# **ChickenRanging NB4**

Network analysis and correlation of network and time warping similarity.

Version: 7.0.

Author: Bernhard Voelkl Created in: *Mathematica* 10.2 Last modified: 01/03/2018

### **Data Import**

### **Functions**

**findFriend**[i] is a function that requires that a matrix pen is globally defined, where pen gives the temporally sorted transition data for one pen for one entire day. Function input i is an integer

counting variable (indicating the ith line of the matrix pen). For each transition of a hen the function finds all other transitions by other hens within 5 seconds at the same antenna (in the same direction). The hen closest in time is considered the "friend" of the focus animal and a an edge {ID focus, ID friend} is created. In case more than one hen crossed the antenna within the same second an edge is created for each pairing with the focus bird {{ID focus, ID friend1}, {ID focus, ID friend2} ..}. In case no other henn crosses the antenna within 5 seconds, an empty list is returned, otherwise the function returns a list of edges.

#### Functions for the standard matrix permutation test

```
In[*]:= rearrange[mat_] := Module[{p},
       p = RandomPermutation[Length[mat]];
      Permute[Transpose[Apply[Permute, List[Transpose[mat], p]]], p]
```

rearrange [mat] takes as input a square matrix and randomly re-arranges lines and rows (both according to the same random permutation -this ensures that the main diagonal remains empty). The output is the re-arranged matrix.

```
ln[*]:= ppmcc[xmat_, ymat_] := Module[{{}},
       k = Length[xmat];
       n = (Length[xmat] (Length[xmat] - 1));
       (Total[xmat * ymat, 2] - n * Total[xmat / n, 2] * Total[ymat / n, 2]) /
          ((Total[(xmat - Total[xmat / n, 2])^2) * (1 - IdentityMatrix[k]), 2])
             Total[(ymat - Total[ymat/n, 2])^2) * (1 - IdentityMatrix[k]), 2]))^0.5 // N]
```

ppmcc[xmat, ymat] is a function that needs two square matrices as input and gives as output the matrix correlation: Pearson's product-moment correlation coefficient, according to Smouse et al 1986 Syst.Zool.

```
In[@]:= newpermutationR[mat1_, mat2_, permut_] := Module[{}},
       statistic = ppmcc[mat1, mat2];
       expectedR = Append[Table[ppmcc[mat1, rearrange[mat2]], {permut - 1}], statistic];
       std = StandardDeviation[expectedR];
       sigma = Abs[Mean[expectedR] - statistic] / std;
       highercounts = Count[expectedR, _?(#1 ≥ statistic &)];
       arr = Graphics \left[ Arrow \left[ \left\{ \left\{ statistic, -\frac{permut}{100} \right\}, \left\{ statistic, 10 \right\} \right\} \right] \right];
       hist = Histogram[expectedR, PlotLabel -> Grid[{
              {StringJoin["Expected r: ", ToString[Mean[expectedR]], " CI95[", ToString[Mean[
                    expectedR] - 1.96 std], ", ", ToString[Mean[expectedR] + 1.96 std], "]"]},
              {StringJoin["Observed r: ", ToString[N[statistic]], ", N(r+) = ",
                ToString[highercounts], " \sigma = ", ToString[sigma]]}}]];
       Show hist, arr, PlotRange → All, ImageSize → 1200, Axes → False,
         Frame → True, FrameStyle → Directive[Thick, Black, 32],
         FrameLabel → {"r²"}, LabelStyle → Directive[24, Black]]]
```

newpermutationR[mat1, mat2, permut] is a function that needs three arguments: mat1 and mat2: the matrices to be correlated and *permut*: the number of permutations. It calculates the matrix

correlation for mat1 × mat2 and for mat1 with the permut times re-arranged matrix mat2. It gives as output "r" of mat1 × mat2, N+, the number of times that the the permuted matrices had a correlation as large or larger than the observed one and the effect size sigma as the difference between observed and expected r in units of standard deviations.

```
In[*]:= newpermutationDataOnly[mat1_, mat2_, permut_] := Module[{}},
       statistic = ppmcc[mat1, mat2];
       expectedR = Append[Table[ppmcc[mat1, rearrange[mat2]], {permut - 1}], statistic];
       std = StandardDeviation[expectedR];
       sigma = Abs[Mean[expectedR] - statistic] / std;
       highercounts = Count[expectedR, _?(#1 ≥ statistic &)];
       {N[statistic], expectedR, sigma}
```

newpermutationDataOnly[mat1,mat2,permut] is a function that needs three arguments: mat1 and mat2: the matrices to be correlated and permut: the number of permutations. It calculates the matrix correlation for mat1 × mat2 and for mat1 with the permut times re-arranged matrix mat2. It gives as output "r" of mat1 × mat2, N+, the number of times that the the permuted matrices had a correlation as large or larger than the observed one and the effect size sigma as the difference between observed and expected r in units of standard deviations.

## 1. Edge Lists for Overall Networks (summed over 72 days)

For the overall pen network (all days pooled) a weighted edge list is created for each pen based on dyadic associations at pens. A link is added for two hens passing the same antenna (in the same direction) within 5 seconds. (See comment for function findFriend[i]).

### Edgelist Pen 11

```
edgelistpen11 = {};
For[d = 1, d ≤ Length[datelist], d++,
 pen = Cases[data, {_, _, _, 11, datelist[d], __}];
 pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
 If[Length[pen] > 1,
    AppendTo[edgelistpen11, ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}]]
 ]
]
edgelistpen11 = DeleteCases[Flatten[edgelistpen11, 2], {}];
edgelistpen11 = Sort /@ edgelistpen11;
edgelistpen11 = Tally[edgelistpen11];
edgelistpen11 = DeleteCases[edgelistpen11, {{v1_, v2_}, _} /; v1 == v2];
Save["edgelistpen11", edgelistpen11];
```

### Edgelist Pen 12

```
edgelistpen12 = {};
  For [d = 1, d \le Length[datelist], d++,
   pen = Cases[data, {_, _, _, 12, datelist[d], __}];
   pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
   If[Length[pen] > 1,
      AppendTo[edgelistpen12, ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}]]
   1
  ]
  edgelistpen12 = DeleteCases[Flatten[edgelistpen12, 2], {}];
  edgelistpen12 = Sort /@ edgelistpen12;
  edgelistpen12 = Tally[edgelistpen12];
  edgelistpen12 = DeleteCases[edgelistpen12, {{v1_, v2_}, _} /; v1 == v2];
  Save["edgelistpen12", edgelistpen12];
Edgelist Pen 13
  edgelistpen13 = {};
  For[d = 1, d ≤ Length[datelist], d++,
   pen = Cases[data, {_, _, _, 13, datelist[d], __}];
   pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
   If[Length[pen] > 1,
      AppendTo[edgelistpen13, ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}]]
   ]
  1
  edgelistpen13 = DeleteCases[Flatten[edgelistpen13, 2], {}];
  edgelistpen13 = Sort /@ edgelistpen13;
  edgelistpen13 = Tally[edgelistpen13];
  edgelistpen13 = DeleteCases[edgelistpen13, {{v1_, v2_}, _} /; v1 == v2];
  Save["edgelistpen13", edgelistpen13];
Edgelist Pen 14
  edgelistpen14 = {};
  For[d = 1, d ≤ Length[datelist], d++,
   pen = Cases[data, {_, _, _, 14, datelist[d], __}];
   pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
   If[Length[pen] > 1,
      AppendTo[edgelistpen14, ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}]]
   1
  ]
  edgelistpen14 = DeleteCases[Flatten[edgelistpen14, 2], {}];
  edgelistpen14 = Sort /@ edgelistpen14;
  edgelistpen14 = Tally[edgelistpen14];
```

```
edgelistpen14 = DeleteCases[edgelistpen14, {{v1_, v2_}, _} /; v1 == v2];
Save["edgelistpen14", edgelistpen14];
```

## 2. Association Indices for Overall Networks (summed over 72 days)

#### Pen 11

```
allhens11 = Cases[allhens, {_, "11", __}];
edgelistpen11 = Get["edgelistpen11"];
data11 = Cases[data, {_, _, _, 11, __}];
indivobservations =
  Table[Length[Cases[data11, {allhens11[i, 3], __}]], {i, Length[allhens11]}];
n = Length[allhens11];
amat = Table[0, {n}, {n}];
For [i = 1, i \le n, i++,
For [j = 1, j \le n, j++,
  pair = Cases[edgelistpen11, {{allhens11[i, 3], allhens11[j, 3]]}, _}];
  If[Length[pair] > 0,
   amat[i, j] = pair[1, 2]
]
]
```

