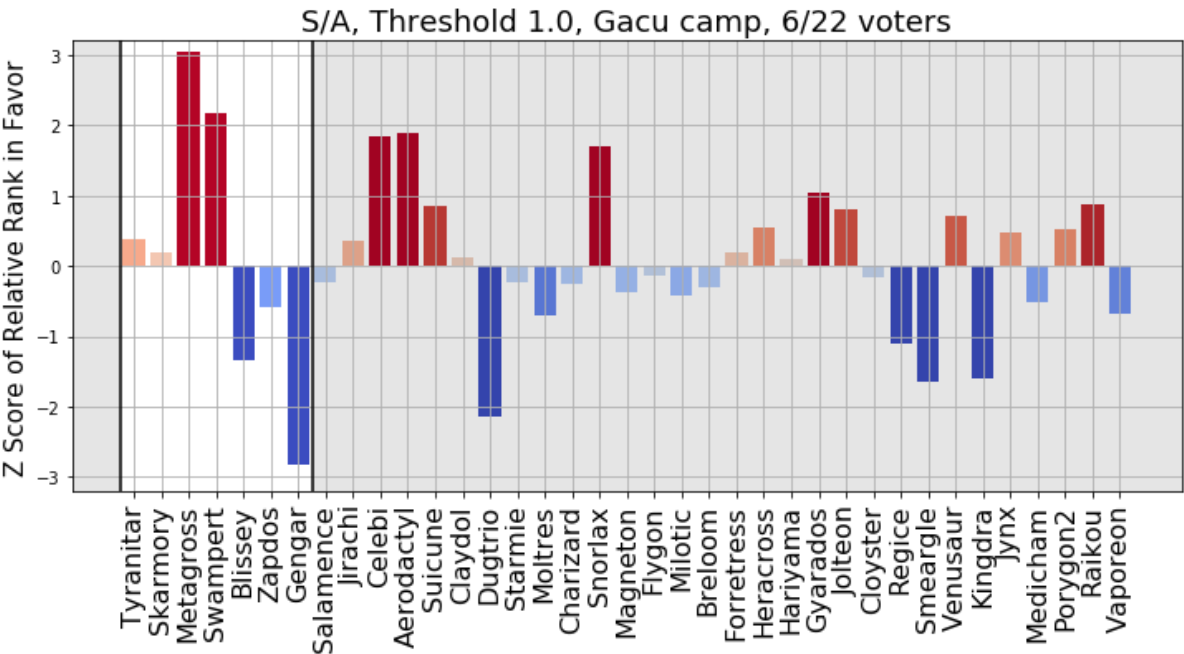
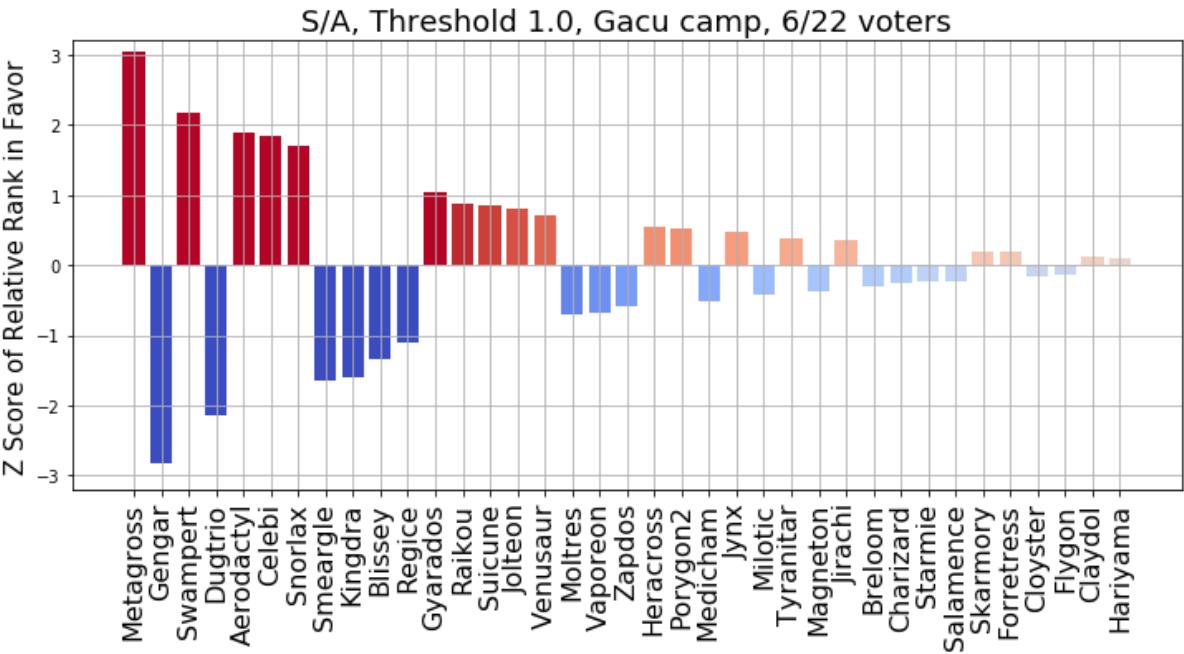
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:118: Runtime Warning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParam `figure.max_open_warning`).

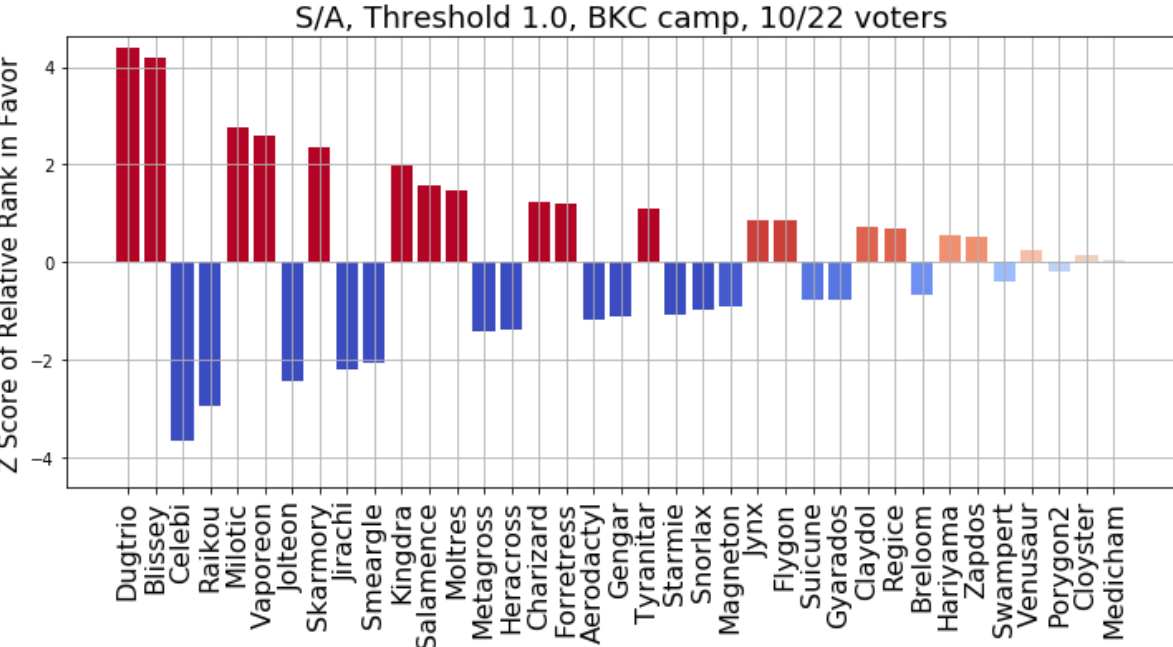
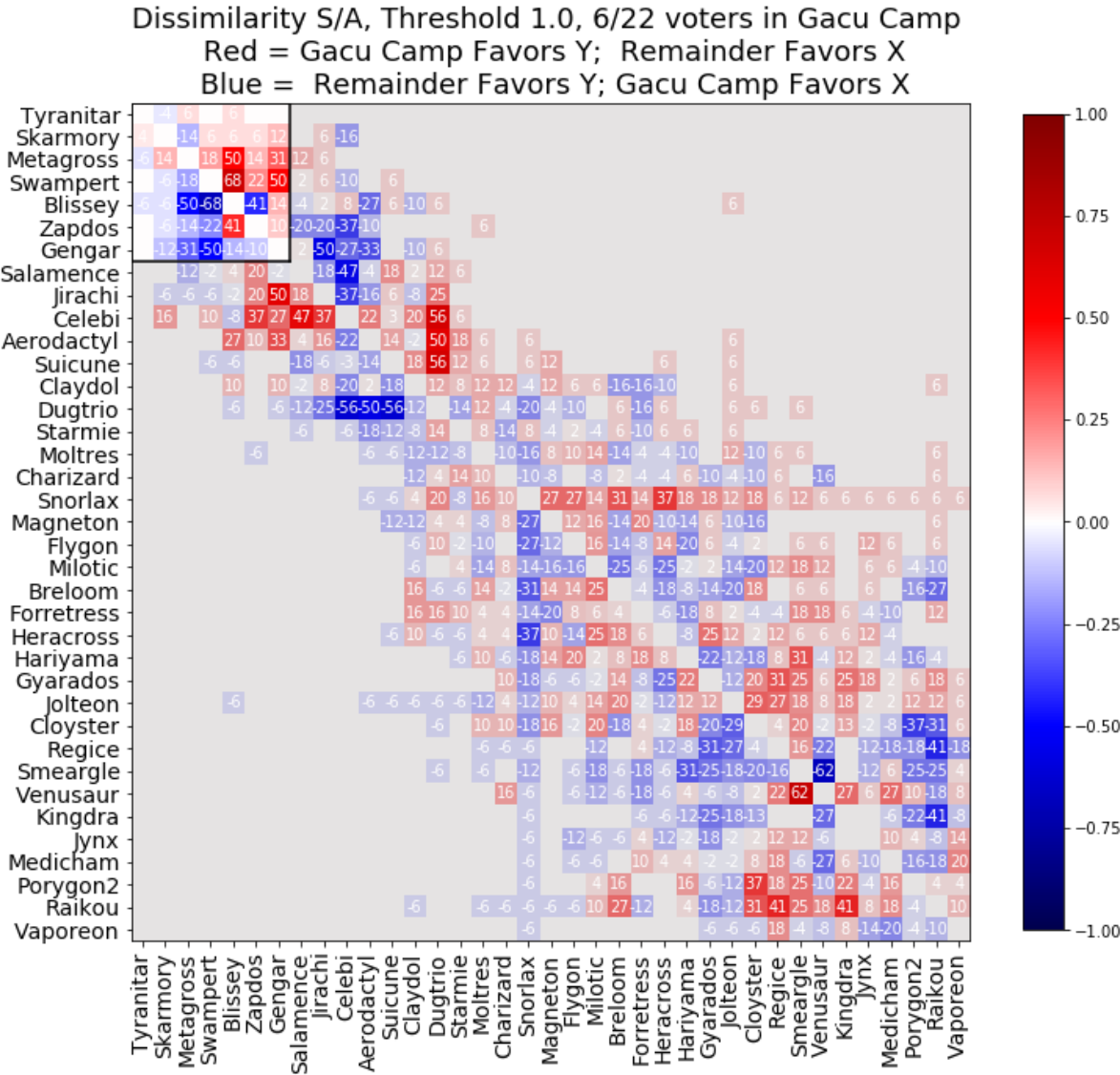
C:\Users\lambj\anaconda3\lib\site-packages\ipykernel_launcher.py:163: Runtime Warning: More than 20 figures have been opened. Figures created through the pyplot interface (`matplotlib.pyplot.figure`) are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParam `figure.max_open_warning`).

