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IAIS

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The science of analytical reasoning facilitated by interactive visual interfaces.

Visual Analytics

Visual analytics tools and techniques are used to

- Synthesize information and derive insight from massive, dynamic, ambiguous, and often conflicting data
- Detect the expected and discover the unexpected
- Provide timely, defensible, and understandable assessments
- Communicate assessment effectively for action



Automatic Data Analysis



IEEE Computer Society 2005 http://nvac.pnl.gov/



EuroGraphics 2010 http://www.vismaster.eu/book



Illuminating the Path Research and Development Agenda for Visual Analytics

Edited by James J. Thomas and Kristin Cook

Visual Analytics of Movement Data: Principal Transformations



Different representations of movement data suitable for different analysis tasks / reporting requirements



Principal Transformations of Movement Data





Visual Data Exploration – Data Understanding & Data Quality



Movement happens in context – airspace design is a 4D spatio-temporal puzzle

Identification of most common types of errors, data gaps, ingestion problems

- ADS-B data: positional messages aggregated into trajectories
- Flight Plans + Regulations
- DDR Airblock/Configuration/Capacity data







Visualizations of DART solutions

The capacity excess events are shown in a space-time cube based on the original (red) and CFMU-regulated (blue) flight data. The vertical dimension, from bottom to top, represents time.



Visualizations of solutions



Flight delays are represented by circles positioned at the sector centroids.

The sizes are proportional to the delay durations.

From top to bottom and from left to right: CFMU, AgentBased, Hierarchical, IndLearners, EdgeBased.

Visualizations of solutions





The space-time cubes show the spatio-temporal distribution of the delays. The time axis is oriented upwards. From top to bottom and from left to right: CFMU, AgentBased, Hierarchical, IndLearners, EdgeBased.

Visually supported Trajectory Comparison



Why compare flight trajectories?

- Flight plan vs. actual flight trajectory
- Two versions of flight plan
- Actual trajectory vs. a typical trajectory of a given flight / representative of a cluster
- Actual trajectory vs. predicted trajectory
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Visually supported Trajectory Comparison





Visually supported Trajectory Comparison



Key ingredient: Computing and visualizing pairwise distances

```
Description of the algorithm:
let M = \langle (1,1) \rangle;
                                                             /* list M contains the pair (1,1) */
let i = 2; let j = 2;
while i <= P.length and j <= Q.length do
                                                               /* scan the trajectories from the start to the end */
   let cp = next candidate pair (P, Q, M, i, j);
                                                               /* account for different inter-point distances, if needed */
   let i = cp.first; let j = cp.second;
                                                               /* indexes of currently considered points in P and Q */
   let mi = M.lastElement[1]; let mj = M.lastElement[2];
                                                                 /* indexes of the last matched points from P and Q */
   let n = argmin(distance(P[i],Q[j]), distance(P[i], Q[mj]), distance(P[mi], Q[j]), distance(P[mi], Q[mi]));
   if n = 2 and
                                                  /* point P[i] is a better match to point Q[mi] than P[mi] */
     distance(P[i], Q[mj]) < distance(P[i], Q[j+1]) /* point P[i] does not match Q[j+1] better than Q[mj] */
   then
       let M = M - (mi, mj) + (i, mj);
                                                    /* replace the pair (mi, mj) in M by (i, mj) */
                                                  /* proceed to the next point in P */
       let i=i+1:
    end if:
    else
      if n = 3 and
                                                    /* point Q[j] is a better match to point P[mi] than Q[mj] */
        distance(P[mi], Q[j]) < distance(P[i+1], Q[j]) /* point Q[j] does not match P[i+1] better than P[mi] */
      then
          let M = M - (mi, mj) + (mi, j);
                                                       /* replace the pair (mi, mj) in M by (mi, j) */
          let i = i +1;
                                                     /* proceed to the next point in Q */
       end_if;
       else
         let n = argmin(distance(P[i],Q[j]), distance(P[i], Q[j+1]), distance(P[i+1], Q[j]));
                                                           /* point Q[i+1] is a better match to point P[i] than Q[i] */
         if n = 2 and
           distance(P[i], Q[j+1]) < distance(P[i+1], Q[j+1]) /* point Q[j+1] is not a better match to P[i+1] than to P[i] */
         then
                                                    /* proceed to the next point in Q */
             let j = j + 1;
         end_if;
         else
            if n = 3 and
                                                             /* point P[i+1] is a better match to point Q[i] than P[i] */
              distance(P[i+1], Q[j]) < distance(P[i+1], Q[j+1]) /* point P[i+1] is not a better match to Q[j+1] than to Q[j] */
            then
               let i = i +1;
                                                    /* proceed to the next point in P */
             end if;
             else
               let M = M + (i,j);
                                                   /* add the pair (i, j) to M */
               let i = i +1; let j = j +1;
                                                   /* proceed to the next points in P and Q */
            end else:
         end else:
      end else:
    end else:
end_while;
return M;
```