The association index (AI) is calculated as: X(AB)/(X(A) + X(B) - X(AB))

```
amat = amat + Transpose[amat];
aimat = N[Table[If[indivobservations[i]] + indivobservations[j]] > 0,
     amat[i, j] / (indivobservations[i] + indivobservations[j] - amat[i, j]), "NA"],
    {i, Length[allhens11]}, {j, Length[allhens11]}]];
Save["aimat11", aimat];
Export["aimat11.csv", aimat];
```

```
distmat = Get["distmat12"];
dmhead = distmat[1];
allhens12 = Cases[allhens, {_, "12", __}];
edgelistpen12 = Get["edgelistpen12"];
data12 = Cases[data, {_, _, _, 12, __}];
```

```
indivobservations =
    Table[Length[Cases[data12, {allhens12[i, 3], __}]], {i, Length[allhens12]}];
  n = Length[allhens12];
  amat = Table[0, {n}, {n}];
  For [i = 1, i \le n, i++,
   For [j = 1, j \le n, j++,
    pair = Cases[edgelistpen12, {{allhens12[i, 3], allhens12[j, 3]}}, _}];
    If[Length[pair] > 0,
      amat[i, j] = pair[1, 2]
    ]
   ]
  ]
  (*Association index: X(AB)/(X(A) + X(B) - X(AB))*)
  amat = amat + Transpose[amat];
  aimat = N[Table[If[indivobservations[i]] + indivobservations[j]] > 0,
        amat[i, j] / (indivobservations[i] + indivobservations[j] - amat[i, j]), "NA"],
       {i, Length[allhens12]}, {j, Length[allhens12]}]];
  Save["aimat12", aimat];
  Export["aimat12.csv", aimat];
Pen 13
  distmat = Get["distmat13"];
  dmhead = distmat[[1]];
  allhens13 = Cases[allhens, {_, "13", __}];
  edgelistpen13 = Get["edgelistpen13"];
  data13 = Cases[data, {_, _, _, 13, __}];
  indivobservations =
    Table[Length[Cases[data13, {allhens13[i, 3], __}]], {i, Length[allhens13]}];
  n = Length[allhens13];
  amat = Table[0, {n}, {n}];
  For [i = 1, i \le n, i++,
   For [j = 1, j \le n, j++,
    pair = Cases[edgelistpen13, {{allhens13[i, 3], allhens13[j, 3]}}, _}];
    If[Length[pair] > 0,
      amat[i, j] = pair[1, 2]
    ]
   ]
  ]
  (*Association index: X(AB)/(X(A) + X(B) - X(AB))*)
  amat = amat + Transpose[amat];
```

```
aimat = N[Table[If[indivobservations[i]] + indivobservations[j]] > 0,
        amat[i, j] / (indivobservations[i] + indivobservations[j] - amat[i, j]), "NA"],
       {i, Length[allhens13]}, {j, Length[allhens13]}]];
  Save["aimat13", aimat];
  Export["aimat13.csv", aimat];
Pen 14
  distmat = Get["distmat14"];
  dmhead = distmat[[1]];
  allhens14 = Cases[allhens, {_, "14", __}];
  edgelistpen14 = Get["edgelistpen14"];
  data14 = Cases[data, {_, _, _, 14, __}];
  indivobservations =
    Table[Length[Cases[data14, {allhens14[i, 3], __}]], {i, Length[allhens14]}];
  n = Length[allhens14];
  amat = Table[0, {n}, {n}];
  For [i = 1, i \le n, i++,
   For [j = 1, j \le n, j++,
    pair = Cases[edgelistpen14, {{allhens14[i, 3], allhens14[j, 3]}, _}];
    If[Length[pair] > 0,
      amat[i, j] = pair[1, 2]
    ]
   ]
  (*Association index: X(AB)/(X(A) + X(B) - X(AB))*)
  amat = amat + Transpose[amat];
  aimat = N[Table[If[indivobservations[i]] + indivobservations[j]] > 0,
        amat[i, j] / (indivobservations[i] + indivobservations[j] - amat[i, j]), "NA"],
       {i, Length[allhens14]}, {j, Length[allhens14]}]];
  Save["aimat14", aimat];
  Export["aimat14.csv", aimat];
```

## 3. Network Descriptors for Overall Networks (summed over 72 days)

```
edgelistpen11 = Get["edgelistpen11"];
aimat = Get["aimat11"];
```

```
pen11aigraph = WeightedAdjacencyGraph[aimat //. {"NA" → 0, 0. → 0},
   EdgeWeight, DirectedEdges → True, EdgeStyle → Arrowheads[0]];
pen11aimultigraph = AdjacencyGraph [Round [ (aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0}) 10000],
   DirectedEdges → True, EdgeStyle → Arrowheads[0]];
N[GraphDensity[pen11aimultigraph]](*Graphdensity*)
0.942714
max = Total[edgelistpen11[All, 2]] (*total number of links*)
134 060
n = Length[aimat];
k = n(n-1);
Show[ListPlot[Transpose[{Range[k] / k,
    PadLeft[Accumulate[Sort[Table[Random[PoissonDistribution[N[max/k]]], {k}]]],
       k, 0] / max}], PlotStyle \rightarrow Gray, Frame \rightarrow True, AspectRatio \rightarrow 1,
  FrameStyle → Directive[Thick, Black, 20], LabelStyle → Directive[18, Black]],
 ListPlot[Transpose[{Range[k] / k,
     PadLeft[Accumulate[Sort[edgelistpen11[All, 2]]], k, 0] / max}], PlotStyle → Red]]
1.0
0.8
0.6
0.4
0.2
0.0
             0.2
                                0.6
                                         8.0
                                                   1.0
   0.0
                      0.4
```

Expected (grey) and observed (red) CDF for edge weights

```
GraphAssortativity[pen11aigraph, FindGraphCommunities[pen11aigraph]]
0.0931133
btw11 = BetweennessCentrality[pen11aimultigraph];
prc11 = PageRankCentrality[pen11aimultigraph, 0.1];
lsc11 = StatusCentrality[pen11aimultigraph];
gc11 = FindGraphCommunities[pen11aigraph];
```

```
vertexlist11 = VertexList[pen11aigraph];
cms11 = Table[Position[gc11, vertexlist11[i]][1, 1], {i, Length[vertexlist11]}];
gc11 // Length (*Number of graph communities*)
5
```

```
edgelistpen12 = Get["edgelistpen12"];
aimat = Get["aimat12"];
pen12aigraph = WeightedAdjacencyGraph[aimat //. \{"NA" \rightarrow 0, 0. \rightarrow 0\},
    EdgeWeight, DirectedEdges \rightarrow True, EdgeStyle \rightarrow Arrowheads[0]];
pen12aimultigraph = AdjacencyGraph [Round [ (aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0}) 10000],
    DirectedEdges → True, EdgeStyle → Arrowheads[0]];
N[GraphDensity[pen12aimultigraph]](*Graphdensity*)
0.931277
max = Total[edgelistpen12[All, 2]] (*total number of links*)
95 219
n = Length[aimat];
k = n(n-1);
Show[ListPlot[Transpose[{Range[k] / k,
     PadLeft [Accumulate | Sort | Table | Random | PoissonDistribution | N | max / k | ] ], {k}]]],
        [k, 0] / max}], PlotStyle \rightarrow Gray, Frame \rightarrow True, AspectRatio \rightarrow 1,
  FrameStyle → Directive[Thick, Black, 20], LabelStyle → Directive[18, Black]],
 ListPlot[Transpose[{Range[k] / k,
     PadLeft[Accumulate[Sort[edgelistpen12[All, 2]]], k, 0] / \max], PlotStyle \rightarrow Red]]
1.0
8.0
0.6
0.4
0.2
0.01
              0.2
                        0.4
                                  0.6
                                            8.0
                                                      1.0
   0.0
```

```
(*Expected (grey) and oberved (red) CDF for edeweights *)
  GraphAssortativity[pen12aigraph, FindGraphCommunities[pen12aigraph]]
  0.0844989
  btw12 = BetweennessCentrality[pen12aimultigraph];
  prc12 = PageRankCentrality[pen12aimultigraph, 0.1];
  lsc12 = StatusCentrality[pen12aimultigraph];
  gc12 = FindGraphCommunities[pen12aigraph];
  vertexlist12 = VertexList[pen12aigraph];
  cms12 = Table[Position[gc12, vertexlist12[i]]][1, 1], {i, Length[vertexlist12]}];
  gc12 // Length
  4
Pen 13
  edgelistpen13 = Get["edgelistpen13"];
  aimat = Get["aimat13"];
  pen13aigraph = WeightedAdjacencyGraph[aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0},
      EdgeWeight, DirectedEdges → True, EdgeStyle → Arrowheads[0]];
  pen13aimultigraph = AdjacencyGraph [Round [ (aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0}) 10000],
      DirectedEdges → True, EdgeStyle → Arrowheads[0]];
  N[GraphDensity[pen13aimultigraph]](*Graphdensity*)
  0.924708
  max = Total[edgelistpen13[All, 2]](*total number of links*)
  114 310
  n = Length[aimat];
  k = n (n - 1);
```

```
Show[ListPlot[Transpose[{Range[k] / k,
        \label{eq:padleft} PadLeft \big[ Accumulate \big[ Sort \big[ Table \big[ Random \big[ PoissonDistribution \big[ N \big[ max \big/ k \big] \big] \big] \big], \ \{k\} \big] \big] \big],
          [k, 0]/max}], PlotStyle \rightarrow Gray, Frame \rightarrow True, AspectRatio \rightarrow 1,
     FrameStyle → Directive[Thick, Black, 20], LabelStyle → Directive[18, Black]],
    ListPlot[Transpose[{Range[k] / k,
        PadLeft[Accumulate[Sort[edgelistpen13[All, 2]]], k, 0] / max}], PlotStyle → Red]]
   1.0
  0.8
  0.6
  0.4
  0.2
  0.0
                0.2
                           0.4
                                     0.6
                                               8.0
      0.0
                                                         1.0
   (*Expected (grey) and oberved (red) CDF for edeweights *)
  GraphAssortativity[pen13aigraph, FindGraphCommunities[pen13aigraph]]
  0.109329
  btw13 = BetweennessCentrality[pen13aimultigraph];
  prc13 = PageRankCentrality[pen13aimultigraph, 0.1];
  lsc13 = StatusCentrality[pen13aimultigraph];
  gc13 = FindGraphCommunities[pen13aigraph];
  vertexlist13 = VertexList[pen13aigraph];
  cms13 = Table[Position[gc13, vertexlist13[i]]][1, 1], {i, Length[vertexlist13]}];
  gc13 // Length
  5
Pen 14
  edgelistpen14 = Get["edgelistpen14"];
  aimat = Get["aimat14"];
  pen14aigraph = WeightedAdjacencyGraph[aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0},
      EdgeWeight, DirectedEdges → True, EdgeStyle → Arrowheads[0]];
```