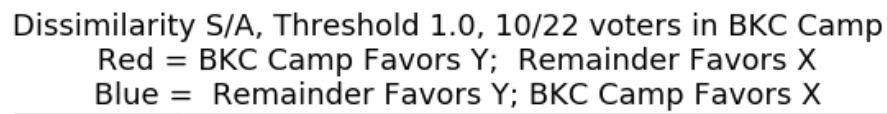


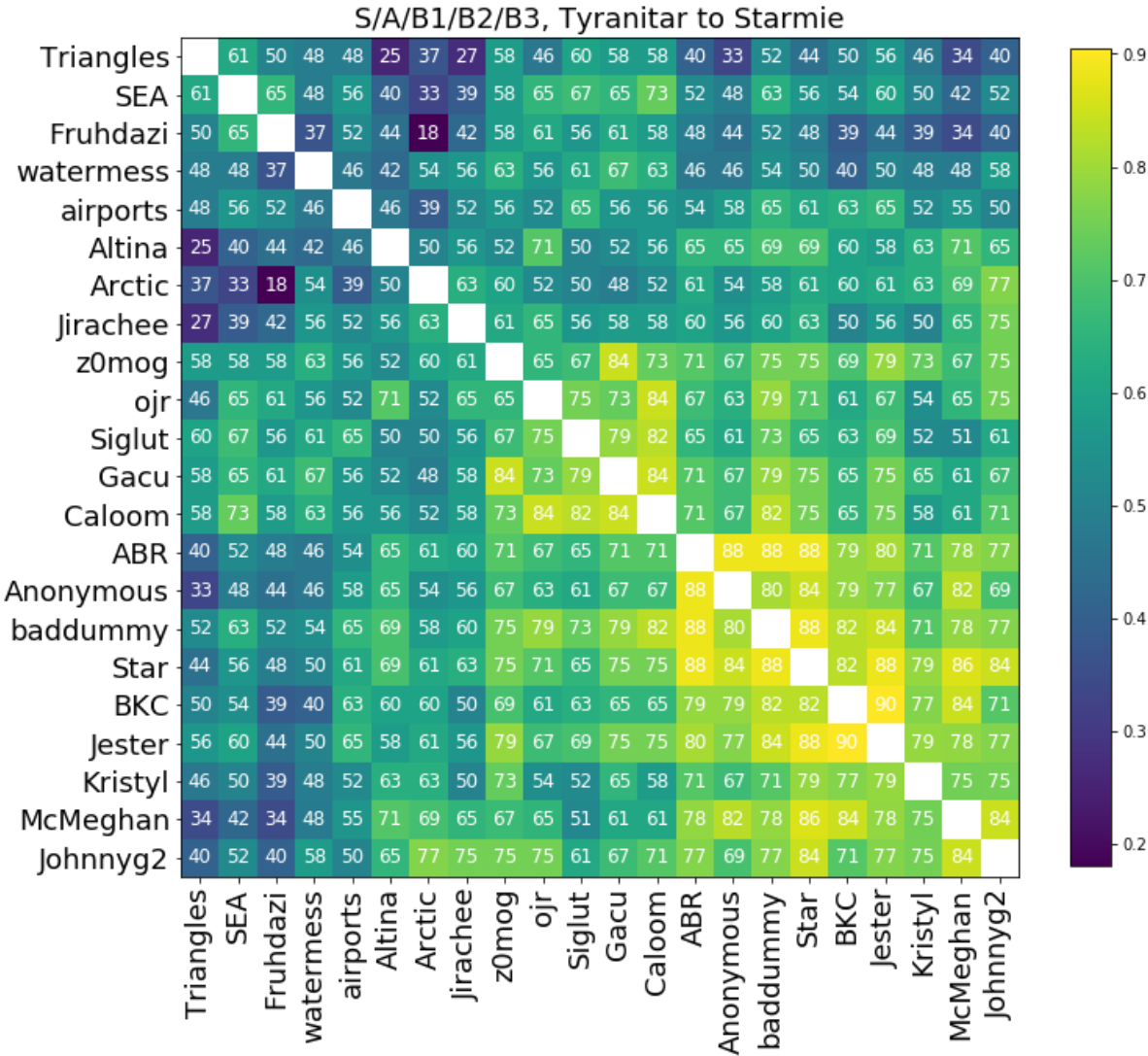
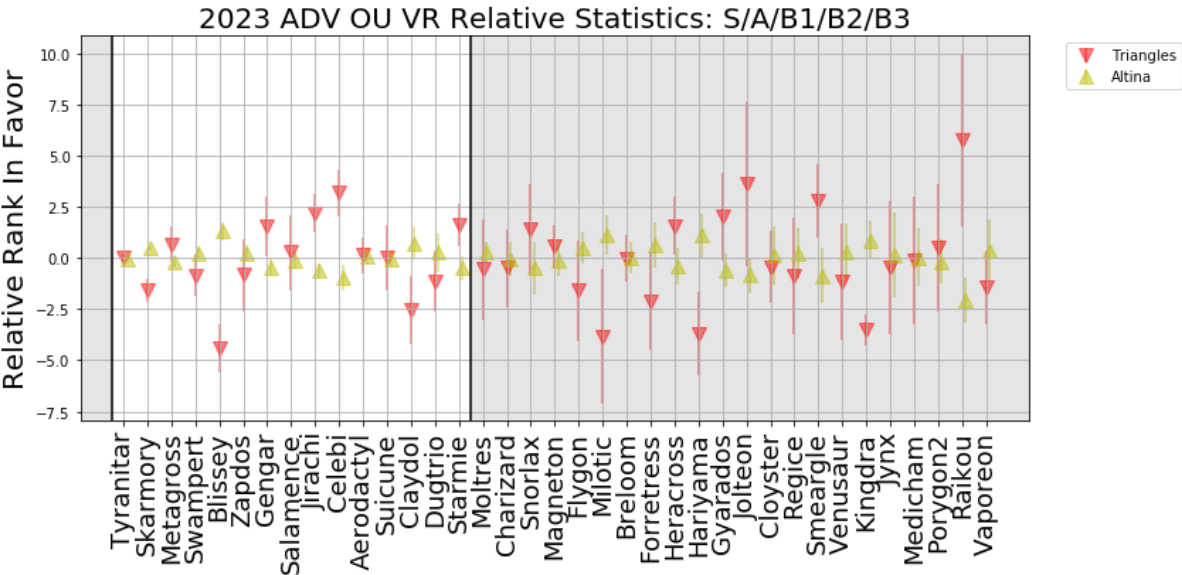


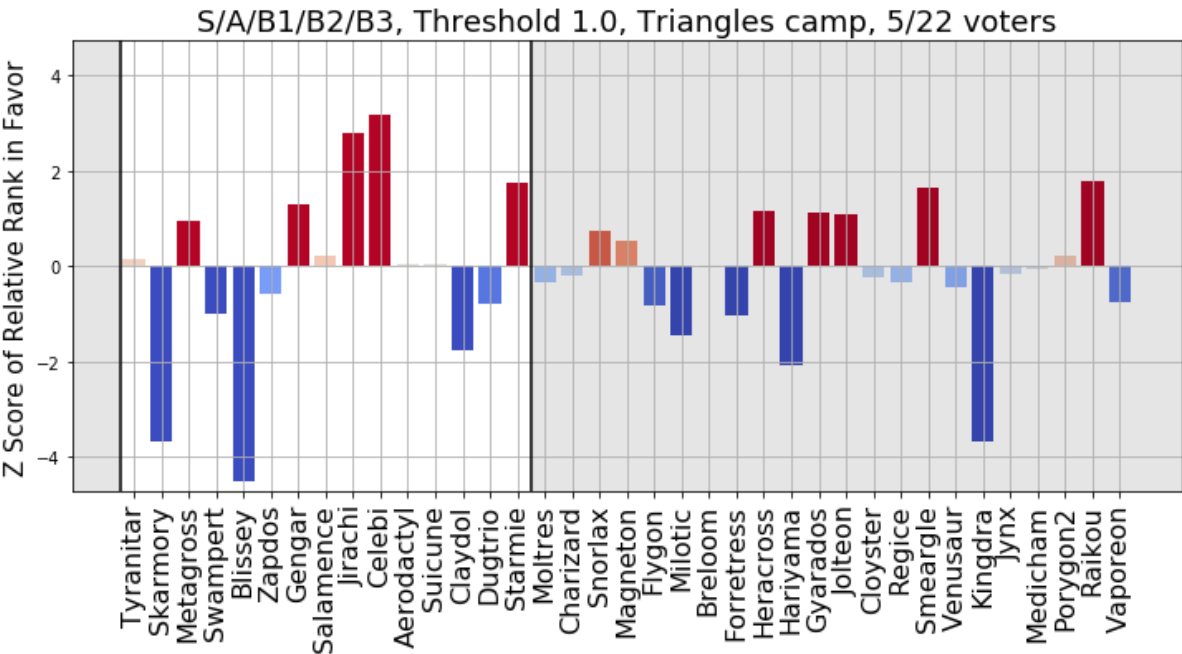
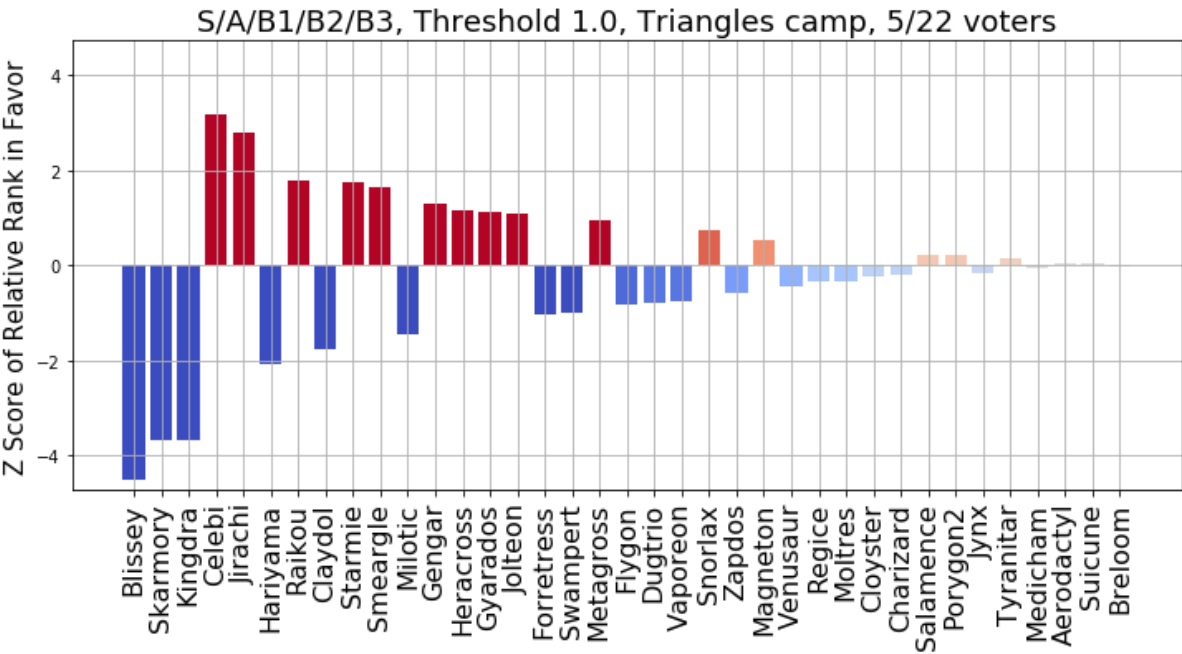


