

Vistribute: Distributing Interactive Visualizations in Dynamic Multi-Device Setups

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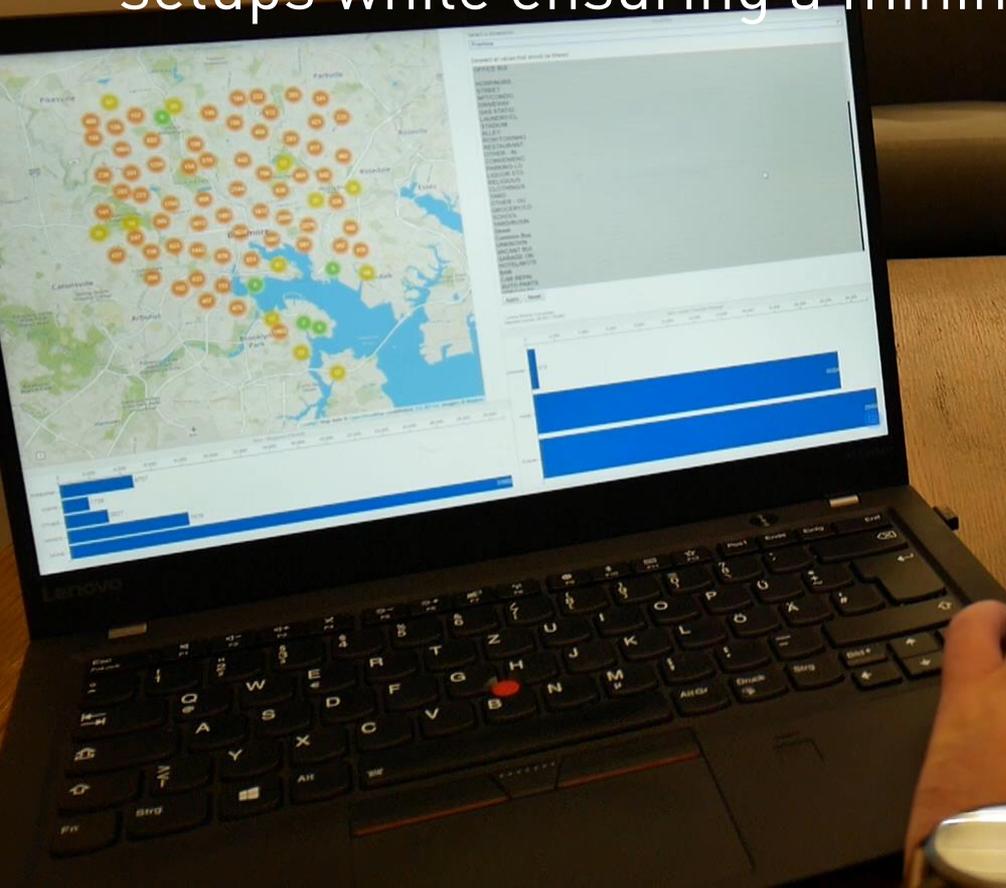
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Nowadays, data analysis can take place in many different environments with various devices



How can we maximize the advantages of multi-device setups while ensuring a minimal user effort?



What we know: devices can fulfill different roles during visual data analysis

Roles emerging from **data exploration patterns**,
e.g., overview+detail, focus+context

Roles emerging from **multi-user constellations**,
e.g., personal toolboxes, shared interaction space

So far:

- Only systems for specific device combinations
- Lacking support for flexibly placing visualizations
- Increasing configuration effort with many devices

Kister et al., CGF '17:
GraSp



Wozinak et al., NordiCHI '14:
Thaddeus



McGrath et al., AVI '12:
Branch-merge-explore



Horak et al., CHI '18:
When David meets Goliath



Plank et al., CHI '17:
Is Two Enough?!