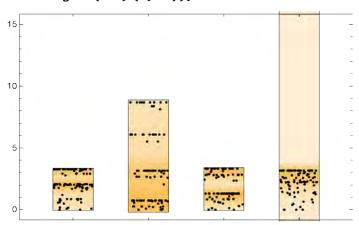
```
pen14aimultigraph = AdjacencyGraph [Round [ (aimat //. {"NA" \rightarrow 0, 0. \rightarrow 0} ) 10000],
   DirectedEdges → True, EdgeStyle → Arrowheads[0]];
N[GraphDensity[pen14aimultigraph]](*Graphdensity*)
0.901792
max = Total[edgelistpen14[All, 2]] (*total number of links*)
101 334
n = Length[aimat];
k = n(n-1);
Show[ListPlot[Transpose[{Range[k]/k,
    PadLeft[Accumulate[Sort[Table[Random[PoissonDistribution[N[max/k]]], \{k\}]]],\\
       k, 0]/max}], PlotStyle \rightarrow Gray, Frame \rightarrow True, AspectRatio \rightarrow 1,
  FrameStyle → Directive[Thick, Black, 20], LabelStyle → Directive[18, Black]],
 ListPlot[Transpose[\{Range[k]/k,
     PadLeft[Accumulate[Sort[edgelistpen14[All, 2]]], k, 0] / max\}], PlotStyle \rightarrow Red]]
1.0
8.0
0.6
0.4
0.2
0.0
             0.2
                                         8.0
                                                  1.0
   0.0
                      0.4
                               0.6
(*Expected (grey) and oberved (red) CDF for edeweights *)
GraphAssortativity[pen14aigraph, FindGraphCommunities[pen14aigraph]]
0.120954
btw14 = BetweennessCentrality[pen14aimultigraph];
prc14 = PageRankCentrality[pen14aimultigraph, 0.1];
lsc14 = StatusCentrality[pen14aimultigraph];
gc14 = FindGraphCommunities[pen14aigraph];
vertexlist14 = VertexList[pen14aigraph];
cms14 = Table[Position[gc14, vertexlist14[i]]][1, 1], {i, Length[vertexlist14]}];
```

gc14 // Length

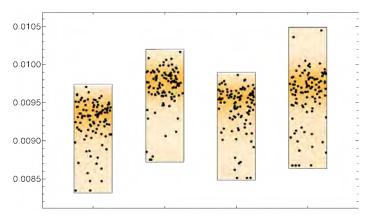
6

#### **Individual Measures**

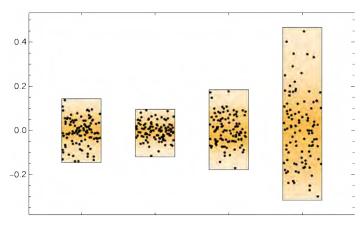
DistributionChart[{btw11, btw12, btw13, btw14}, PlotRange  $\rightarrow$  {All, {0, 15}}, ChartElementFunction  $\rightarrow$  "PointDensity"]



DistributionChart[{prc11, prc12, prc13, prc14}, ChartElementFunction → "PointDensity"]



DistributionChart[{lsc11, lsc12, lsc13, lsc14}, ChartElementFunction → "PointDensity"]



#### Network measure Matrix

```
networkmeasures =
  {Join[allhens11[All, 1], allhens12[All, 1], allhens13[All, 1], allhens14[All, 1]],
   Join[Table[11, {Length[vertexlist11]}], Table[12, {Length[vertexlist12]}],
    Table[13, {Length[vertexlist13]}], Table[14, {Length[vertexlist14]}]],
   Join [btw11, btw12, btw13, btw14], Join [prc11, prc12, prc13, prc14],
   Join [lsc11, lsc12, lsc13, lsc14] //. Indeterminate \rightarrow 0,
   Join[0.1 cms11 + 11, 0.1 cms12 + 12, 0.1 cms13 + 13, 0.1 cms14 + 14];
networkmeasures = Prepend[Transpose[networkmeasures],
   {"Bird", "Pen", "Betweenness", "Centrality", "Status", "Community"}];
Export["networkmeasuresnew.csv", networkmeasures]
networkmeasuresnew.csv
```

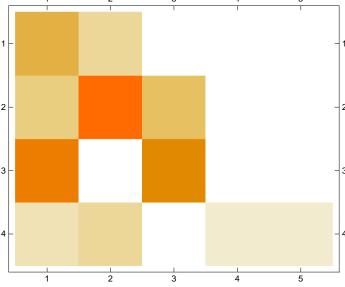
## Correlating NW measures with DTW Clusters

#### Pen 11

```
cl11 = Import["cl11.csv", "table", FieldSeparators → ";"];
```

Classifications based on dynamic time warping (computed in R) are imported as .csv files.

```
dtwclusters11 = Table[Cases[cl11, {_, ci}] [All, 1], {ci, 1, 4}]
ngmodules11 = Table[Cases[networkmeasures, {__, 11 + 0.1 ci}] [All, 1], {ci, 1, 5}];
MatrixPlot[Table[
  Table[Length[Intersection[dtwclusters11[[t]], ngmodules11[[u]]]], {u, 1, 5}], {t, 1, 4}]]
```

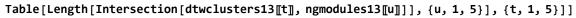


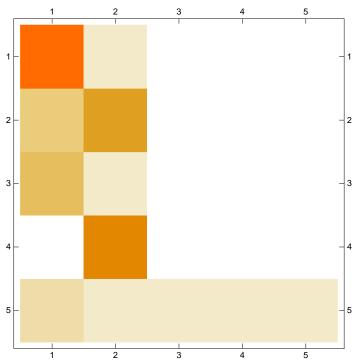
#### Pen 12

```
cl12 = Import["cl12.csv", "table", FieldSeparators → ";"];
dtwclusters12 = Table[Cases[cl12, {_, ci}] [All, 1], {ci, 1, 4}]
ngmodules12 = Table[Cases[networkmeasures, {__, 12 + 0.1 ci}] [All, 1], {ci, 1, 4}];
MatrixPlot[Table[
  Table [Length [Intersection [dtwclusters 12 [t]], ngmodules 12 [u]]]], \{u, 1, 4\}], \{t, 1, 4\}]]
```

```
cl13 = Import["cl13.csv", "table", FieldSeparators → ";"];
dtwclusters13 = Table[Cases[cl13, {_, ci}] [All, 1], {ci, 1, 5}]
ngmodules 13 = Table [Cases [network measures, {\_\_, 13 + 0.1 ci}] [All, 1], {ci, 1, 5}]
```

#### MatrixPlot[Table[





#### Pen 14

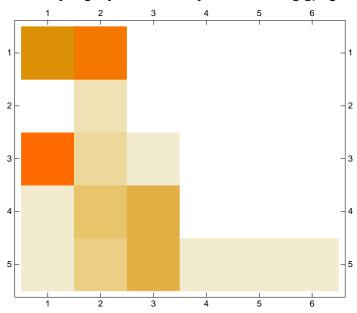
```
cl14 = Import["cl14.csv", "table", FieldSeparators → ";"];
```

dtwclusters14 = Table[Cases[cl14, {\_, ci}] [All, 1], {ci, 1, 5}]

 $ngmodules 14 = Table [Cases [network measures, \{\_\_, 14 + 0.1 ci\}] [All, 1], \{ci, 1, 6\}];$ 

#### MatrixPlot[Table[

 $Table [Length [Intersection [dtwclusters 14[t]], ngmodules 14[u]]], \{u, 1, 6\}], \{t, 1, 5\}]]\\$ 



#### Combined probability

G statistics for the contingency tables were calculated in the R script "GTestClustersModules". The combined probability for all four pens is summarized below

```
chisquared =
 -2 \left( \log [7.1 \times 10^{-14}] + \log [6.3 \times 10^{-10}] + \log [1 \times 10^{-13}] + \log [2.1 \times 10^{-7}] \right)
N[Probability[x >= chisquared, x \approx ChiSquareDistribution[8]]]
1.46365 \times 10^{-37}
```