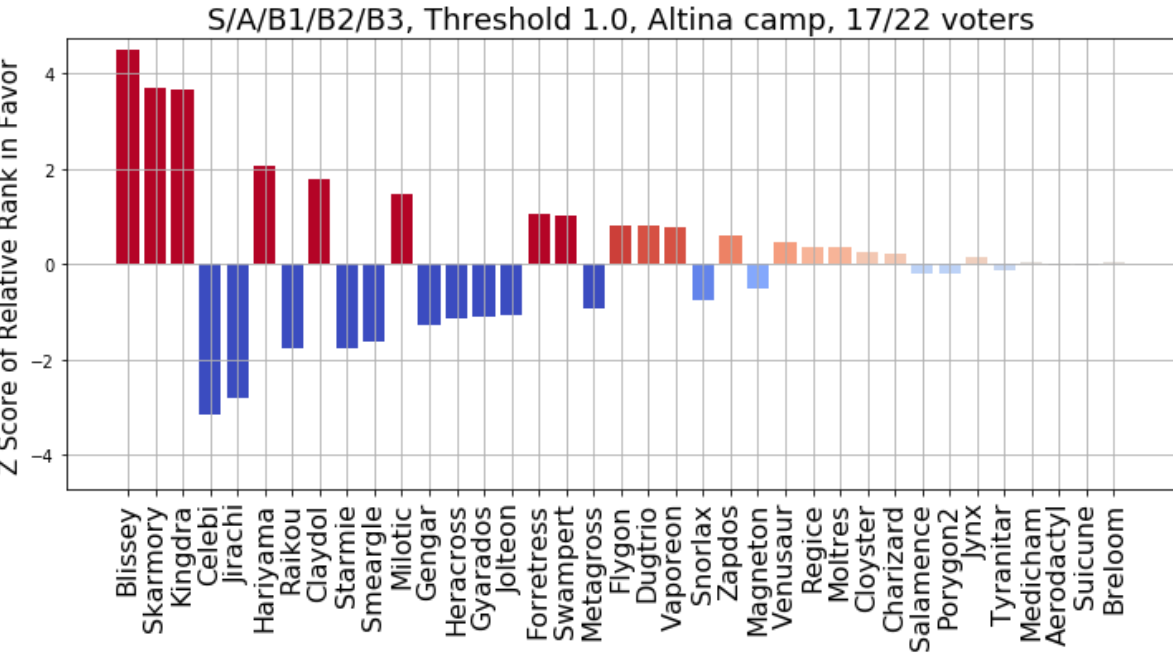
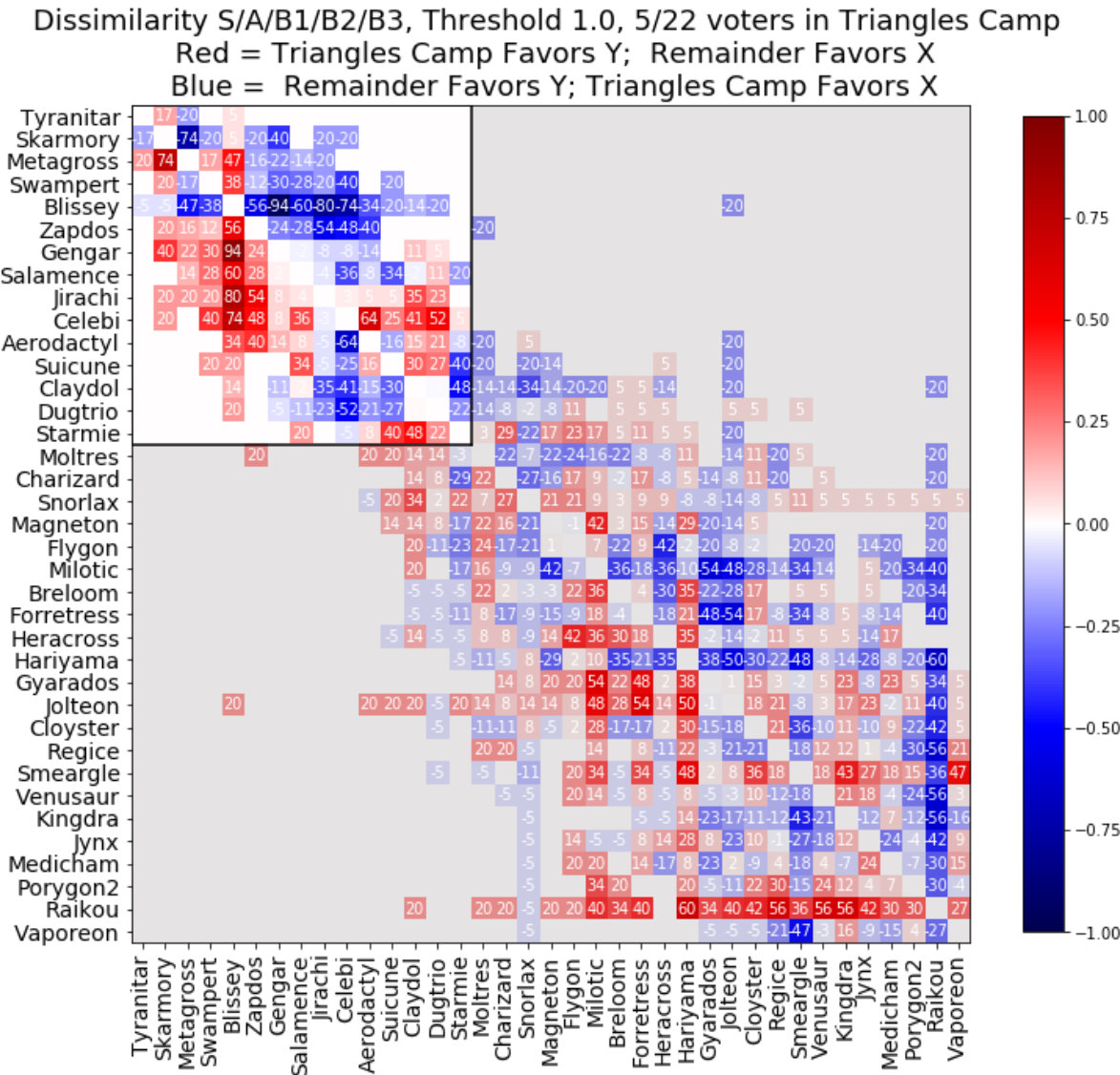


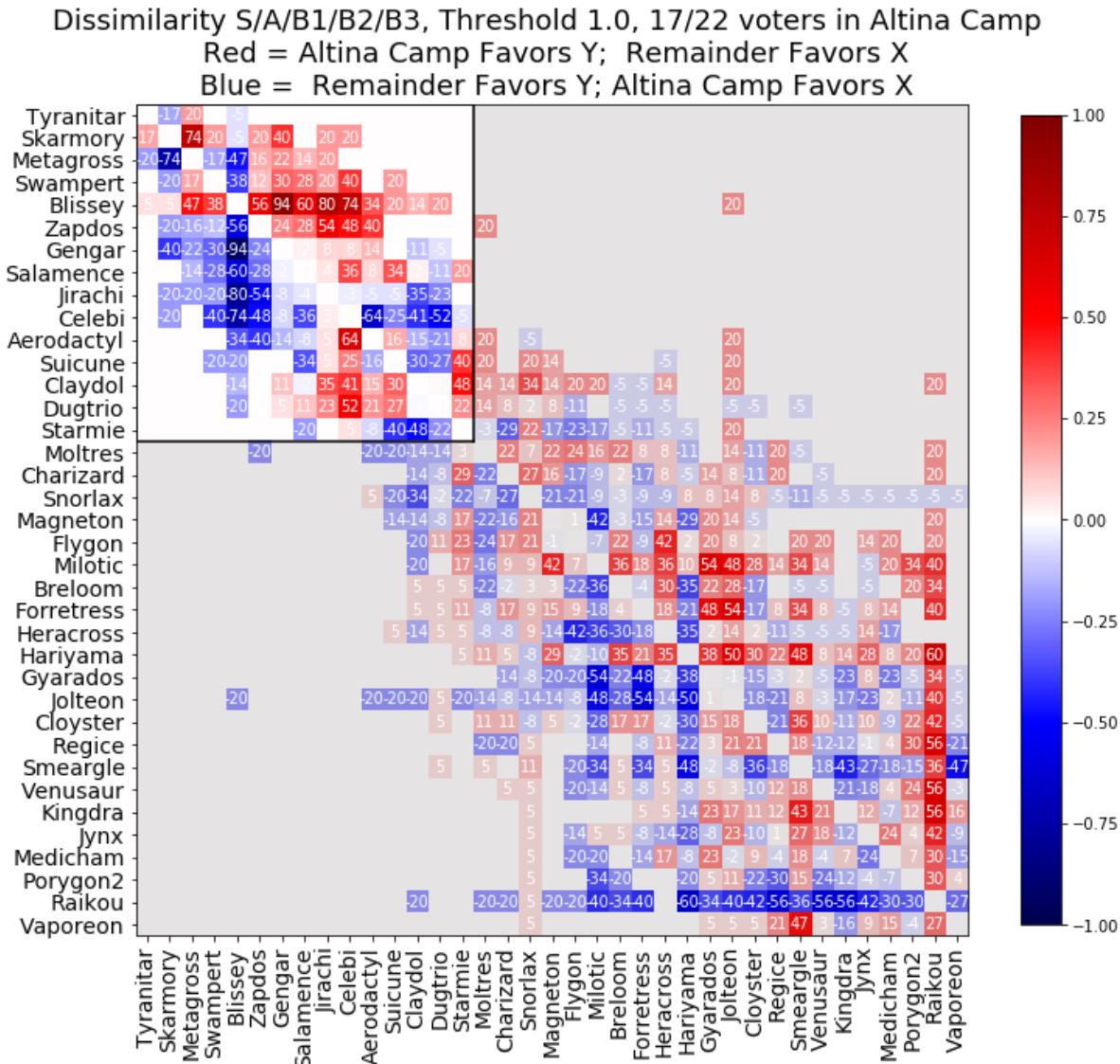
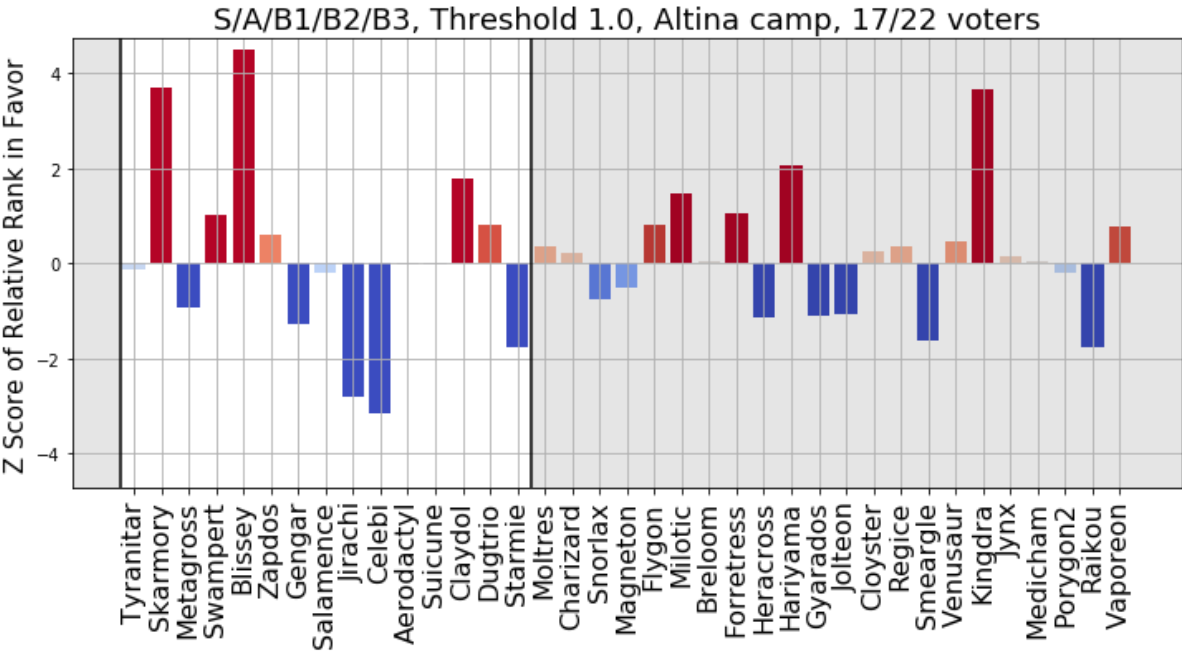