Langner et al., VIS '18:
VisTiles



What we know: various frameworks for cross-device development exist, but rarely focus on visualizations

Synchronization frameworks:

Support for synchronizing elements or events across devices

Badam and Elmqvist 2014: *PolyChrome*

Badam et al. 2015: *Munin*

Houben & Marquardt 2015: *WATCHCONNECT*

Klokrose et al. 2015: *Webstrates*

Schreiner et al. 2015: *Connichiwa*

Distribution frameworks:

Automatic distribution of components based on manually defined constraints

Yang & Wigdor 2014: *Panelrama*

Nebeling & Dey 2016, Nebeling 2017: *XDBrowser*

Husmann et al. 2018: *Out of Office Software Development*

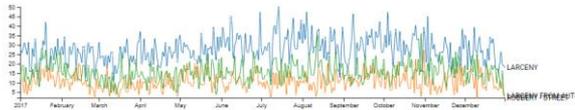
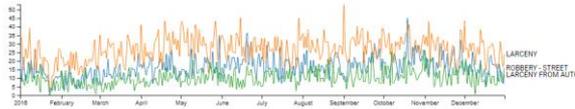
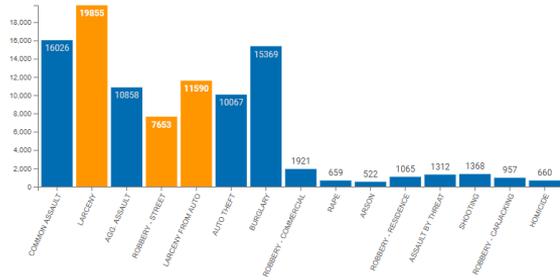
Park et al. 2018: *AdaM*

So far, all frameworks...

- rely on additional input from developers or users
- do rarely consider visualization-specific aspects

Visualizations are more than “just” views

Visualizations have a rich body of characteristics and certain relationships to other visualizations



Visualization Type
Data Points
Axis

Encoding
Visual Density
Data Source

Size
Internal State

Idea: Considering these aspects alongside device properties and user preferences

We contribute the Vistribute framework



Design Space

Exploring the properties and relationships between visualizations, devices, and the user



6 Heuristics

High-level constraints for deriving a view-sensitive distribution and layout



Vistribute System

Open source implementation representing one possible instance of our heuristics

Each heuristic contributes to different aspects of a distribution

Grouping & alignment based on view relationships

*1 Visual Similarity *2 Data Similarity *3 Input Connectivity

View adjustments and device assignments

*4 Data Density *5 Device Suitability

Allowing adaptations by users

*6 User Preferences

Grouping & alignment based on view relationships

*1 **Visual Similarity** promotes comparison

*If two views are **visually very similar**, they should be both **juxtaposed and aligned**.*

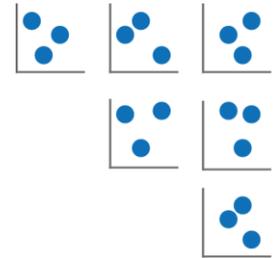
Example: Small multiples



*2 **Data Similarity** indicates alternative representations

*If two views have a **high degree of data similarity** and a corresponding visual similarity, they should be **placed close to each other**.*

Example: Scatterplot matrix



*3 **Input Connectivity** fosters the data exploration

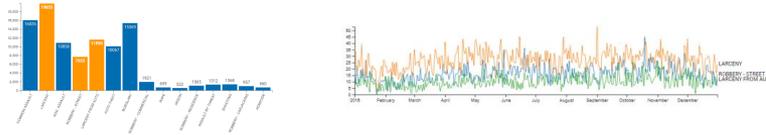
*If an interface component serves as **data input for others**, it should be **placed close to the affected** components.*

Example: Dashboard



View adjustments and device assignments

*4 Data Density influences the space requirement



A view should be **allocated space proportional** to the **number of data points** it encodes.

*5 Device Suitability differs for all visualizations

If devices are diverse, **view assignments** should be **guided by device suitability**.