## **Daily Networks**

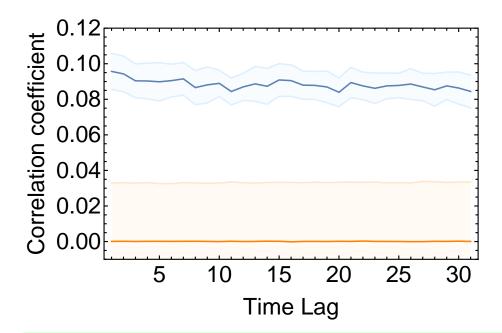
```
dayofyear = {158, 192, 197, 198, 199, 200, 201, 202, 203, 204, 205, 206,
      207, 208, 209, 210, 211, 213, 220, 222, 223, 224, 225, 226, 227, 228, 229,
      230, 234, 235, 236, 237, 238, 239, 240, 242, 243, 244, 245, 246, 247, 249,
      250, 251, 252, 253, 254, 255, 256, 257, 264, 265, 266, 267, 268, 269, 270,
      271, 272, 273, 274, 276, 277, 278, 279, 280, 281, 283, 284, 285, 286, 289};
  shift = 0;
Pen 11
  edgelistpen11 = {};
  For[d = 1, d ≤ Length[datelist], d++,
   pen = Cases[data, {_, _, _, 11, datelist[d], __}];
   pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
   If[Length[pen] > 1,
        dailyedgelist =
      DeleteCases[Tally[Sort /@DeleteCases[Flatten[ParallelTable[findFriend[i],
             {i, 1, Length[pen] - 1}], 1], {}]], {{v1_, v2_}, _} /; v1 == v2];
      AppendTo[edgelistpen11, dailyedgelist],
    Print[StringJoin["No data for this day: ", ToString[d]]]
   ]
  ]
  No data for this day: 68
  Save["dailyedgelistpen11", edgelistpen11];
  Length[edgelistpen11]
  71
```

edgelistpen11 is a nested list with each of the 71 sub-lists being the edgelist for a single day.

```
edgelistpen11 = Get["dailyedgelistpen11"];
wms = Table[WeightedAdjacencyMatrix[Graph[Cases[allhens, {_, "11", __}][All, 3],
     Apply[UndirectedEdge, edgelistpen11[i, All, 1], 1],
     EdgeWeight → edgelistpen11[i, All, 2]]], {i, Length[edgelistpen11]}];
```

#### This creates for pen 11 for each day a weighted adjacency matrix

```
days = Length[wms];
newacorrSN11 = Table[
   Table [newpermutationDataOnly [wms [[q]], wms [[i]], 1000], \{i, q+1, days\}], \{q, 1, 40\}]; \\
Save["newacorrSN11", newacorrSN11]
obsl = Table [Mean [Table [newacorrSN11[i, lag, 1]], {i, 40}]], {lag, 31}];
ciobs =
  1.96 * (Table[StandardDeviation[Table[newacorrSN11[i, lag, 1]], {i, 40}]], {lag, 31}] /
      Sqrt [40]);
exm = Table[Mean[Table[Mean[newacorrSN11[i, lag, 2]]], {i, 40}]], {lag, 31}];
exci = Table [Mean [Table [Sort [newacorrSN11[i, lag, 2]] [975]], {i, 40}]], {lag, 31}];
SNconsistency11 = Show[ListLinePlot[exci, PlotRange → {All, {-0.01, 0.12}},
   PlotStyle → LightOrange, Filling → -0.01, Frame → True,
   FrameStyle → Thickness[0.003], FrameTicksStyle → Directive[Black, 24],
   FrameLabel → {{"Correlation coefficient", ""}, {"Time Lag", ""}},
   LabelStyle -> Directive[Black, 24]], ListLinePlot[exm, PlotStyle → Orange],
  ListLinePlot[{obsl + ciobs, obsl - ciobs}, PlotStyle \rightarrow LightBlue, Filling \rightarrow {1 \rightarrow {2}}],
  ListLinePlot[obsl], ImageSize → 800]
```

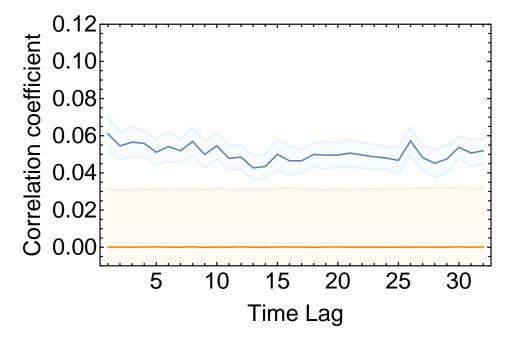


This plot shows the observed (blue) and expected (orange) correlation coefficients of days t = 1 to 40 with days with timelags t+1 to t+31.

```
Export["SNconsistency11.tif", SNconsistency11]
SNconsistency11.tif
```

```
edgelistpen12 = {};
For[d = 1, d ≤ Length[datelist], d++,
 pen = Cases[data, {_, _, _, 12, datelist[d], __}];
 pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
 If[Length[pen] > 1,
     dailyedgelist =
   DeleteCases[Tally[Sort /@ DeleteCases[Flatten[ParallelTable[findFriend[i]],
          {i, 1, Length[pen] - 1}], 1], {}]], {{v1_, v2_}, _} /; v1 == v2];
    AppendTo[edgelistpen12, dailyedgelist],
  Print[StringJoin["No data for this day: ", ToString[d]]]
 ]
]
Save["dailyedgelistpen12", edgelistpen12];
edgelistpen12 = Get["dailyedgelistpen12"];
Length[edgelistpen12]
72
wms = Table[WeightedAdjacencyMatrix[Graph[Cases[allhens, {_, "12", __}] [All, 3],
     Apply[UndirectedEdge, edgelistpen12[i, All, 1], 1],
     EdgeWeight → edgelistpen12[i, All, 2]]], {i, Length[edgelistpen12]}];
days = Length[wms];
newacorrSN12 = Table[ParallelTable[
    newpermutationDataOnly[wms[q], wms[i], 1000], {i, q + 1, days}], {q, 1, 40}];
Save["newacorrSN12", newacorrSN12]
obsl = Table[Mean[Table[newacorrSN12[i, lag, 1]], {i, 40}]], {lag, 32}]
ciobs =
 1.96 * (Table[StandardDeviation[Table[newacorrSN12[i, lag, 1], {i, 40}]], {lag, 32}]/
    Sqrt [40])
exm = Table[Mean[Table[Mean[newacorrSN12[i, lag, 2]]], {i, 40}]], {lag, 32}]
exci = Table [Mean [Table [Sort [newacorrSN12 [i, lag, 2]] [975]], {i, 40}]], {lag, 32}]
```

```
SNconsistency12 = Show[ListLinePlot[exci, PlotRange \rightarrow {All, {-0.01, 0.12}},
   PlotStyle → LightOrange, Filling → -0.01, Frame → True,
   FrameStyle → Thickness[0.003], FrameTicksStyle → Directive[Black, 24],
   FrameLabel → {{"Correlation coefficient", ""}, {"Time Lag", ""}},
   LabelStyle -> Directive[Black, 24]], ListLinePlot[exm, PlotStyle → Orange],
  ListLinePlot[\{obsl + ciobs, obsl - ciobs\}, PlotStyle \rightarrow LightBlue, Filling \rightarrow \{1 \rightarrow \{2\}\}],
  ListLinePlot[obsl], ImageSize → 800]
```

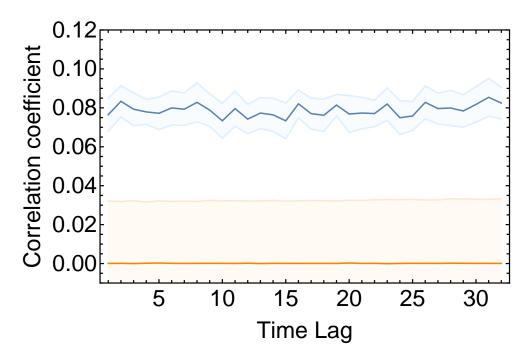


Export["SNconsistency12.tif", SNconsistency12]

SNconsistency12.tif

```
edgelistpen13 = {};
For[d = 1, d ≤ Length[datelist], d++,
 pen = Cases[data, {_, _, _, 13, datelist[d], __}];
 pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
 If[Length[pen] > 1,
     dailyedgelist =
   DeleteCases [Tally [Sort /@ DeleteCases [Flatten [ParallelTable [findFriend[i],
          {i, 1, Length[pen] - 1}], 1], {}]], {{v1_, v2_}, _} /; v1 == v2];
    AppendTo[edgelistpen13, dailyedgelist],
  Print[StringJoin["No data for this day: ", ToString[d]]]
 ]
]
Save["dailyedgelistpen13", edgelistpen13];
edgelistpen13 = Get["dailyedgelistpen13"];
Length[edgelistpen13]
72
```

```
wms = Table[WeightedAdjacencyMatrix[Graph[Cases[allhens, {_, "13", __}] [All, 3],
      Apply [UndirectedEdge, edgelistpen13[i, All, 1], 1],
      EdgeWeight → edgelistpen13[i, All, 2]]], {i, Length[edgelistpen13]}];
days = Length[wms];
newacorrSN13 = Table[ParallelTable[
    newpermutationDataOnly[wms[q]], wms[i]], 1000], \{i, q+1, days\}], \{q, 1, 40\}];
Save["newacorrSN13", newacorrSN13]
obsl = Table[Mean[Table[newacorrSN13[i, lag, 1]], {i, 40}]], {lag, 32}]
 1.96 * (Table[StandardDeviation[Table[newacorrSN13[i, lag, 1], {i, 40}]], {lag, 32}] /
    Sqrt [40])
exm = Table[Mean[Table[Mean[newacorrSN13[i, lag, 2]]], {i, 40}]], {lag, 32}]
exci = Table [Mean [Table [Sort [newacorrSN13 [i, lag, 2]] [975]], {i, 40}]], {lag, 32}]
SNconsistency13 = Show[ListLinePlot[exci, PlotRange → {All, {-0.01, 0.12}},
   PlotStyle → LightOrange, Filling → -0.01, Frame → True,
   FrameStyle → Thickness[0.003], FrameTicksStyle → Directive[Black, 24],
   FrameLabel → {{"Correlation coefficient", ""}, {"Time Lag", ""}},
   LabelStyle -> Directive[Black, 24]], ListLinePlot[exm, PlotStyle → Orange],
  ListLinePlot[{obsl + ciobs, obsl - ciobs}, PlotStyle \rightarrow LightBlue, Filling \rightarrow {1 \rightarrow {2}}],
  ListLinePlot[obsl], ImageSize → 800]
```