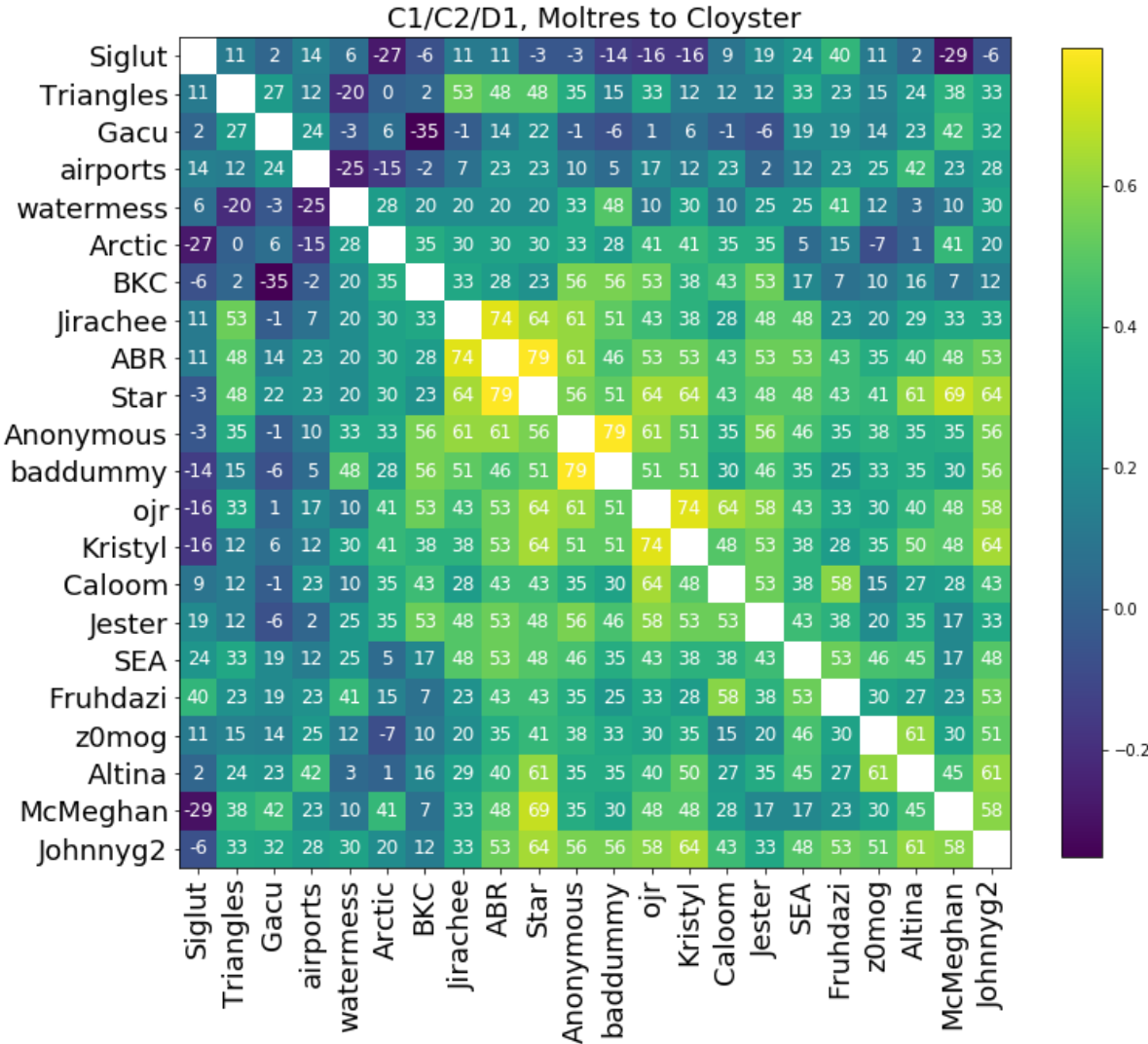
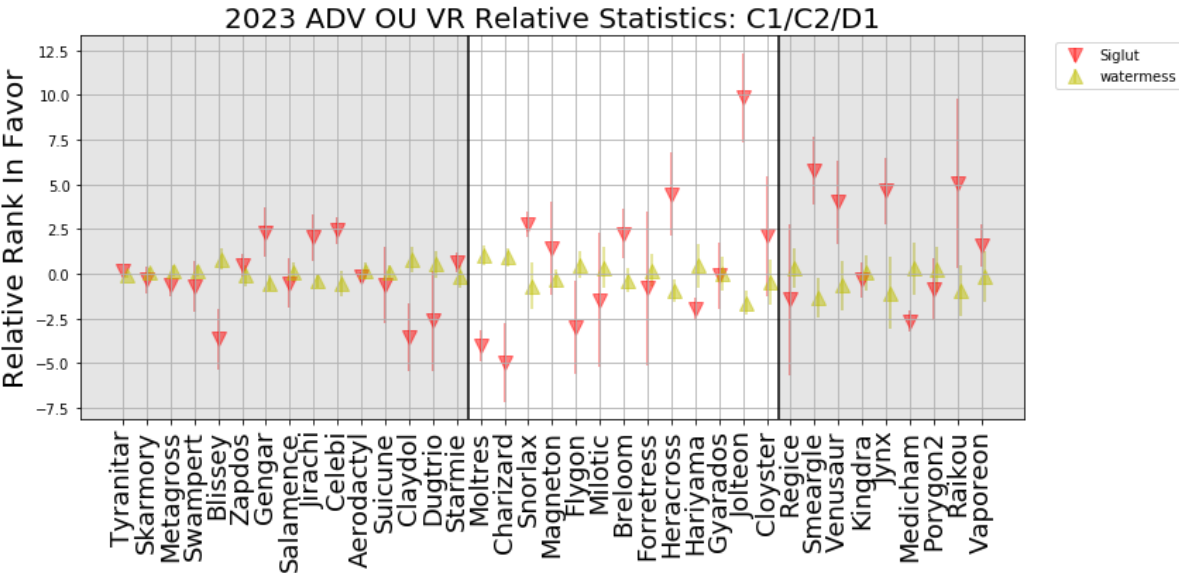


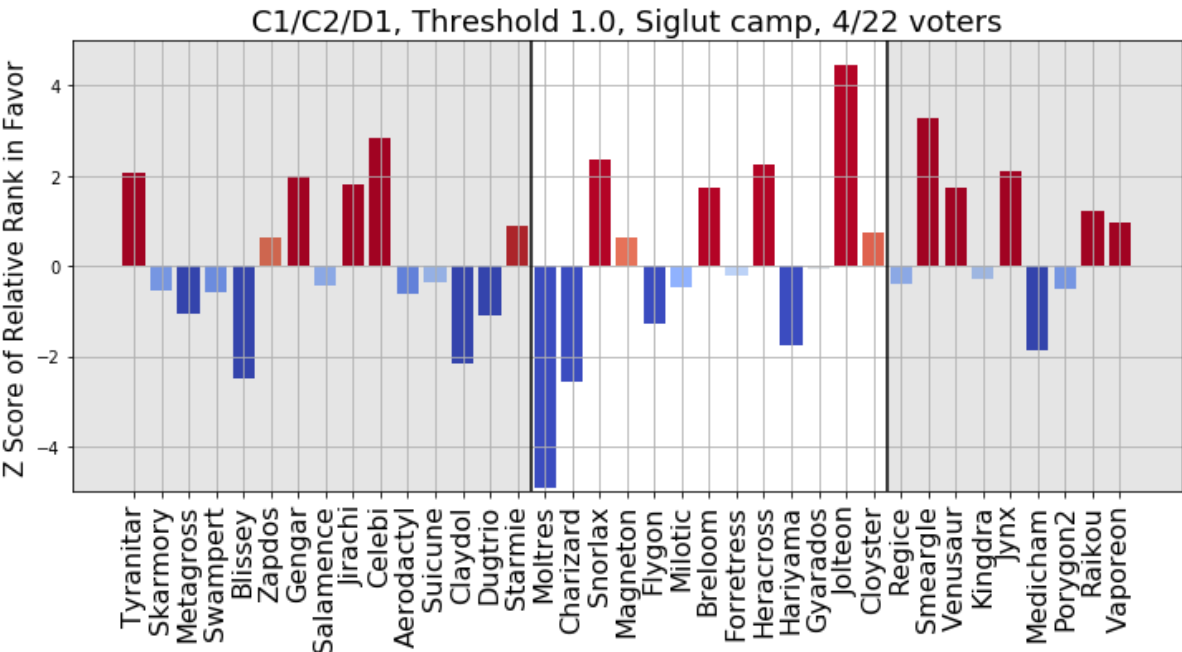
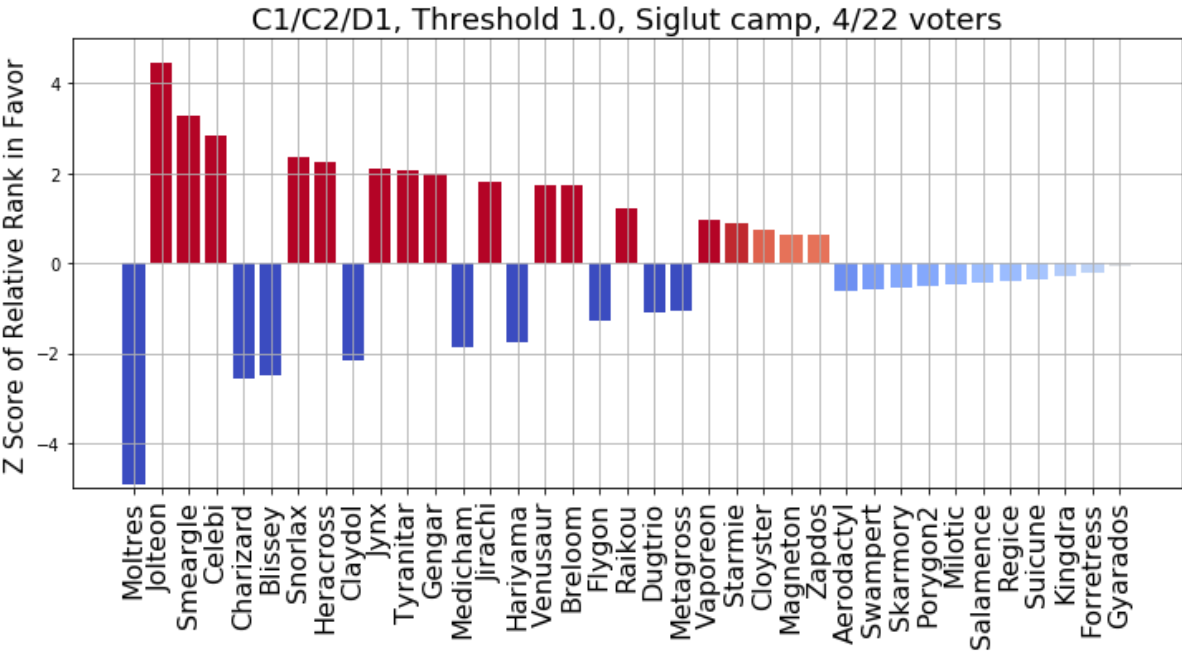