| function next_candidate_pair (P, Q, M, i, j): | |
|--|----|
| if i > P.length or j > Q.length return null; | |
| <pre>let mi = M.lastElement[1]; let mj = M.lastElement[2];</pre> | / |
| (mi, mj) is the last matched pair */ | |
| <pre>let d1 = distance(P[i], P[mi]); let d2 = distance(Q[j], Q[mj]);</pre> | |
| /* inter-point distances in P and Q */ | |
| if d1 > d2 then | |
| while d1 > d2 * TT and j < Q.length do | /* |
| much larger inter-point gap in P */ | |
| <pre>let j = j+1; let d2 = distance(Q[j], Q[mj]);</pre> | /* |
| proceed to the next point in Q */ | |
| end_while; | |
| end_if; | |
| else | |
| <pre>while d2 > d1 * TT and i < P.length do</pre> | /* |
| much larger inter-point gap in Q */ | |
| <pre>let i = i+1; let d1 = distance(P[i], P[mi]);</pre> | /* |
| proceed to the next point in P */ | |
| end_while; | |
| end_else; | |
| return (i, j); | |

WP2: Single Trajectory Prediction



D2.2 Visually supported Trajectory Comparison





WP2: Single Trajectory Prediction



D2.2 Visually supported Trajectory Comparison











Case Study I – Approach Route Analysis



This project has received funding from the SESAR Joint Undertaking under the European Union's Horizon 2020 research and innovation programme under grant agreement No [number]





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Task: Extract the airport approach routes and observe the wind impact





Challenge: The holding loops are <u>not</u> essential parts of the approach routes

Visual support to interactive filtering

Workflow: <u>*filter*</u> \rightarrow cluster \rightarrow summarize \rightarrow analyse



Filter-aware rendering

Filtering \rightarrow Boolean attribute \rightarrow visual encoding Different level of detail for active and inactive parts

Progressive clustering



Workflow: *filter* \rightarrow *cluster* \rightarrow *summarize* \rightarrow *analyse*

Application of clustering with different distance functions or parameter settings to different data subsets

Particularly, subsets may be defined by previous clustering runs

Useful when clustering of the whole set produces clusters of differing quality

 "Bad" clusters (with high internal variation) can be refined through further clustering and good clusters can be preserved





An example of cluster refinement





Workflow: *filter* \rightarrow *cluster* \rightarrow *summarize* \rightarrow *analyse*