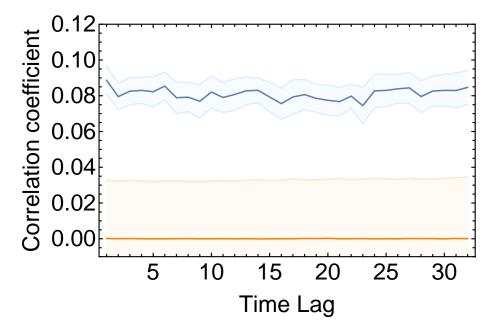


Export["SNconsistency13.tif", SNconsistency13]

SNconsistency13.tif

```
edgelistpen14 = {};
For[d = 1, d ≤ Length[datelist], d++,
 pen = Cases[data, {_, _, _, 14, datelist[d], __}];
 pen = Sort[pen, #1[[6]] < #2[[6]] &];</pre>
 If[Length[pen] > 1,
     dailyedgelist =
   DeleteCases[Tally[Sort /@ DeleteCases[Flatten[ParallelTable[findFriend[i]],
          {i, 1, Length[pen] - 1}], 1], {}]], {{v1_, v2_}, _} /; v1 == v2];
    AppendTo[edgelistpen14, dailyedgelist],
  Print[StringJoin["No data for this day: ", ToString[d]]]
 ]
]
Save["dailyedgelistpen14", edgelistpen14];
edgelistpen14 = Get["dailyedgelistpen14"];
Length[edgelistpen14]
72
wms = Table[WeightedAdjacencyMatrix[Graph[Cases[allhens, {_, "14", __}] [All, 3],
      Apply[UndirectedEdge, edgelistpen14[i, All, 1], 1],
      EdgeWeight → edgelistpen14[i, All, 2]]], {i, Length[edgelistpen14]}];
days = Length[wms];
newacorrSN14 = Table[ParallelTable[
    newpermutationDataOnly[wms[q], wms[i], 1000], {i, q + 1, days}], {q, 1, 40}];
Save["newacorrSN14", newacorrSN14]
obsl = Table[Mean[Table[newacorrSN14[i, lag, 1]], {i, 40}]], {lag, 32}]
ciobs =
 1.96 * (Table [StandardDeviation [Table [newacorrSN14 [i, lag, 1]], {i, 40}]], {lag, 32}] /
    Sqrt [40])
exm = Table[Mean[Table[Mean[newacorrSN14[i, lag, 2]]], {i, 40}]], {lag, 32}]
exci = Table[Mean[Table[Sort[newacorrSN14[i, lag, 2]]][975]], {i, 40}]], {lag, 32}]
```

```
SNconsistency14 = Show[ListLinePlot[exci, PlotRange → {All, {-0.01, 0.12}},
   PlotStyle → LightOrange, Filling → -0.01, Frame → True,
   FrameStyle → Thickness[0.003], FrameTicksStyle → Directive[Black, 24],
   FrameLabel → {{"Correlation coefficient", ""}, {"Time Lag", ""}},
   LabelStyle -> Directive[Black, 24]], ListLinePlot[exm, PlotStyle → Orange],
  ListLinePlot[\{obsl + ciobs, obsl - ciobs\}, PlotStyle \rightarrow LightBlue, Filling \rightarrow \{1 \rightarrow \{2\}\}],
  ListLinePlot[obsl], ImageSize → 800]
```



Export["SNconsistency14.tif", SNconsistency14]

SNconsistency14.tif

### Correlation AI and DTW

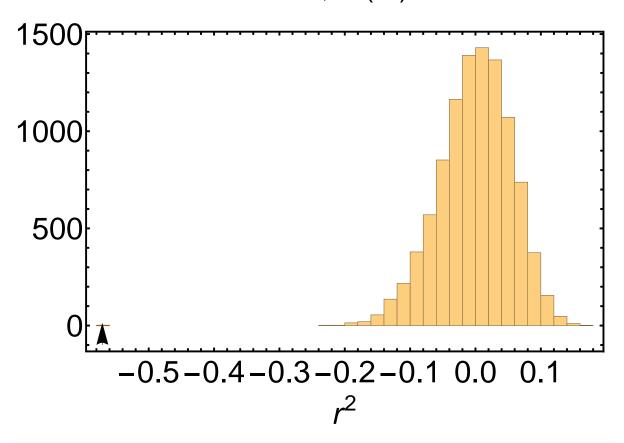
```
dm11 = Get["distmat11"];
 Importing dynamic time warping distance matrices.
dmhead = dm11[[1]];
chicks11 = Table[StringDrop[dmhead[i], -3], {i, 2, Length[dmhead], 72}];
allhens11 = Cases[allhens, {_, "11", __}];
n = Length[chicks11];
dmat = Table[0, {n}, {n}];
```

```
For[i = 1, i ≤ Length[allhens11], i++,
 For [j = 1, j \le Length[allhens11], j++,
  chicki = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks11[i]] ~~ ___]]]];
  chickj = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks11[j]] ~~ \___]]]]; \\
  dmat[i, j] = N[Mean[Table[dm11[chicki[u]], chickj[u]]], {u, Length[chicki]}]]]
Save["dmat11", dmat];
```

This is re-formatting the dtw distance matrix (same format as adjacency matrix from social networks).

```
aimat11 = Get["aimat11"];
correlpen11 = newpermutationR[aimat11 //. "NA" → 0, dmat, 10000]
```

Expected r: -0.000377889 Cl<sub>95</sub>[-0.108191, 0.107 Observed r: -0.571074, N(r+)= 10000  $\sigma$  = 10.3



Correlation of overall (over all days for a pen) adjacency matric with dtw distance matrix using Mantel matrix permutation.

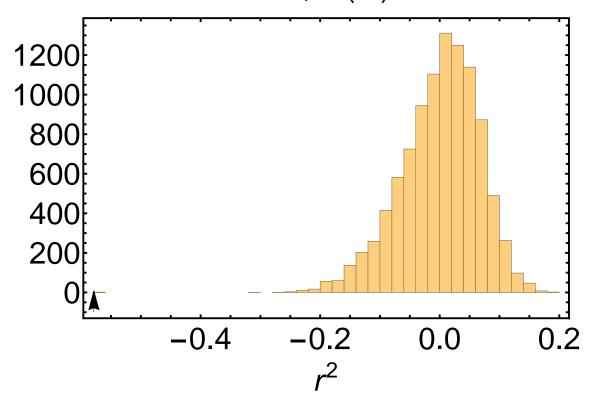
```
Export["DTWxAIcorPen11.tiff", correlpen11]
```

DTWxAIcorPen11.tiff

```
dm12 = Get["distmat12"];
dmhead = dm12[[1]];
chicks12 = Table[StringDrop[dmhead[i]], -3], {i, 2, Length[dmhead], 72}];
allhens12 = Cases[allhens, {_, "12", __}];
n = Length[chicks12];
dmat = Table[0, {n}, {n}];
For [i = 1, i \le Length[allhens12], i++,
 For [j = 1, j \le Length[allhens12], j++,
  chicki = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks12[i]] ~~ ___]]]];
  chickj = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks12[j]] ~~ \___]]]]; \\
  dmat[i, j] = N[Mean[Table[dm12[chicki[u], chickj[u]]], {u, Length[chicki]}]]]
]
]
Save["dmat12", dmat];
dmat = Get["dmat12"];
aimat12 = Get["aimat12"];
```

correlpen12 = newpermutationR[aimat12 //. "NA" → 0, dmat, 10000]

Expected r: 0.000600277 Cl<sub>95</sub>[-0.127299, 0.128 Observed r: -0.578701,  $N(r+)=10000 \sigma = 8.877$ 



Export["DTWxAIcorPen12.tiff", correlpen12]