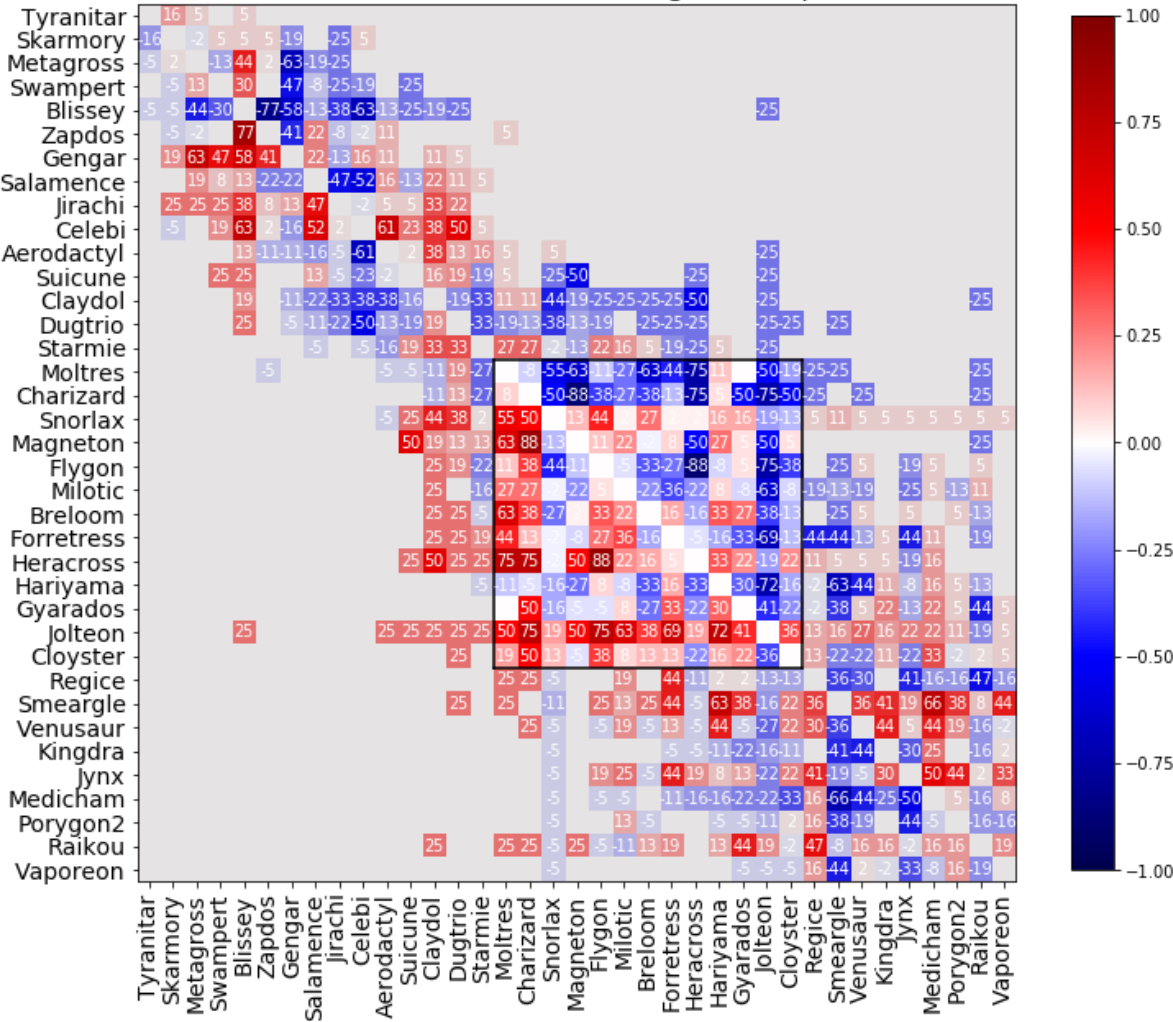




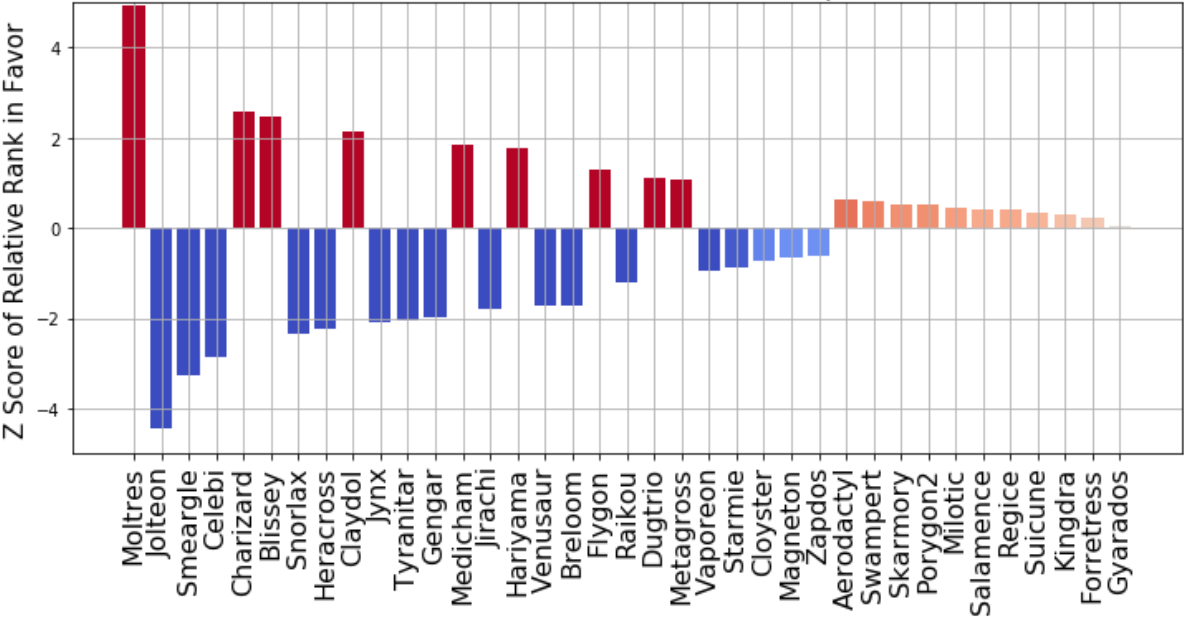
Dissimilarity C1/C2/D1, Threshold 1.0, 4/22 voters in Siglut Camp

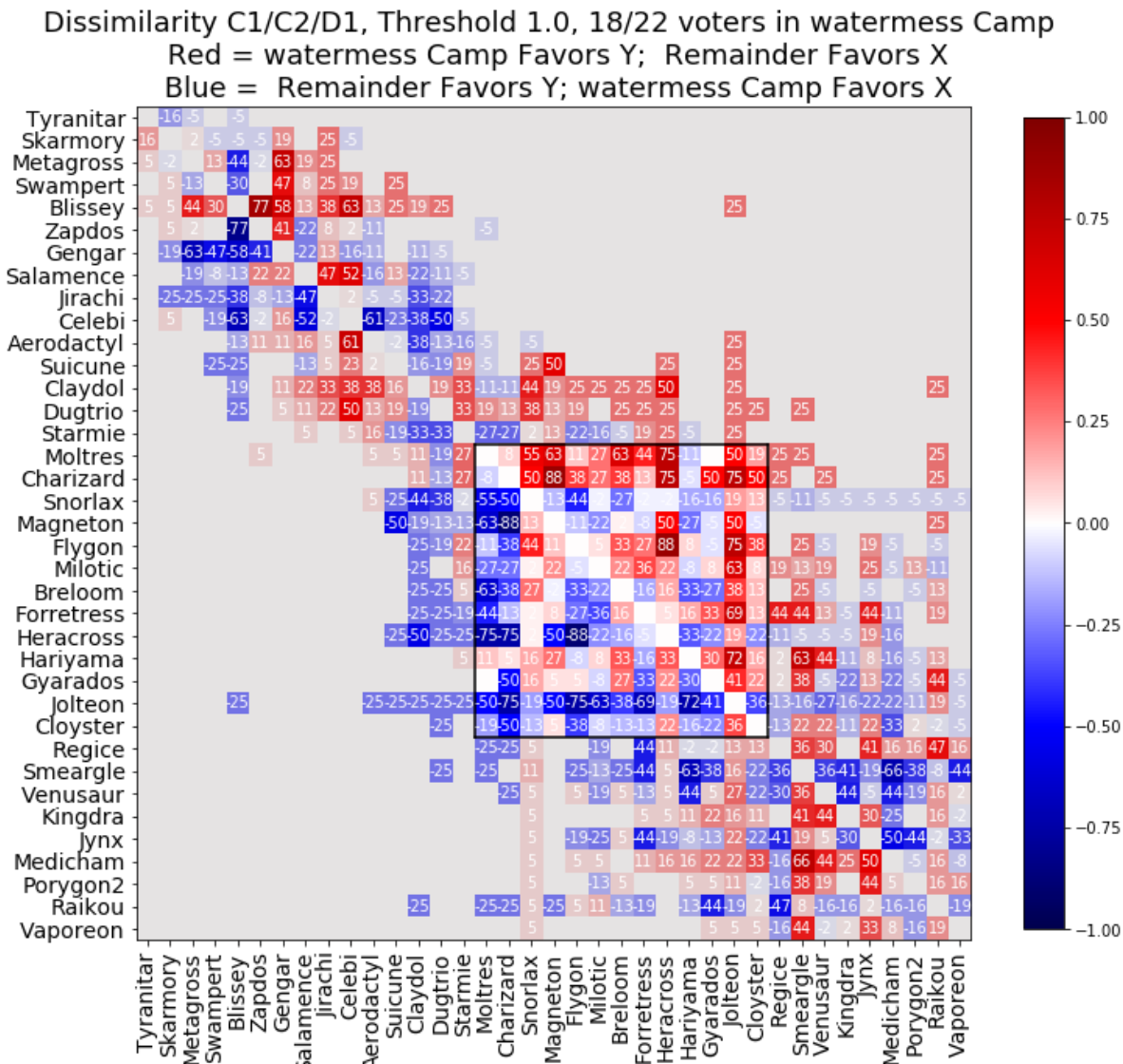
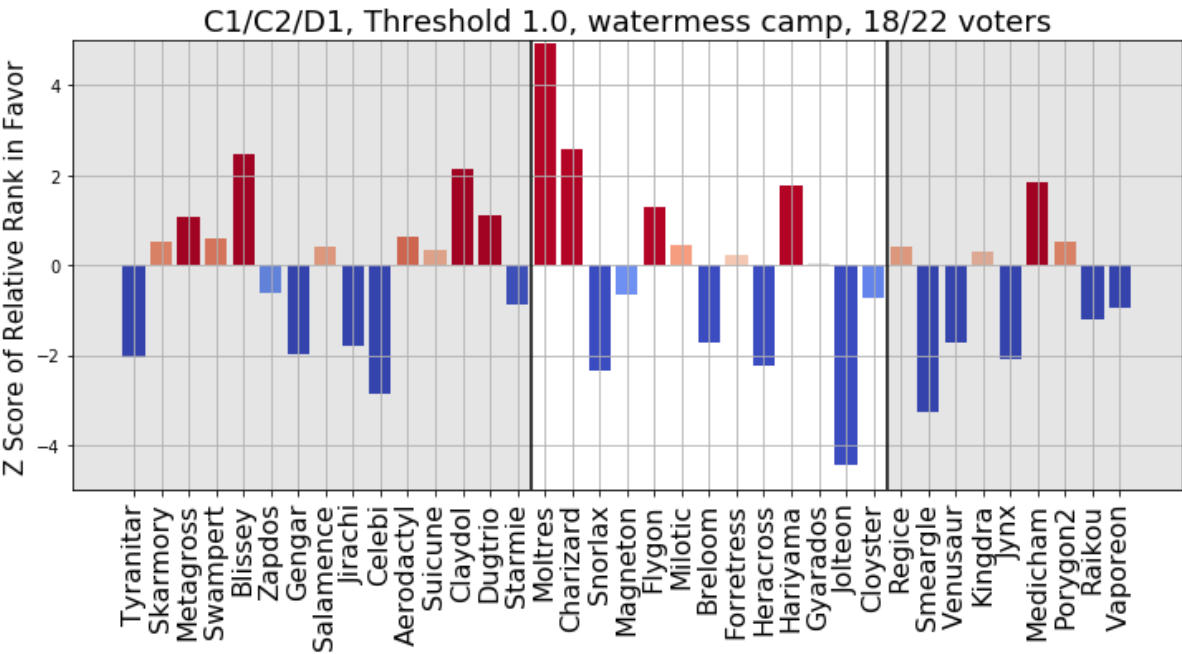
Red = Siglut Camp Favors Y; Remainder Favors X