Loop statistics by the routes





| N | l members | N | % | Loop | Loop | Loop | Loop | Loop | Loop | Loop | Loop |
|---|------------------|--------------|--------------|------------------------|------------|------------|--------------------|------------|------------|------------|-------------------------|
| | | trajectories | trajectories | duration | duration | duration | duration | duration | duration | duration | duration |
| L | | with loops | with loops | (minutes), | (minutes), | (minutes), | (minutes), | (minutes), | (minutes), | (minutes), | (minutes), |
| L | | | | sum | mean | std.dev. | min | q1 | median | q3 | max |
| | 9 <mark>9</mark> | | | | | | | | | | |
| | 36 | | | | | | | | | | |
| | 36 | | | | | | | | | | |
| | 11 | | | | | | | | | | |
| | 32 | | | | | | | | | | |
| | 258 | 184 | 71.3 | 3 1892.10 | 10.283 | 4.4243 | 2.02 | 7.03 | 8.78 | 12.72 | 29. |
| | 152 | 91 | 59.9 | 3 784.30 | 8.619 | 3.7594 | 2.68 | 5.82 | 8.27 | 11.02 | 19. |
| - | 137 | 80 | 58.4 | 4 644 <mark>.92</mark> | 8.061 | 3.4982 | 2.78 | 5.33 | 7.14 | 10.80 | 17. |
| | 59 | 34 | 57.6 | <u>336.4</u> 0 | 9.894 | 4.0622 | 3.15 | 7.04 | 9.22 | 12.64 | 17. |
| L | 89 | 50 | 56. | 402.63 | 8.053 | 3.6846 | 3.17 | 6.00 | 6.75 | 10.83 | 17. |
| | 342 | 1// | 51.0 | 3 1 <u>340.78</u> | 7.575 | 4.1491 | 1.78 | 4.25 | 6.37 | 9.97 | 19. |
| | 141 | <u>/1</u> | 50.4 | 4 764.42 | 10.766 | 4.8091 | 4.((| 7.55 | 9.28 | 12.87 | 25. |
| | /46 | 374 | 50.1 | 2534.43 | 6.777 | 3.6097 | 2.52 | 3.86 | 4.97 | 9.34 | 19. |
| _ | 12 | 6 | 50.0 | 42.42 | 7.069 | 2.7334 | 4.58 | 4.73 | 5.63 | 10.85 | 10 |
| _ | 10 | 5 | 50.0 | J 46.17 | 9.233 | 2.2277 | 7.15 | 7.33 | 8.07 | 11.72 | 13. |
| _ | 314 | 130 | 43. | 141.33 | 9.172 | 4.5198 | 3.13 | 4.30 | 9.18 | 12.03 | 22. |
| _ | 30 | 15 | 42. | 1 132.80 | 8.853 | 2.3794 | 4.92 | 1.13 | 8.20 | 12.00 | 13. |
| | 104 | 0 0 | 30.0 | 0 510.07 | 8.001 | 2.7680 | 2 <mark>.95</mark> | 0.00 | 8.71 | 10.23 | 19. |
| | 113 | 23 | 20.4 | 4 128.00 | 5.505 | 1.9033 | 1.52 | 4.07 | 5,15 | 5.83 | 10. |
| | 84 | 14 | 10. | 7 109.05 | 1.189 | 2.2088 | 4.90 | 0.38 | 7.10 | 8.08 | 1 <mark>3.</mark> 12 |
| | 10 | 8 | 10. | 7 74.07 | 9.333 | 2.4880 | 3.82 | 8.14 | 9.82 | 7.40 | 12. |
| ļ | 170 | 10 | 9. | 32.90 | 0.080 | 0.8281 | 0.28 | 0.77 | 0.57 | 7.40 | 1. |
| | 12 | 10 | 0. 7 | 7 6.70 | 6,700 | 0.8754 | 0.37 | 6.07 | 0.73 | 7.80 | .0 A |
| 1 | 13 | 7 | 1. | 7 52.02 | 7.421 | 2 1660 | 6.20 | 5.67 | 6.10 | 9.0 | 0. |
| h | 123 64 | 2 | J., | 22.02 | 7.431 | 2,1000 | 4 70 | 3.07 | 9.62 | 0.03 | |
| | 40 0 a | 3 | 4. | 17.49 | 5 222 | 1 8267 | 2 22 | 3.73 | 7.02 | 7.22 | 3 |
| | 206 | 13 | 4 | 87.90 | 6 754 | 1 7163 | 4.23 | 5.03 | 6.75 | 8.47 | (. Q |
| | 201 | 13 | 4 | 41 38 | 5173 | 0.8310 | 3.88 | 4 50 | 513 | 5 95 | 5 |
| | 83 | | 4.0 | 20.87 | 6,956 | 1 8969 | 4 47 | 4.00 | 7.33 | 9.07 | 9. 9 |
| | 61 | 2 | 3 | 12 78 | 6 392 | 0.1250 | 6.27 | 6.27 | 6.39 | 6.50 | 6 |
| | 33 | 1 | 31 | 5.67 | 5 667 | 0.1230 | 5.67 | 5.67 | 5.67 | 5.67 | 5 |
| | 363 | 0 | 21 | 45.70 | 5.078 | 1 1 4 67 | 3.07 | 4 23 | 4.51 | 6.26 | 7 |
| F | 112 | 1 | 0.1 | 4 00 | 4 000 | 0.0000 | 4.00 | 4.0 | 4.00 | 4.00 | 4 |
| 1 | / | | | 4.00 | 4.000 | 0.0000 | | 4.00 | 4.00 | 4.00 | 4.1 |