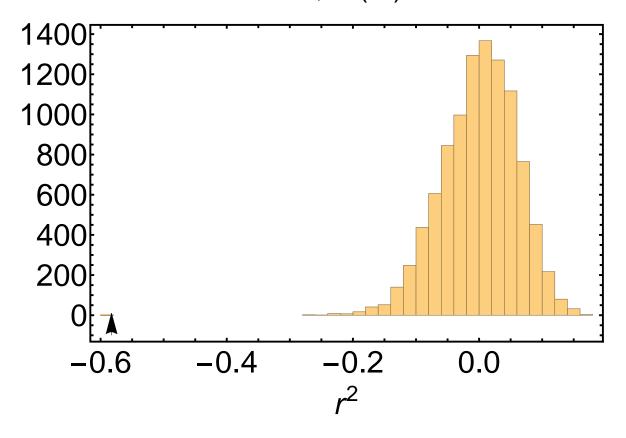
#### Pen 13

DTWxAIcorPen12.tiff

```
dm13 = Get["distmat13"];
dmhead = dm13[1];
chicks13 = Table[StringDrop[dmhead[i], -3], {i, 2, Length[dmhead], 72}];
allhens13 = Cases[allhens, {_, "13", __}];
n = Length[chicks13];
dmat = Table[0, {n}, {n}];
For[i = 1, i ≤ Length[allhens13], i++,
For [j = 1, j \le Length[allhens13], j++,
  chicki = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks13[i]] ~~ ___]]];
  chickj = Quiet[Flatten[Position[dmhead, x_/; StringMatchQ[x, chicks13[j]] ~~ \___]]]]; \\
  dmat[i, j] = N[Mean[Table[dm13[chicki[u]], chickj[u]]], {u, Length[chicki]}]]]
 ]
]
```

```
Save["dmat13", dmat];
dmat = Get["dmat13"];
aimat13 = Get["aimat13"];
correlpen13 = newpermutationR[aimat13 //. "NA" → 0, dmat, 10000]
```

Expected r: 0.000671103 CI<sub>95</sub>[-0.115554, 0.1168 Observed r: -0.582698,  $N(r+)=10000 \sigma = 9.837$ 

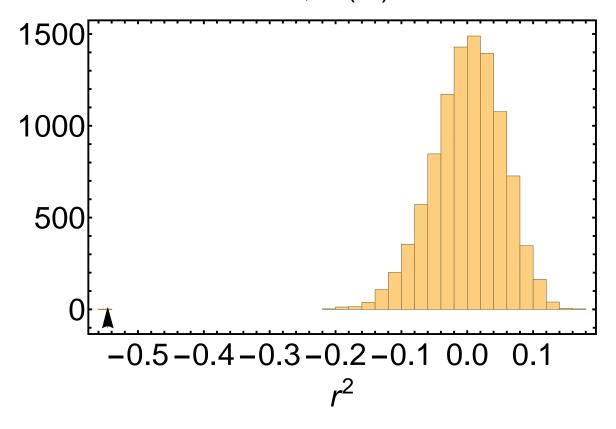


Export["DTWxAIcorPen13.tiff", correlpen13] DTWxAIcorPen13.tiff

```
dm14 = Get["distmat14"];
dmhead = dm14[1];
chicks14 = Table[StringDrop[dmhead[i], -3], {i, 2, Length[dmhead], 72}];
allhens14 = Cases[allhens, {_, "14", __}];
n = Length[chicks14];
dmat = Table[0, {n}, {n}];
```

```
For [i = 1, i \le Length[allhens14], i++,
 For [j = 1, j \le Length[allhens14], j++,
  chicki = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks14[i]] ~~ ___]]]];
  chickj = Quiet[Flatten[Position[dmhead, x_ /; StringMatchQ[x, chicks14[j]] ~~ \___]]]]; \\
  dmat[i, j] = N[Mean[Table[dm14[chicki[u]], chickj[u]]], {u, Length[chicki]}]]]
]
Save["dmat14", dmat];
dmat = Get["dmat14"];
aimat14 = Get["aimat14"];
correlpen14 = newpermutationR[aimat14 //. "NA" → 0, dmat, 10000]
```

Expected r: 0.000562177 Cl<sub>95</sub>[-0.104208, 0.1053 Observed r: -0.54602, N(r+)= 10000  $\sigma$  = 10.22



Export["DTWxAIcorPen14.tiff", correlpen14] DTWxAIcorPen14.tiff

### **DTW Distane Matrix**

Here matrix correlations (Pearson product moment correlations) between AI and DTW distance are calculated for each day.

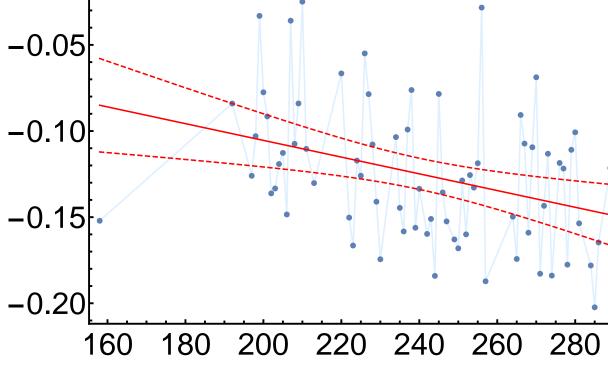
#### Pen 11

```
distmat = Get["distmat11"];
hens = Cases[allhens, {_, "11", __}];
dailyedgelistpen = Get["dailyedgelistpen11"];
order = {51, 13, 25, 34, 16, 55, 66, 44, 5, 26, 35, 18, 57, 69,
   47, 8, 29, 10, 12, 62, 41, 4, 24, 32, 15, 54, 65, 43, 17, 56, 68, 46, 7,
   28, 37, 59, 71, 38, 1, 21, 30, 50, 61, 40, 3, 23, 31, 14, 53, 64, 67, 45,
   6, 27, 36, 19, 58, 70, 48, 9, 20, 11, 49, 60, 39, 2, 22, 52, 63, 42, 33};
 (*dailyedgelistpen and dtw table is not in chronological order*)
wms = Table[WeightedAdjacencyMatrix[
    Graph[hens[All, 3], Apply[UndirectedEdge, dailyedgelistpen[i, All, 1], 1],
      EdgeWeight → dailyedgelistpen[i, All, 2]]], {i, Length[dailyedgelistpen]}];
dm = Delete[
   Table[distmat[Table[i, {i, 2+u, 7777+u, 72}], Table[i, {i, 2+u, 7777+u, 72}]],
     {u, 0, 71}], 13]; (*Oct 10 is deleted because only zeros*)
mcs = Table[ppmcc[dm[order[i]]], wms[order[i]]]], {i, 1, 71}];
(*for pen 11 no data for day 68, so this one was dropped*)
mcs = Transpose[{Drop[dayofyear, {68}], mcs}];
reg = LinearModelFit[mcs, {1, x}, x]
FittedModel [ | -0.00832358-0.000485527 x
```

A linear regression model with day of the year as predictor for the Pearson product moment correlation between DTW and AI.

```
line = reg["BestFit"]
-0.00832358 - 0.000485527 x
cis = reg["MeanPredictionBands"];
```

```
similSNTW = Show[ListLinePlot[mcs, PlotStyle → LightBlue],
                       ListPlot[mcs], Plot[line, \{x, 158, 289\}, PlotStyle \rightarrow Red],
                      Plot[cis[1], \{x, 158, 289\}, PlotStyle \rightarrow \{Red, Dashed\}],
                      \label{eq:plot_cis_2} $$\operatorname{Plot}[\operatorname{cis}_2], \{x, 158, 289\}, \operatorname{PlotStyle} \to \{\operatorname{Red}, \operatorname{Dashed}\}], \operatorname{PlotRange} \to \operatorname{All}, $$\operatorname{Plot}_{\operatorname{Red}, \operatorname{Dashed}}(\operatorname{PlotRange}) \to \operatorname{All}, $$\operatorname{PlotRange} \to \operatorname{All
                      ImageSize → 800, Axes → False, Frame → True, FrameStyle → Directive[Thick, Black, 32],
                      FrameLabel → {"Matrix correlation coefficient for AI and DTW"},
                       LabelStyle → Directive[24, Black]]
```



## Matrix correlation coefficient for AI and D

#### reg["ANOVATable"]

	DF	SS	MS	F-Statistic	P-Value
х	1	0.0142053	0.0142053	9.70249	0.00268009
Error	69	0.101022	0.00146409		
Total	70	0.115227			

#### reg["AdjustedRSquared"]

0.110575

Time (as linear covariate) explains 11% of the overall variation in the correlation coefficients.

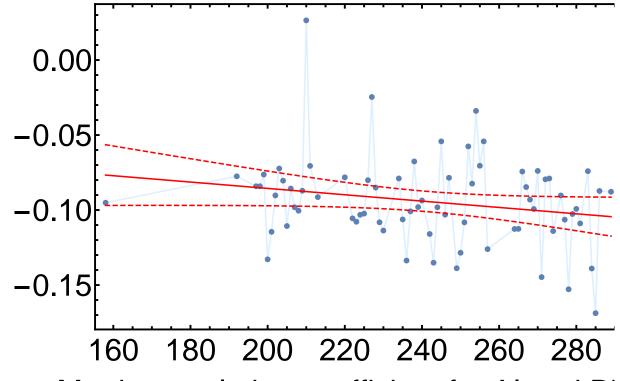
Export["AIxDTWoverTimePen11.tif", similSNTW]

AIxDTWoverTimePen11.tif

```
distmat = Get["distmat12"];
hens = Cases[allhens, {_, "12", __}];
```

```
dailyedgelistpen = Get["dailyedgelistpen12"];
datelist = Get["datelist"]; (*sorted*)
order = {52, 14, 26, 35, 17, 56, 67, 45, 5, 27, 36, 19, 58, 70, 48,
   8, 30, 10, 12, 63, 42, 4, 25, 33, 16, 55, 66, 44, 18, 57, 69, 47, 7, 29,
   38, 60, 72, 39, 1, 22, 31, 51, 62, 41, 3, 24, 32, 15, 54, 65, 68, 46, 6,
   28, 37, 20, 59, 71, 49, 9, 21, 11, 50, 61, 40, 2, 23, 13, 53, 64, 43, 34};
 (*dailyedgelistpen and dtw table is not in chronological order*)
wms = Table[WeightedAdjacencyMatrix[
    Graph[hens[All, 3], Apply[UndirectedEdge, dailyedgelistpen[i, All, 1], 1],
      EdgeWeight → dailyedgelistpen[i, All, 2]]], {i, Length[dailyedgelistpen]}];
dm = Table[distmat[Table[i, {i, 2 + u, 7417 + u, 72}],
    Table[i, {i, 2 + u, 7417 + u, 72}]], {u, 0, 71}];
mcs = Table[ppmcc[dm[order[i]]], wms[order[i]]]], {i, 1, 72}];
mcs = Transpose[{dayofyear, mcs}];
reg = LinearModelFit[mcs, {1, x}, x]
FittedModel [ | -0.0432518 - 0.000211642 x | ]
line = reg["BestFit"]
-0.0432518 - 0.000211642 x
cis = reg["MeanPredictionBands"];
```

```
similSNTW = Show[ListLinePlot[mcs, PlotStyle → LightBlue, PlotRange → All],
                       ListPlot[mcs], Plot[line, \{x, 158, 289\}, PlotStyle \rightarrow Red],
                     Plot[cis[1]], \{x, 158, 289\}, PlotStyle \rightarrow \{Red, Dashed\}],
                     \label{eq:plot_cis_2} $$\operatorname{Plot}[\operatorname{cis}_2], \{x, 158, 289\}, \operatorname{PlotStyle} \to \{\operatorname{Red}, \operatorname{Dashed}\}], \operatorname{PlotRange} \to \operatorname{All}, $$\operatorname{Plot}_{\operatorname{Red}, \operatorname{Dashed}}(\operatorname{PlotRange}) \to \operatorname{All}, $$\operatorname{PlotRange} \to \operatorname{All
                     ImageSize → 800, Axes → False, Frame → True, FrameStyle → Directive[Thick, Black, 32],
                       FrameLabel → {"Matrix correlation coefficient for AI and DTW"},
                       LabelStyle → Directive[24, Black]]
```



## Matrix correlation coefficient for AI and Di

#### reg["ANOVATable"]

	DF	SS	MS	F-Statistic	P-Value
Х	1	0.00278007	0.00278007	3.3546	0.0712725
			0.000828733		
Total	71	0.0607914			

reg["AdjustedRSquared"]

0.0320989

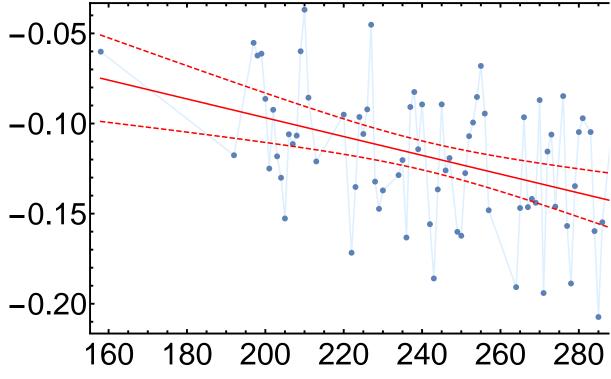
Export["AIxDTWoverTimePen12.tif", similSNTW]

AIxDTWoverTimePen12.tif

```
distmat = Get["distmat13"];
hens = Cases[allhens, {_, "13", __}];
dailyedgelistpen = Get["dailyedgelistpen13"];
```

```
datelist = Get["datelist"]; (*sorted*)
order = {52, 14, 26, 35, 17, 56, 67, 45, 5, 27, 36, 19, 58, 70, 48,
   8, 30, 10, 12, 63, 42, 4, 25, 33, 16, 55, 66, 44, 18, 57, 69, 47, 7, 29,
   38, 60, 72, 39, 1, 22, 31, 51, 62, 41, 3, 24, 32, 15, 54, 65, 68, 46, 6,
   28, 37, 20, 59, 71, 49, 9, 21, 11, 50, 61, 40, 2, 23, 13, 53, 64, 43, 34};
 (*dailyedgelistpen and dtw table is not in chronological order*)
wms = Table[WeightedAdjacencyMatrix[
     Graph[hens[All, 3], Apply[UndirectedEdge, dailyedgelistpen[i, All, 1], 1],
      EdgeWeight → dailyedgelistpen[i, All, 2]]], {i, Length[dailyedgelistpen]}];
Length[distmat]
7633
Length[dailyedgelistpen]
72
dm = Table[distmat[Table[i, {i, 2 + u, 7632 + u, 72}],
     Table[i, {i, 2 + u, 7632 + u, 72}]], {u, 0, 71}];
mcs = Table[ppmcc[dm[order[i]]], wms[order[i]]]], {i, 1, 72}];
mcs = Transpose[{dayofyear, mcs}];
reg = LinearModelFit[mcs, {1, x}, x]
FittedModel | 0.00751738 - 0.000521575 x
line = reg["BestFit"]
0.00751738 - 0.000521575 x
cis = reg["MeanPredictionBands"];
```

```
similSNTW = Show[ListLinePlot[mcs, PlotStyle → LightBlue, PlotRange → All],
                       ListPlot[mcs], Plot[line, \{x, 158, 289\}, PlotStyle \rightarrow Red],
                     Plot[cis[1]], \{x, 158, 289\}, PlotStyle \rightarrow \{Red, Dashed\}],
                     \label{eq:plot_cis_2} $$\operatorname{Plot}[\operatorname{cis}_2], \{x, 158, 289\}, \operatorname{PlotStyle} \to \{\operatorname{Red}, \operatorname{Dashed}\}], \operatorname{PlotRange} \to \operatorname{All}, $$\operatorname{Plot}[\operatorname{Cis}_2], \{x, 158, 289\}, \operatorname{PlotStyle} \to \{\operatorname{Red}, \operatorname{Dashed}\}], $$\operatorname{PlotRange} \to \operatorname{All}, $$\operatorname
                       ImageSize → 800, Axes → False, Frame → True, FrameStyle → Directive[Thick, Black, 32],
                       FrameLabel → {"Matrix correlation coefficient for AI and DTW"},
                       LabelStyle → Directive[24, Black]]
```



## Matrix correlation coefficient for AI and D

#### reg["ANOVATable"]

	DF	SS	MS	F-Statistic	P-Value
х	1	0.0168844	0.0168844	14.5463	0.000291579
Error	70	0.0812515	0.00116074		
Total	71	0.0081350			

#### reg["AdjustedRSquared"]

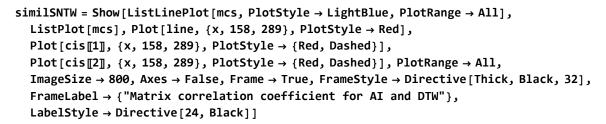
0.160224

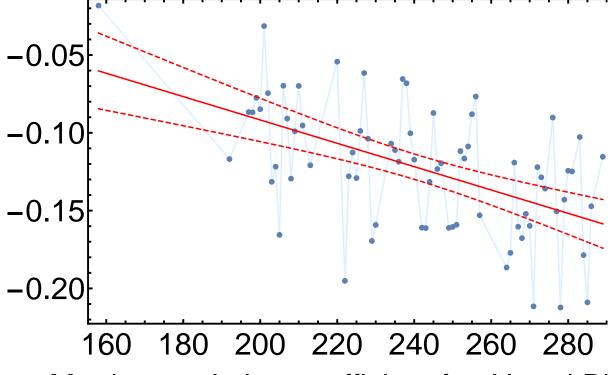
Export["AIxDTWoverTimePen13.tif", similSNTW]