Allowing adaptations by the user:

*6 User preferences always exists

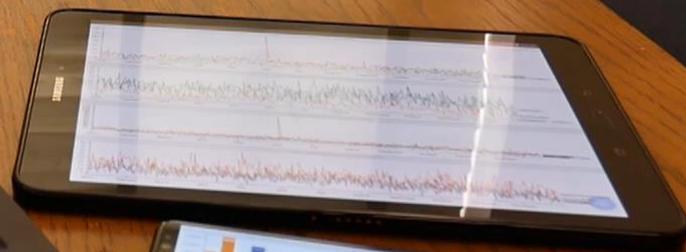
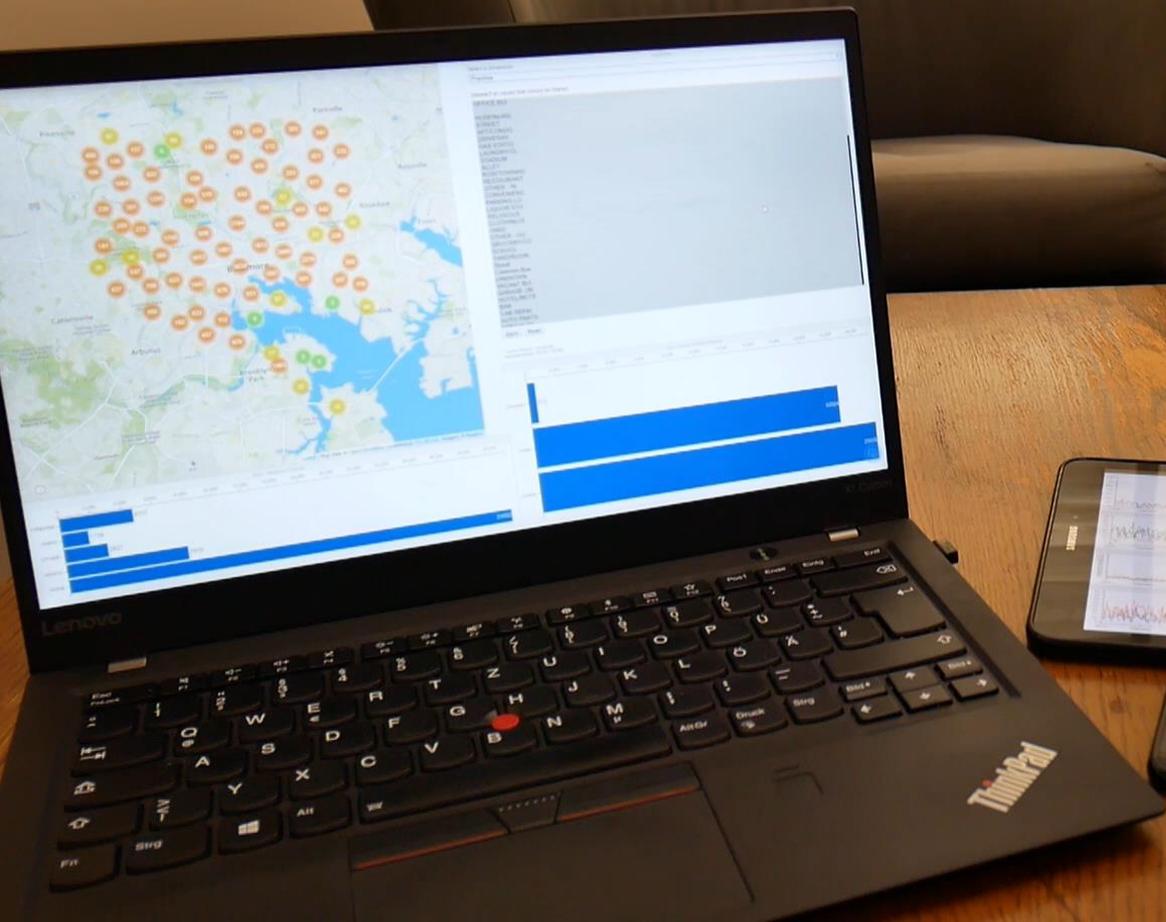
*If user **preferences** are applicable, they **outweigh all other heuristics**.*