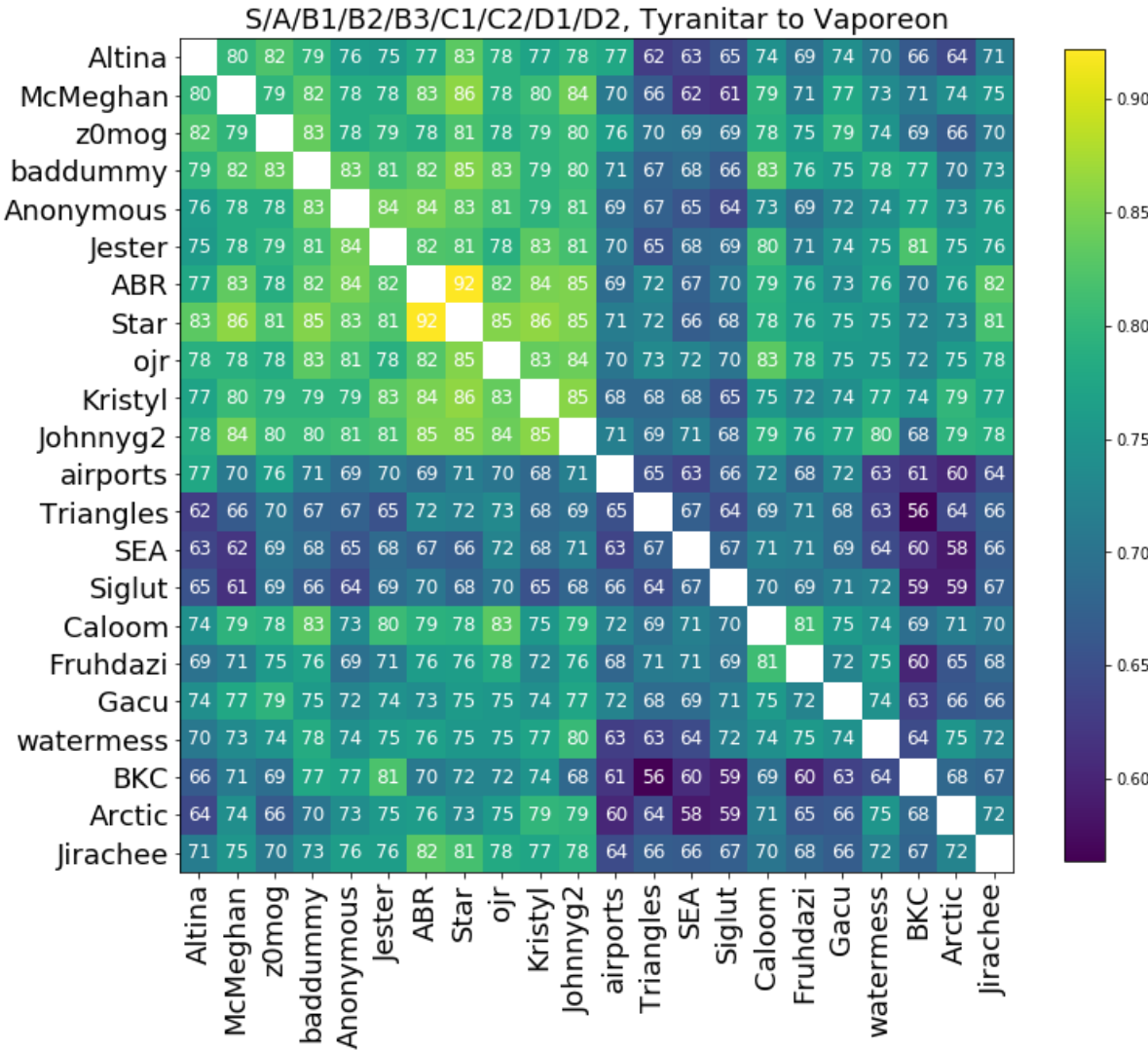
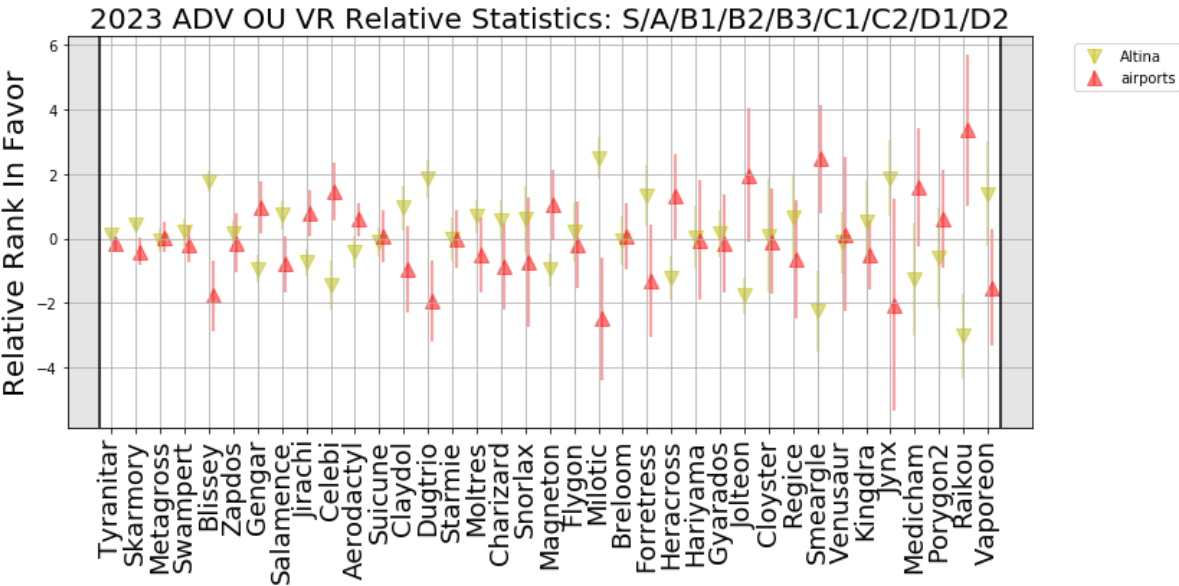
Blue = Remainder Favors Y; Siglut Camp Favors X

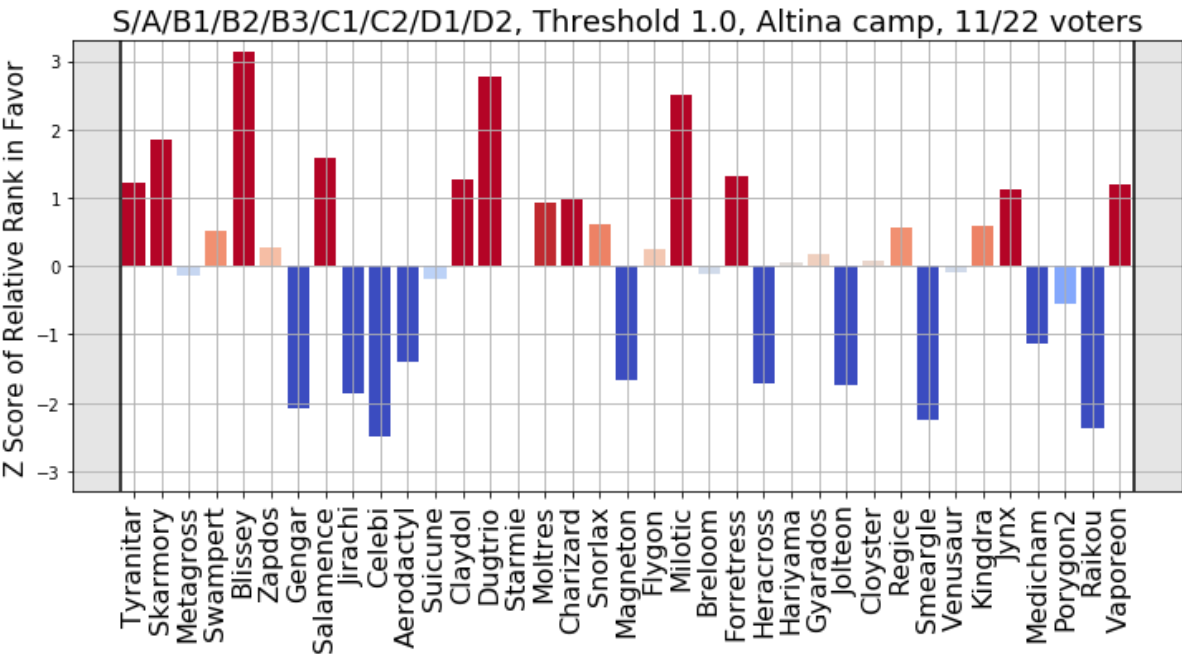
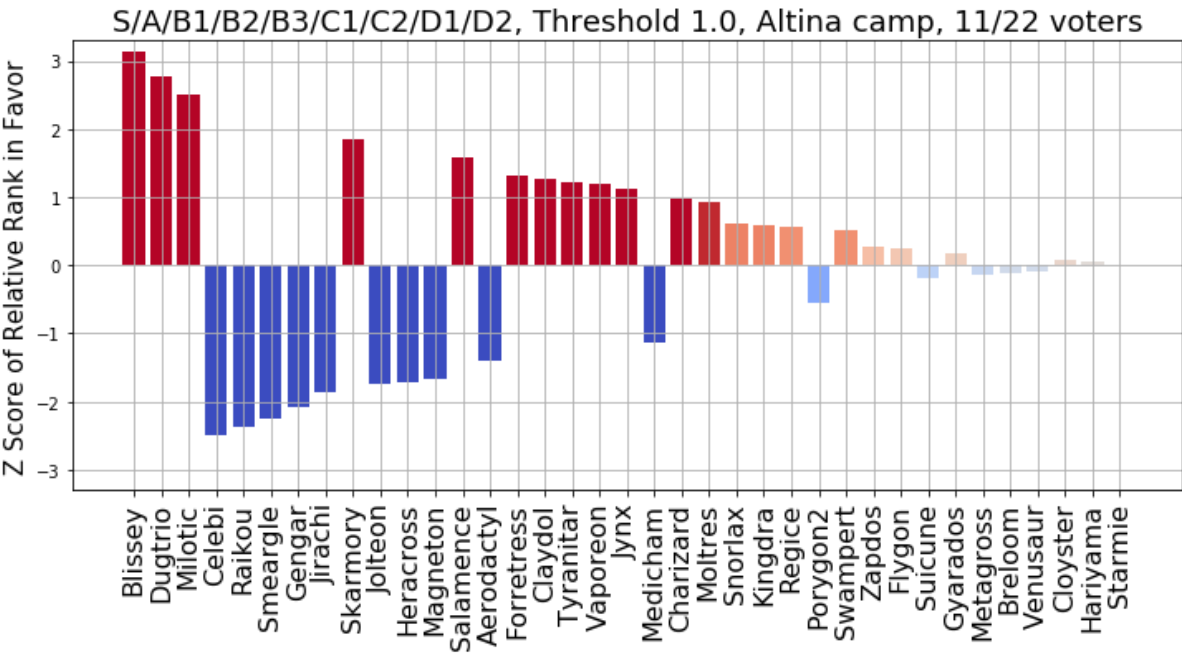