AIxDTWoverTimePen13.tif

```
distmat = Get["distmat14"];
hens = Cases[allhens, {_, "14", __}];
dailyedgelistpen = Get["dailyedgelistpen14"];
```

```
datelist = Get["datelist"]; (*sorted*)
order = {52, 14, 26, 35, 17, 56, 67, 45, 5, 27, 36, 19, 58, 70, 48,
   8, 30, 10, 12, 63, 42, 4, 25, 33, 16, 55, 66, 44, 18, 57, 69, 47, 7, 29,
   38, 60, 72, 39, 1, 22, 31, 51, 62, 41, 3, 24, 32, 15, 54, 65, 68, 46, 6,
   28, 37, 20, 59, 71, 49, 9, 21, 11, 50, 61, 40, 2, 23, 13, 53, 64, 43, 34};
 (*dailyedgelistpen and dtw table is not in chronological order*)
wms = Table[WeightedAdjacencyMatrix[
     Graph[hens[All, 3], Apply[UndirectedEdge, dailyedgelistpen[i, All, 1], 1],
      EdgeWeight → dailyedgelistpen[i, All, 2]]], {i, Length[dailyedgelistpen]}];
Length[distmat]
7489
Length[dailyedgelistpen]
72
dm = Table[distmat[Table[i, {i, 2 + u, 7489 + u, 72}],
     Table[i, {i, 2 + u, 7489 + u, 72}]], {u, 0, 71}];
mcs = Table[ppmcc[dm[order[i]]], wms[order[i]]]], {i, 1, 72}];
mcs = Transpose[{dayofyear, mcs}];
reg = LinearModelFit[mcs, {1, x}, x]
FittedModel | 0.0582984 - 0.000750173 x
line = reg["BestFit"]
0.0582984 - 0.000750173 x
cis = reg["MeanPredictionBands"];
```





## Matrix correlation coefficient for AI and D'

#### reg["ANOVATable"]

	DF	SS	MS	F-Statistic	P-Value
x	1	0.0349282	0.0349282	29.1398	$8.70975 \times 10^{-7}$
Error	70	0.083905	0.00119864		
Total	71	0.118833			

#### reg["AdjustedRSquared"]

0.283839

Export["AIxDTWoverTimePen14.tif", similSNTW]

AIxDTWoverTimePen14.tif

#### Combined probability

```
chisquared = -2 \left( Log[0.027] + Log[0.071] + Log[0.0003] + Log[8.7 \times 10^{-7}] \right)
56.647
```

```
N[Probability[x >= chisquared, x \approx ChiSquareDistribution[8]]]
2.11011 \times 10^{-9}
```

### **Random Numbers**

For the time shift test a random list is created. 36 days (half of all days) are randomly selected, on each day 100 dyads are selected for each pen (N=14400 in total). For these dyads DTW distances are re-calculated after shifting one time series by 60 seconds (done in R) and AI are too re-calculated (next section).

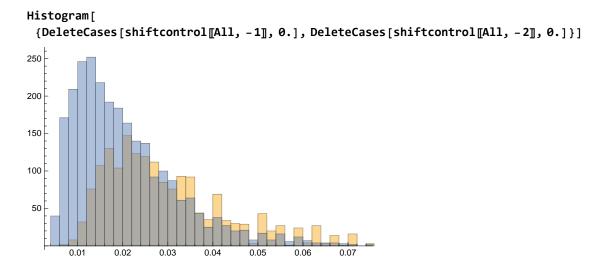
```
In[*]:= SeedRandom[54123]
In[*]:= daysample = Sort[RandomSample[Range[72], 36]]
37, 39, 41, 43, 46, 50, 51, 54, 55, 59, 60, 61, 63, 65, 66, 70, 71, 72}
In[*]: randomlist = Round[Flatten[Table[Table[Table[Join[{daysample[d]]},
            RandomSample[Cases[allhens, {_, ToString[p], __}], 2] [All, 1]],
           \{100\}], \{d, Length[daysample]\}], \{p, 11, 14\}], 2]];
    Export["randomlist.csv", randomlist]
    randomlist.csv
```

### **Shift Control**

```
In[*]:= datelist = datelist[Ordering[Table[AbsoluteTime[{datelist[i]],
            {"DayName", " ", "Day", "/", "Month", "/", "Year"}}], {i, Length[datelist]}]]];
```

```
shiftcontrol = {};
For[i = 1, i ≤ Length[randomlist], i++,
pair = randomlist[i];
pennumber = ToExpression[Cases[allhens, {pair[2], __}][1, 2]];
day = datelist[Round[pair[1]]];
bird1 = Cases[allhens, {pair[2], __}] [1, 3];
bird2 = Cases[allhens, {pair[3], __}][1, 3];
 pen = Cases[data, {_, _, _, pennumber, day, __}];
pen = Sort[penday, #1[[6]] < #2[[6]] &];</pre>
 shift = 0;
 friends = Flatten[ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}], 1];
 ai = (Count[friends, {bird1, bird2}] + Count[friends, {bird2, bird1}]) /
   (Count[friends, {___, bird1, ___}] + Count[friends, {___, bird2, ___}] -
     Count[friends, {bird1, bird2}] - Count[friends, {bird2, bird1}]);
 shift = 60;
 shiftedfriends = Flatten[ParallelTable[findFriend[i], {i, 1, Length[pen] - 1}], 1];
 shiftedai =
  (Count[shiftedfriends, {bird1, bird2}] + Count[shiftedfriends, {bird2, bird1}]) /
   (Count[shiftedfriends, {___, bird1, ___}] +
     Count[shiftedfriends, {___, bird2, ___}] -
     Count[shiftedfriends, {bird1, bird2}] - Count[shiftedfriends, {bird2, bird1}]);
AppendTo[shiftcontrol, {day, pennumber, bird1, bird2, N[ai], N[shiftedai]}]
```

Here AI is calculated for a list of dyads (for given days and pens) twice: once normal and once after shifting the transition times of one bird by 5 seconds. The outcome is a matrix, where each line gives: day, pennumber, birdID1, birdID2, unshifted AI, shifted AI.



Histogram showing Als of randomly chosen dyads without and with time shift of 60s.

```
shiftcontrol = shiftcontrol //. Indeterminate → 0.;
Save["shiftcontrol", shiftcontrol]
shiftcontrol = Prepend[shiftcontrol,
   {"Date", "Pen", "Individual 1", "Individual 2", "AI", "shifted AI"}];
```

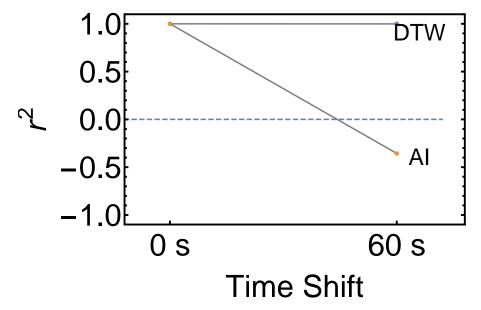
```
Export["shiftcontrol.csv", shiftcontrol]
shiftcontrol.csv
```

## Test with Time Shifted Data

Here the Pearson product moment correlation of the AI of the unshifted and shifted time series and of the DTW distances of unshifted and shifted time series are calculated and graphed.

```
aiunshiftedshifted = Import["shiftcontrol.csv"];
 dtwunshifted = Import["randomlistunshiftedconverted.csv"];
 dtwshifted = Import["randomlist_full_new.csv"]; (*file "randomlist" was re-named*)
makematrix[line_] := Module[{in, dtws, hex1, hex2, date},
  in = dtwunshifted[line];
  dtws = Cases[dtwshifted, {_, in[1], in[2], in[3], _}][1, 5];
  hex1 = Cases[allhens, {N[in[2]], __}][1, 3];
  hex2 = Cases[allhens, {N[in[3]], __}][1, 3];
  date = datelist[in[1]];
  Join[in, {dtws}, Cases[aiunshiftedshifted, {date, _, hex1, hex2, __}] [1, 5;; 6]]
timeshifttable = Prepend[Table[makematrix[i], {i, 2, Length[dtwunshifted]}],
   {"day", "hen1", "hen2", "dtw", "dtwshifted", "ai", "aishifted"}];
Save["timeshifttable", timeshifttable]
tstshort = DeleteCases[timeshifttable, {__, 0., 0., ___}];
tstshort // Length
3830
dtwcorr = Transpose[{Drop[tstshort[All, 4], 1], Drop[tstshort[All, 5], 1]}];
N[Correlation[dtwcorr]]
\{\{1., 0.999906\}, \{0.999906, 1.\}\}
Covariance Matrix for DTW
aicorr = Transpose[{Drop[tstshort[All, 6]], 1], Drop[tstshort[All, 7]], 1]}];
N[Correlation[aicorr]]
\{\{1., -0.355851\}, \{-0.355851, 1.\}\}
Covariance Matrix for Al
```

```
scplot = Show[ListLinePlot[{{{1, 1}, {2, 0.999}}}, {{1, 1}, {2, -0.356}}},
                       PlotRange \rightarrow \{\{0.8, 2.3\}, \{-1.1, 1.1\}\}, PlotStyle \rightarrow Gray, PlotRange \rightarrow All],
              ListPlot[\{\{1, 1\}, \{2, 0.999\}\}, \{\{1, 1\}, \{2, -0.356\}\}\}],
              Plot[0, \{x, 0.8, 2.2\}, PlotStyle \rightarrow Dashed],
              Graphics[{Inset[Text[Style["DTW", Large]], {2.1, 0.9}],
                               Inset[Text[Style["AI", Large]], \{2.1, -0.4\}]\}], ImageSize \rightarrow 800, Axes \rightarrow False, Axes 
              Frame \rightarrow True, FrameTicks \rightarrow {{True, True}, {{{1, "0 s"}}, {2, "60 s"}}, False}},
              FrameStyle → Directive[Thick, Black, 32],
               FrameLabel → {"Time Shift", "r²"}, LabelStyle → Directive[24, Black]]
```



Export["shiftcontrolplot.tiff", scplot]

shiftcontrolplot.tiff