🗌 group by classe Dy: |% trajectories with loops Descending ~ 14

Use of the routes over time



KCRAT'18 Data-Enhanced TBO Workshop, Barcelona 2018-06-25

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SESA JOINT UNDERTAKIN

Understanding the wind impact



Workflow: *filter* \rightarrow *cluster* \rightarrow *summarize* \rightarrow *analyse*

Consol LUTOR - CONSOLIST ANSTED Consol LUTOR - CONSOLIST ANSTED CONSOLITION - CONSOLITION



December 2016 Weather in London — Graph



Fraunhofer

The routes to Stansted





Workflow: *filter* \rightarrow *cluster* \rightarrow *summarize* \rightarrow *analyse*

datAcron









Case Study II – Route Choice Criteria Analysis



This project has received funding from the SESAR Joint Undertaking under the European Union's Horizon 2020 research and innovation programme under grant agreement No [number]





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Task: Identify and compare the major flight routes from Paris to Istanbul





Gennady Andrienko, Natalia Andrienko, Georg Fuchs, Jose Manuel Cordero Garcia

Clustering Trajectories by Relevant Parts for Air Traffic Analysis

IEEE Transactions on Visualization and Computer Graphics (proceedings IEEE VAST 2017), 2018, vol. **24**(1): 34-44

The movements around the airports are not parts of the routes





Route popularity and navigation charges





| | N members | charges, min | charges, q1 | charges, mean | charges, median | charges, q3 | charges, max | | | |
|---|-------------------|---------------------|---------------------|----------------------|--------------------|--------------------|---------------------|--|--|--|
| | | | | | | | | | | |
| 2 | 217 | 434.9 | 457.8 | 459.10 | 459.4 | 460.2 | 492.8 | | | |
| 4 | 10 <mark>5</mark> | 4 <mark>86.4</mark> | 5 <mark>08.1</mark> | 5 <mark>12.67</mark> | <mark>513.8</mark> | 517.0 | 52 <mark>8.2</mark> | | | |
| 3 | 14 | 47 <mark>6.6</mark> | <mark>512.4</mark> | 51 <mark>1.43</mark> | 515.5 | <mark>517.2</mark> | 517 <mark>.5</mark> | | | |
| 1 | 1031 | 47 <mark>2.2</mark> | <mark>512.8</mark> | 5 <mark>14.36</mark> | <mark>515.6</mark> | <mark>517.8</mark> | 5 <mark>47.3</mark> | | | |
| 3 | 12 <mark>5</mark> | <mark>500.6</mark> | <mark>519.2</mark> | 5 <mark>21.45</mark> | <mark>521.8</mark> | <mark>523.2</mark> | 53 <mark>6.5</mark> | | | |
| 7 | 20 | 519.9 | 521.3 | 5 <mark>22.90</mark> | 522.2 | <mark>523.0</mark> | 53 <mark>5.9</mark> | | | |
| 5 | 7 <mark>6</mark> | 509.7 | 533.7 | 539.24 | 540.3 | 542.7 | 573.1 | | | |
| noise | 91 | 454.2 | 510.0 | 565.12 | 544.6 | 650.0 | 745.8 | | | |
| 3 | 11 | 609.3 | 609.4 | 610.60 | 609.6 | 612.5 | 614.2 | | | |
| 6 | 27 | 646.6 | 654.2 | 655.47 | 655.7 | 656.4 | 660.3 | | | |
| | ₹ | | | | | | E. | | | |
| 5 76 509.7 533.7 539.24 540.3 542.7 573.1 noise 91 454.2 510.0 565.12 544.6 650.0 745.8 3 11 609.3 609.4 610.60 609.6 612.5 614.2 4 27 646.6 654.2 655.47 655.7 656.4 660.3 9 group by classes Sort by: charges, median Ascending TableLens condensed Attribute. | | | | | | | | | | |