C1/C2/D1, Threshold 1.0, watermess camp, 18/22 voters



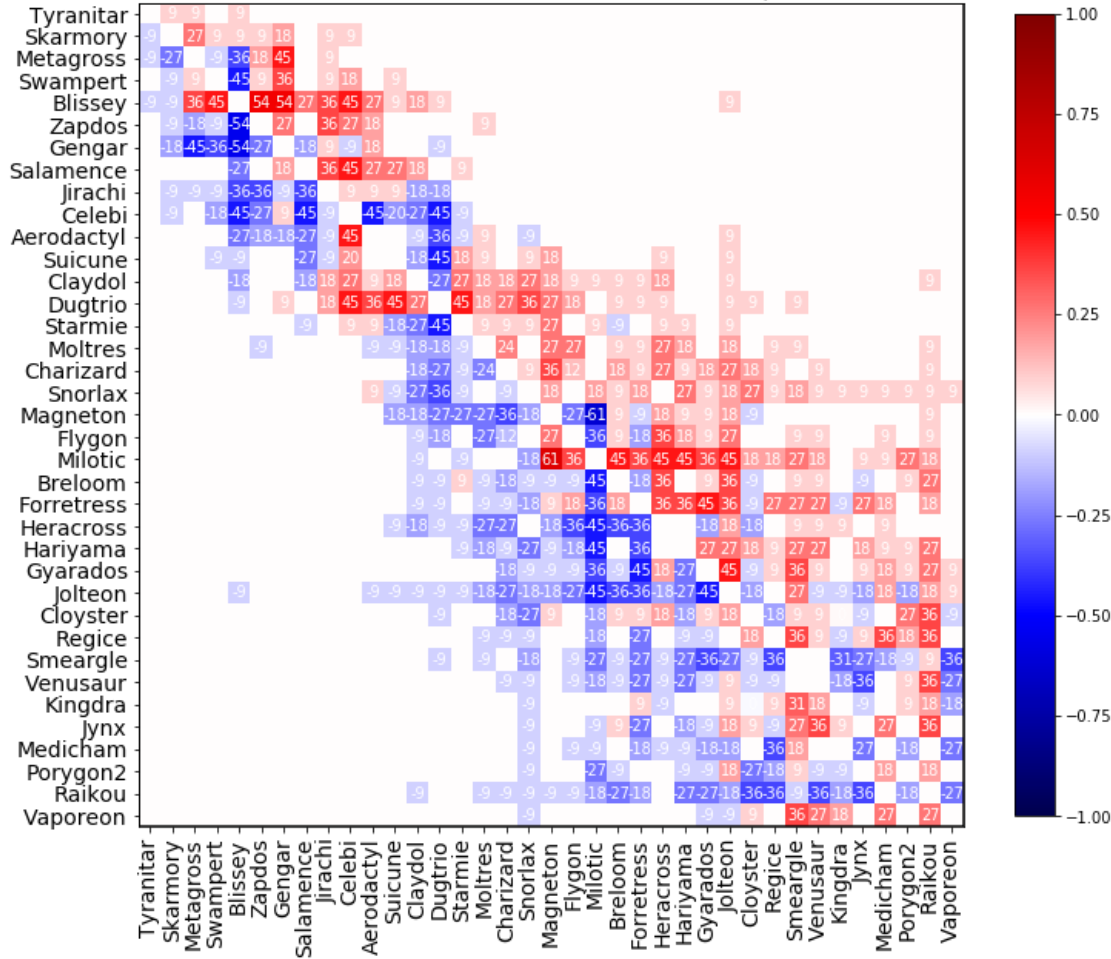




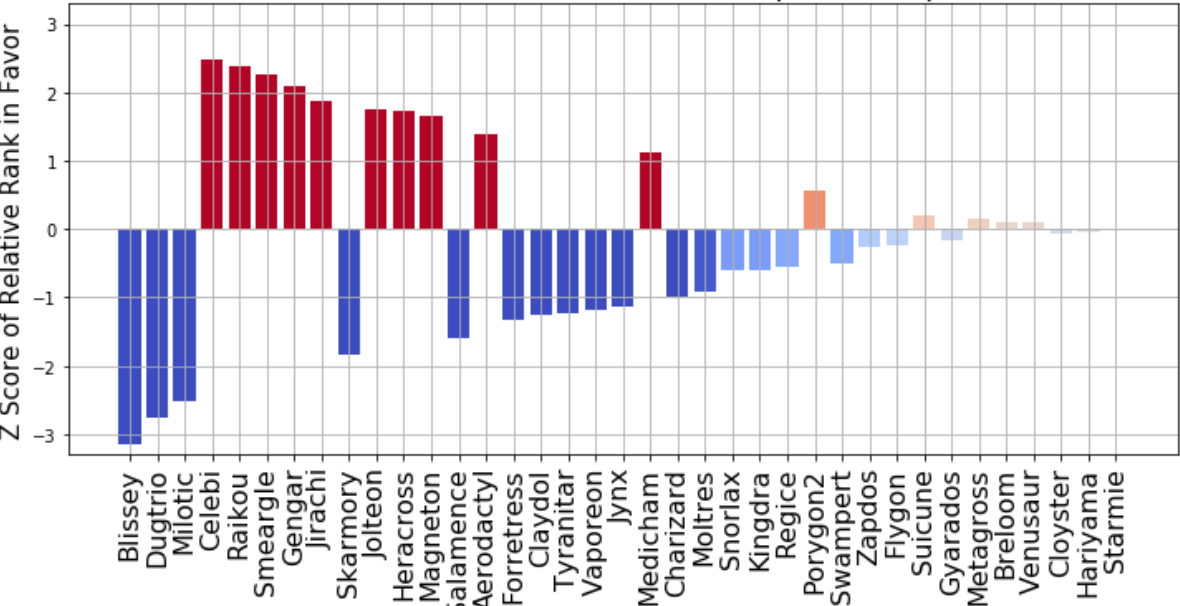


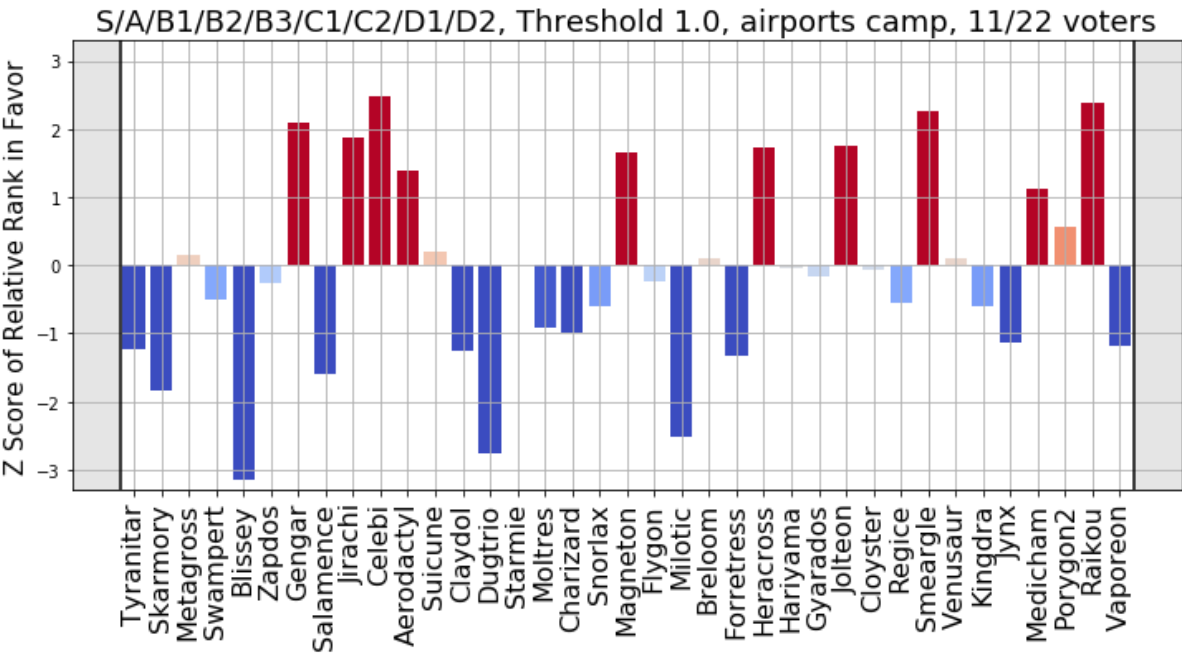
Dissimilarity S/A/B1/B2/B3/C1/C2/D1/D2, Threshold 1.0, 11/22 voters in Altina Camp

Red = Altina Camp Favors Y; Remainder Favors X
Blue = Remainder Favors Y; Altina Camp Favors X



S/A/B1/B2/B3/C1/C2/D1/D2, Threshold 1.0, airports camp, 11/22 voters

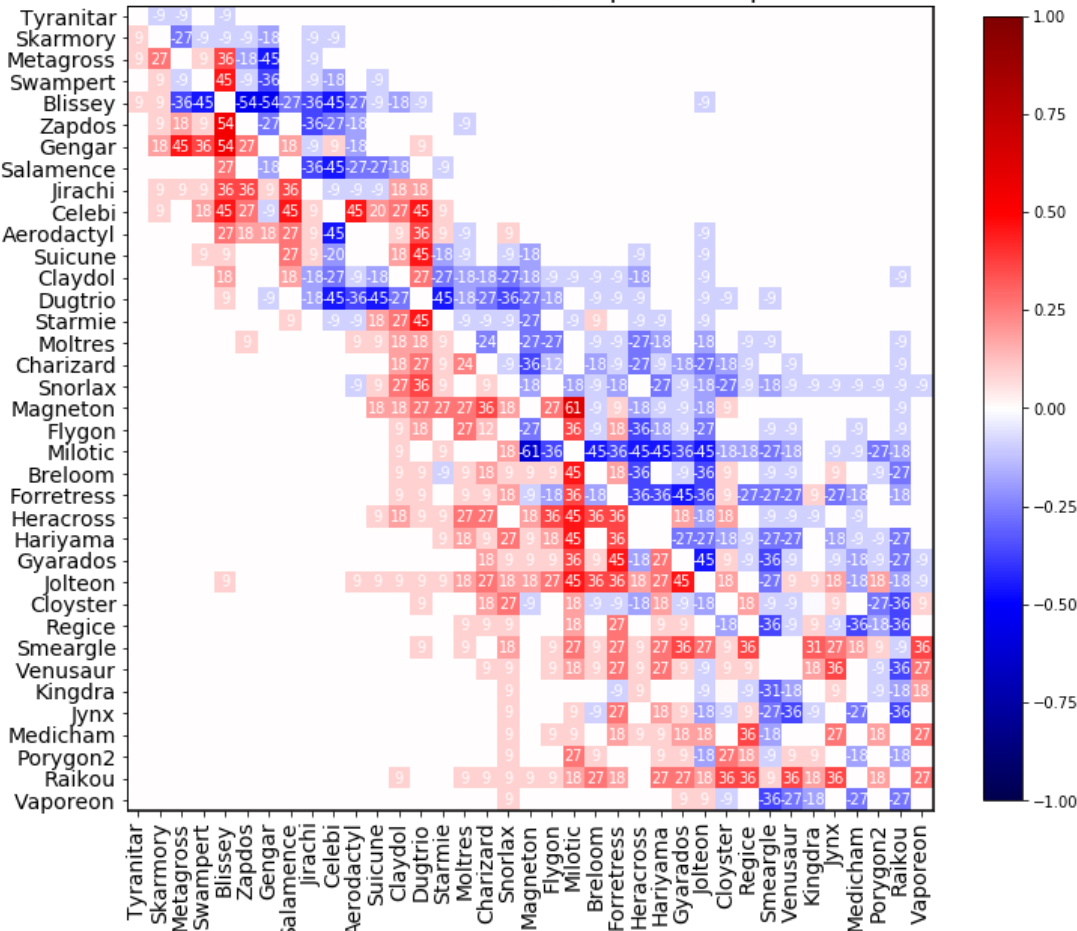




Dissimilarity S/A/B1/B2/B3/C1/C2/D1/D2, Threshold 1.0, 11/22 voters in airports Camp