Users can have **static** preference about specific distribution details

In the context of analysis tasks, **temporary** user interest can occur



Web-based prototype serving as an example implementation



User-created distributions versus Vistribute: a small-scale comparison study

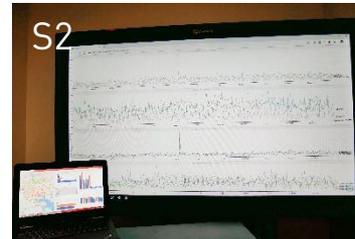
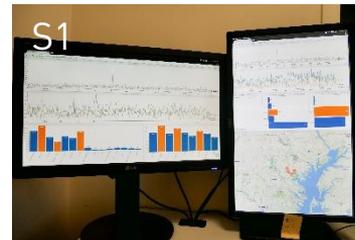
- 👤 6 participants (1 female, 5 male; active in the field > 3 years)
- 🕒 2 phases; approx. 60 minutes per session
- 💬 Think-aloud protocol

Phase 1:

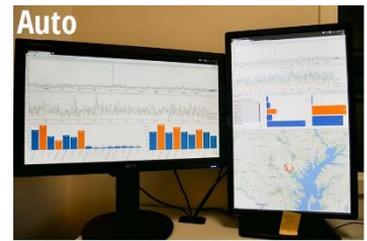
Manually distributing 10 visualizations in 3 different setups

Phase 2:

Per setup, rating of 3 existing distributions
(2 created by other participants, 1 by Vistribute)



In most cases, multiple reasonable distributions exist



Personal preferences have a strong influence



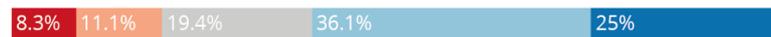
User considered similar aspects as our heuristics



Manual distributions rated slightly better

Manual

$M = 3.9$



Automatic

$M = 3.6$



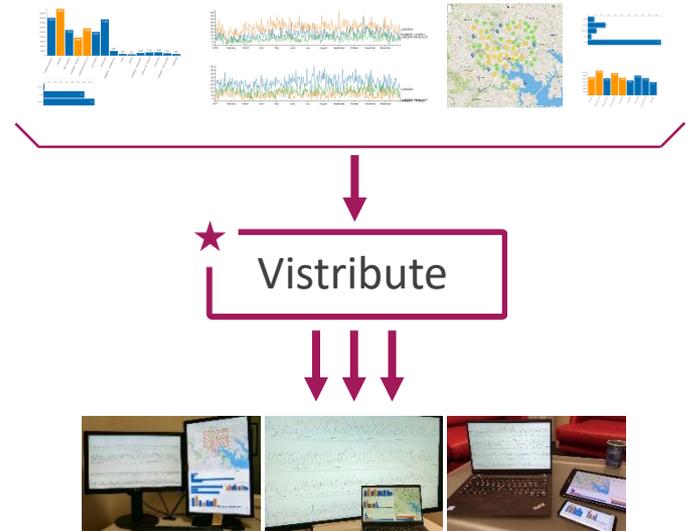
1 (Unsuitable Distribution) 2 3 4 5 (Optimal Distribution)



Towards effortless multi-device environments

- ☹️ Manually distributing is *“exhausting”*, *“there should be an optimization for this”*
- 🕒 On average, participants spent 8 minutes on one distribution

Vistribute provides reasonable distributions without requiring additional user input



Towards effortless multi-device environments



Next: Investigating how analysts work in MDEs

Refinement of heuristics and investigate cross-device interactions

From heuristics towards formalism

Incorporating AI mechanisms to further improve distributions

From distribution towards visualization generation

Generating suitable visualization for the user's current goals



Vistribute: Distributing Interactive Visualizations in Dynamic Multi-Device Setups

Open positions for
PhD students and Postdocs
> imld.de/jobs

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