Route choices







Operator



| | N members | charges, min | charges, q1 | charges, mean | charges, median | charges, q3 | charges, max |
|------|-------------------|---------------------|---------------------|----------------------|--------------------|--------------------|---------------------|
| | | | | | | | |
| | 217 | 434.9 | 457.8 | 459.10 | 459.4 | 460.2 | 492.8 |
| | 10 <mark>5</mark> | 4 <mark>86.4</mark> | 5 <mark>08.1</mark> | 5 <mark>12.67</mark> | <mark>513.8</mark> | 517.0 | 52 <mark>8.2</mark> |
| | 14 | 47 <mark>6.6</mark> | 512.4 | 51 <mark>1.43</mark> | <mark>515.5</mark> | <mark>517.2</mark> | 517 <mark>.5</mark> |
| | 1031 | 47 <mark>2.2</mark> | 512.8 | 5 <mark>14.36</mark> | 515.6 | 517.8 | 5 <mark>47.3</mark> |
| | 1 <mark>25</mark> | <mark>500.6</mark> | <mark>519.2</mark> | 5 <mark>21.45</mark> | <mark>521.8</mark> | <mark>523.2</mark> | 53 <mark>6.5</mark> |
| | 20 | 519.9 | 521.3 | 522.90 | 522.2 | 523.0 | 53 <mark>5.9</mark> |
| | 7 <mark>6</mark> | 509.7 | 533.7 | 539.24 | 540.3 | 542.7 | 573.1 |
| oise | 91 | 454.2 | 510.0 | 565.12 | 544.6 | 650.0 | 745.8 |
| | 11 | 609.3 | 609.4 | 610.60 | 609.6 | 612.5 | 614.2 |
| | 27 | 646.6 | 654.2 | 655.47 | 655.7 | 656.4 | 660.3 |
| | < | | | | | | P. |
| _ | | | | | | | |

group by classes – Sort by: | charges, median ▼ ▼ TableLens ▼ condensed Attribute... Г Ascending N members length, min length, q1 length, mean length, median length, q3 length, max 76 1242.0 1255.3 1269.17 1268.5 1276.3 1321.2 14 1261.8 1272.19 1272.8 1283.5 1253. 1287.2 20 1272 1278.8 1274 1282 1301. 1186.8 127 1285. 128 12<mark>94</mark>. 13<mark>15</mark> 12 1266. 12<mark>75.</mark> 128<mark>5.8</mark> 12<mark>88.</mark> 12<mark>94.1</mark> 13<mark>17.</mark> 10 1225 127 1285 12 12<mark>94</mark>. 131 1243.1 1304.98 noise 91 1280.2 1295.4 1320.4 1409.2 27 1294.6 1296. 1302. 13<mark>15.</mark> 1283. 1243. 1296.4 1304. 1325 11 1338.2 1340.8 1349.56 1349.0 1352.9 1377.3 🗸 Ascending 🛛 🗸 🗹 TableLens 🗹 condensed Attribute.. group by classes Sort by: length, median

ICRAT'18 Data-Enhanced TBO Workshop, Barcelona 2018-06-25

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Route variability









Visual Analytics – human expert in the loop "Detect the expected, discover the unexpected"

- Data Exploration & Understanding (raw data)
- Pattern Extraction & Feature Engineering (refined data)
- Visual exploration of modeling results
- "What-if" Analyses

Application to the ATM domain @ DART, datAcron

- Complex data (4D trajectories; weather, airspace design, ... contextual data)
- Complex algorithms in moving towards TBO
- Decision support (vs. autonomous "black box" decision making)





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Thank you very much for your attention!



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