Red = airports Camp Favors Y; Remainder Favors X

Blue = Remainder Favors Y; airports Camp Favors X



```

In [65]: ### Individual Relative Ranks

# Define range for clustering
refRange = 1
# Order voters by camps
pair = segments[refRange]
order = Dlist[refRange]['leaves']
namesInd = [names[i] for i in order]

## Option 1 plots relative rank wrt mean rank
rankFinalSubtractedOrder = rankMeanSubtracted[:,order]

## Option 2 plots relative rank wrt integer rank
# rankFinalSubtracted = rankval - np.repeat(np.array([np.arange(1,N[0]+1)]).tr
anspose(),rankval.shape[1],axis=1)
# rankFinalSubtractedOrder = rankFinalSubtracted[:,order]

# Eliminate underranked Pokemon
rankFinalSubtractedOrder = rankFinalSubtractedOrder[orderReduced,:]
Nr = np.shape(rankFinalSubtractedOrder)
monListReduced = monList[orderReduced]

fig = plt.figure(figsize=(Nr[1]/3,Nr[0]/3))
ax = plt.gca()
h = ax.imshow(-rankFinalSubtractedOrder,interpolation='none',cmap='seismic',vm
in=-10,vmax=10)

fig.colorbar(h,ax=ax,shrink=0.8)
title = 'Individual Relative Rankings\n Sorted By Camps in '
title += tierNames[pair[0]]
for m in range(pair[0]+1,pair[1]):
    title += '/' + tierNames[m]
title += '\nPokemon with <' + str(int(100*rankFracThreshold)) + '% votes are r
emoved'
ax.set_title(title,fontsize=18)
ax.set_xticks(range(0,Nr[1]))
ax.set_yticks(range(0,Nr[0]))

ax.get_xaxis().set_visible(False)
ax1 = ax.twinx()
ax1.xaxis.tick_top()
ax2 = ax.twinx()
ax2.xaxis.tick_bottom()

ax.get_yaxis().set_visible(False)
ax3 = ax.twinx()
ax3.yaxis.tick_left()
ax4 = ax.twinx()
ax4.yaxis.tick_right()

ax1.set_xticks(np.append(np.arange(0.5,Nr[1]),Nr[1]))
ax2.set_xticks(np.append(np.arange(0.5,Nr[1]),Nr[1]))
ax3.set_yticks(np.append(Nr[0],np.arange(Nr[0]-0.5,0,-1)))
ax4.set_yticks(np.append(Nr[0],np.arange(Nr[0]-0.5,0,-1)))
ax1.set_xticklabels([namesInd[i] for i in range(Nr[1])],fontsize=520/cutoff)
ax2.set_xticklabels([namesInd[i] for i in range(Nr[1])],fontsize=520/cutoff)

```



```

ax3.set_yticklabels(['']+monListReduced[i] for i in range(Nr[0])),fontsize=52
0/cutoff)
ax4.set_yticklabels(['']+monListReduced[i] for i in range(Nr[0])),fontsize=52
0/cutoff)
plt.setp(ax1.get_xticklabels(), rotation=90, ha="left",va="center",rotation_mo
de="anchor");
plt.setp(ax2.get_xticklabels(), rotation=90, ha="right",va="center",rotation_m
ode="anchor");
for ii in range(0,Nr[0]):
    for jj in range(0,Nr[1]):
        if not np.isnan(-rankFinalSubtractedOrder[ii,jj]):
            text = ax.text(jj, ii, int(np.round(-rankFinalSubtractedOrder[ii,j
j])),
                                ha="center", va="center", color="w",fontsize=360/cu
toff)

ax.set_aspect('auto')
for n in np.arange(1,Nr[1]/5):
    ax.plot([n*5-0.5,n*5-0.5],[-0.5,ax.get_ylim()[0]],color='k')

tempBool = np.zeros(N[0])
tempBool[clusterIndicesMod[clusterIndicesMod!=N[0]]] = 1
tempBool = tempBool[orderReduced]
clusterIndicesModReduced = np.nonzero(tempBool)
clusterIndicesModReduced = np.append(clusterIndicesModReduced,Nr[0])

for n in clusterIndicesModReduced:
    ax.plot([-0.5,ax.get_xlim()[1]],[n-0.5,n-0.5],color='c')
plt.delaxes(fig.axes[1])

```


Individual Relative Rankings
Sorted By Camps in S/A/B1/B2/B3
Pokemon with <72% votes are removed

	Triangles	SEA	Fruhdazi	watermess	airports	Altina	Arctic	jrachee	zomog	oijr	Siglut	Gacu	Caloom	ABR	Anonymous	baddummy	Star	BKC	Jester	Kristyl	McMeghan	Johnnyg2	
Tyranitar		-1					-2	-1						-1									Tyranitar
Skarmory	-3	-2	-2	-3	0	0	0	1	-1	0	0	0	0	1	0	0	0	0	0	0	0	0	Skarmory
Metagross		3	2	1	-2	-1	-2		2	1	-1	1	1	-1	-2		1	-1	1		-1	1	Metagross
Swampert	-4	1	0	2	-1	0	-1	1	1	2	2	2	2	3		1	1	-1	-1	-2	0	0	Swampert
Blissey	-3	-3	-5	-1	-8	3	5	-2	0	1	-2	0	1	3	2	3	2	3	2	3	3	2	Blissey
Zapdos	-1	-6	-3	4	2	-1	2	3	2	-3		1	-3		1	-1				1	2	1	Zapdos
Gengar	4	3	-4	0	3	-4	1	-3	-2	-1	3	-2	0	-2	-3	0	-1	2	1	0	1	-1	Gengar
Salamence	-3	2	3	-6	3	4	-4	-2	-1	0	-2	-2	-2	1	1	0	0	-1	-1	1	1	-1	Salamence
Jirachi	6	2	2	1	1	-3	-1	-3	2	-2		2	2		-1			-1	1		-2	-1	Jirachi
Celebi	4	2	7	1	1	1	-4	-1	-2	4	3	1	2	-2	-2		-3	-2	-2	-5	-3	-4	Celebi
Aerodactyl	1	0	3	-2	-1	-5	0	4	0	-1	0	0	0	1	0	0	0	0	0	-2	1	0	Aerodactyl
Suicune	5	0	1	1	-5	0	4	0	3	0	-2	1	0	-1	-1	-1	-1	0	0	3	0	1	Suicune
Claydol	-7	-5	-1	2	0	5	5	7	-1	3	0	-6	1	0	0	0	1	-3	-1	-1	3	5	Claydol
Dugtrio		-6		-2	3	2		-3		-1	-10	-3	-4	2	5	1	3	5	3	2	5		Dugtrio
Starmie	2	1	1	6	0	1	1	-4	1	1	0	2	0	1	0	1	1	-5	-2	4	-5	1	Starmie
Moltres	-6	8	-1		-5	1			-2	2	-4	-2	-1	-1	1	1	-1	3	3	2	-2	1	Moltres
Charizard	-8	2	1	2	-1	1	-3	-1	1	1	-2	-10	3	-2	3	2	-2	4	2	1	-4		Charizard
Snorlax	3	5	3	-7	4	5	-6	-2	3	1	1	4	5	2	4	-6	3	-16	-2	1	3	3	Snorlax
Magnetron	-1	3		-2	3	-1	-2	6	-2	-2	8	-4		2		1	-1	1	1	-3	-3	-2	Magnetron
Flygon	2	-6	2	3	-9	-6	4	5	-4	-1		-2	-1	4	1		3		3	1	2	-1	Flygon
Milotic	7	-8	-8	-10	1	6	-1	5	1	3	-11	-2	-2	1	4	4	4	4	-1	1	5	2	Milotic
Breloom	1	2	-4	2	1	-1	3	1	5	-2		6	-6	1	2	0	0	-1	-3	-1	4	-1	Breloom
Forretress	-7	-8		3	3	5	4	-6	0	0	-8	10	0	-1	-2	2	1	-2	-2	3	8	4	Forretress
Heracross	6	3	2	-1	-2	-2	-7	-1	-2	-1	9	5	1		1	-1	-1		3	-6	-2	-2	Heracross
Hariyama	-2	-12	4	0	-2	-4	8	-2	-2	6	0	-4	8	0	-2	0	0	8	4	4	2	-4	Hariyama
Gyarados	1	4	2	7	-6	-4	-9	1	0	0	2	1	-1	1	0	4	0	-8	-4	1	0	3	Gyarados
Jolteon		-6	5		16	-1		-4		-1	8	6	2	-4	-2	-2	-5	-6	-1	-3		1	Jolteon
Cloyster	-5	-4	0	5		5	-9	-2	9	-8	11	1	-4	0	-12		-1	2	4	2	1	-1	Cloyster
Regice	9	-4	0	1	-8	-5	6	6	-3	3	3	-8	-8	7	5	-3	5	1	-1	7	-4	1	Regice
Smeargle	5	4	-4	4	6	-1	8	-2	-4	2	11	2	-7	-1	1	-8	-3		0	-2	-10	4	Smeargle
Venusaur		7	-2	-9	-2	-2	-15	0	1	5	7	6	-4	-2	-2	5	1	10	4	0	1	-5	Venusaur
Kingdra		-2	-5	-5	-2	-3	-4	-2	-1	-4	-1	2			9		-3	6	6	-3	3	1	Kingdra
Jynx	4	-1	-6	-9	9	11	-29	6	4	2	5		-1	3		-1	5		-2	-2	1	-2	Jynx
Medicham		10	-2	-6	-4	-11	5	10	-11	0		-2	-4	1	-2	-4	-1	4	4	5	-5	6	Medicham
Porygon2	2	-9	9	4	-4		1	1	-6	2	2	-4	3	5	-10	4	4		-5	-4	3		Porygon2
Raikou	15	10	8	-10	4	-7	2	-3	3	-1		-5	6		-11	2	-8		4	-3	-3	-2	Raikou
Vaporeon		-2	0	-6	4	5	-13	-1	7	-10		0	-1		5	6	4	9	6	-2	-2	-1	Vaporeon
Machamp		9	5	4	2	-2	-10		0	4		4	3		-12	-5	-3		1	5		-2	Machamp
Registeel		-6	7	5		-4	2			-3			-7				-12	8		7	4	-5	Registeel
Ludicolo	8	-1	-1	-1	0	8	-19	3	-2	-4		5	-3			3	-10		5	-1	11	6	Ludicolo
Weezing		2		3		5	2			1	13				-11	2	-5			-1		-6	Weezing
Misdreavus		-12		6		14	-1		-1	4					4	2	3			-8	1	0	Misdreavus
Blaziken		15					-4			4			-1			-2	5					-8	Blaziken
Steelix				-3	1		-3	3	9	-6	12	6	-8		-7		-6	10	2	-1		-7	Steelix
Regirock		-3					8						-4		12	-4	-5				-1	-3	Regirock

In []:

