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Saliency Prediction in the Data Visualization Design Process

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Advisor: Prof. Fulvio Corno

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Dan Witzner Hansen



**Politecnico
di Torino**



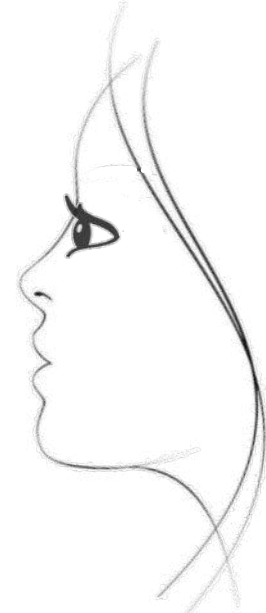
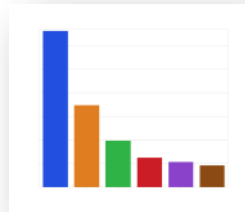
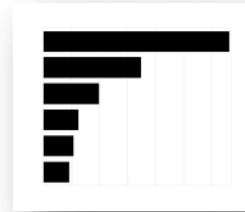
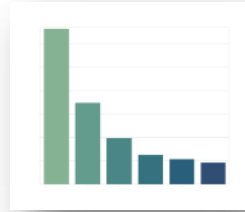


*"Design graphic representations of data by taking into account **human sensory capabilities** in such a way that important data elements and data patterns can be quickly perceived."* by Colin Ware [1]



Graph Designers

DataViz



*Human Sensory
Capabilities*

Observer

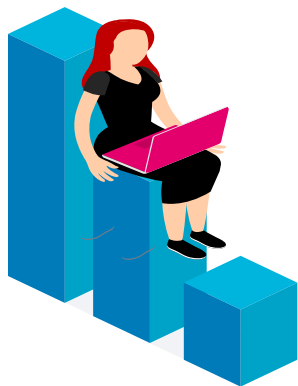


Human Vision Attention

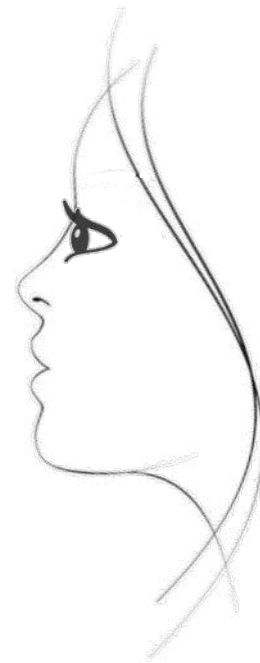
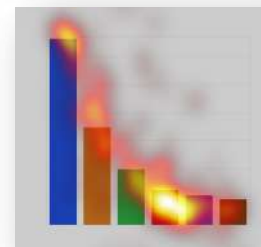
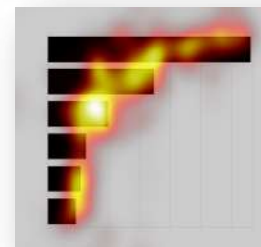
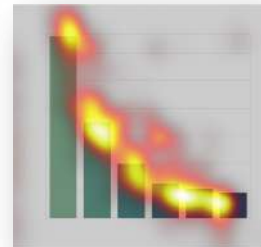
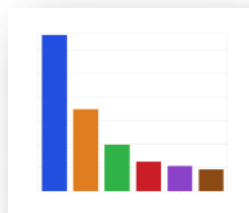
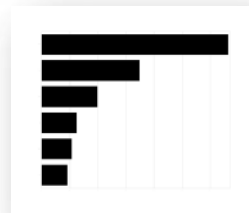
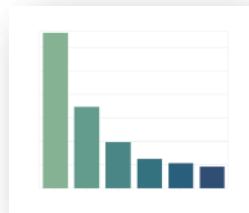


DataViz Images

Observers Focus Attention



Same data,
different visual
elements

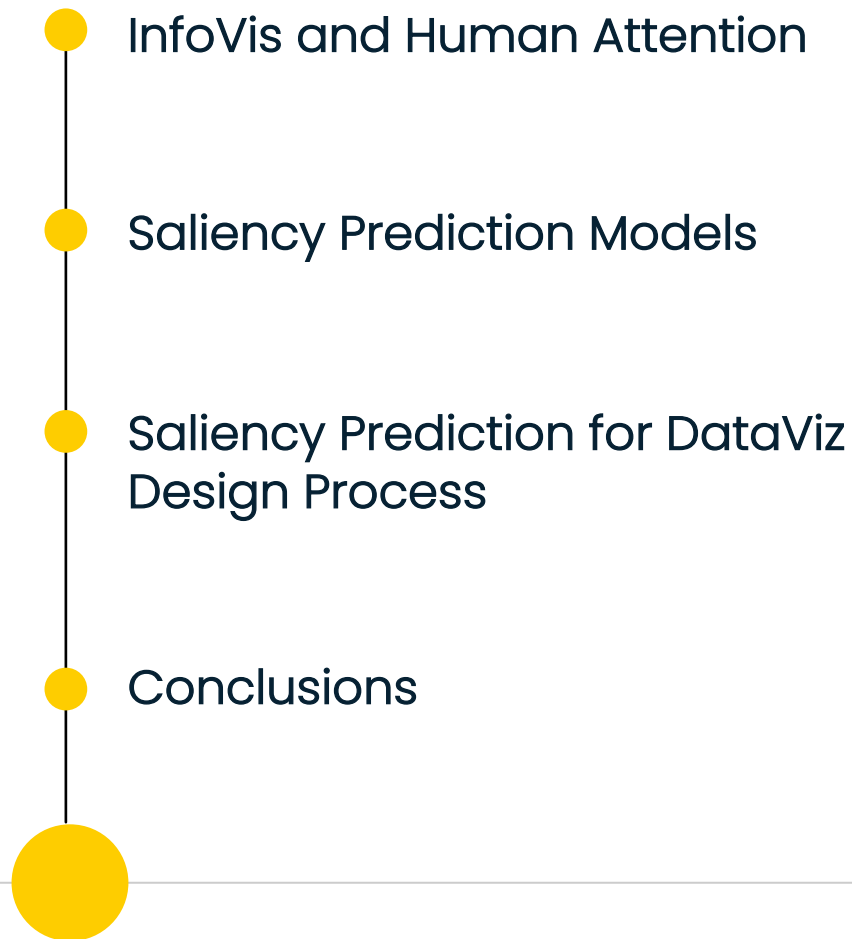




Research Aim

Integrate **visual-cognitive concepts** into the **Data Visualization Design Process and tools**. We intend to bring those concepts to the **graph designer's** context and provide insight into how their design choices might affect the **observer's perception**.

Research Development Process



Research Development Process

InfoVis and Human Attention

- Key Concepts
- Literature Review

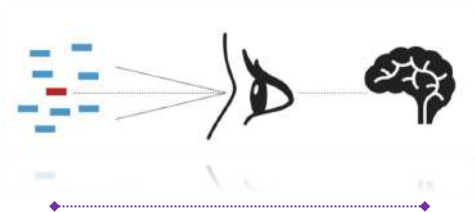
Saliency Prediction Models

Saliency Prediction for DataViz Design Process

Conclusions



Preattentive Process

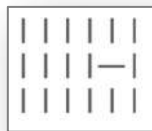


Decides what visual attributes are ***offered up*** to our attention and easy to find in the next fixation

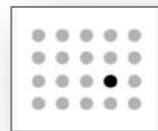
“ [21]



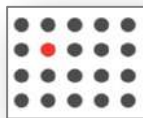
Preattentive Attributes



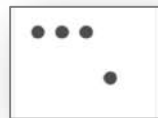
Orientation



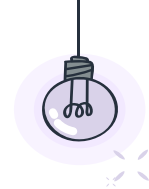
Color Intensity



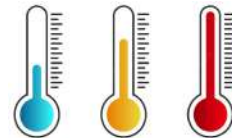
Color Hue



Position



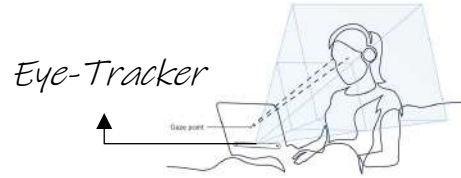
Saliency



The degree to which a target stimulus “***pops out***” in a set of stimuli.” [22]



Heat Map



Saliency Prediction



Computer Models that imitate human attention process

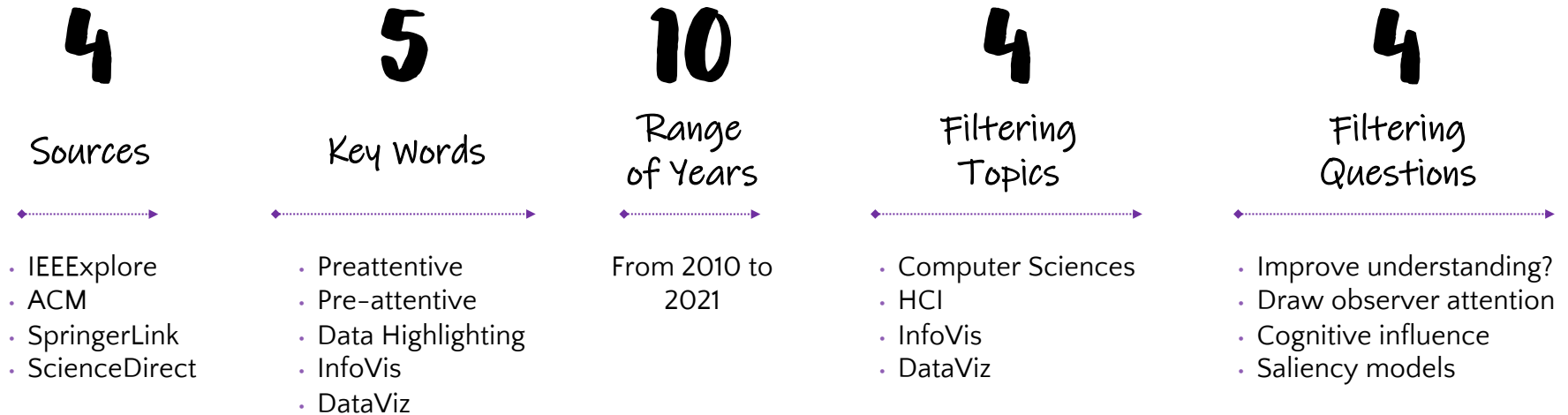


Saliency Map



How is the preattentive visual process used in information visualization? (RQ)

Oriented to improve the observer's cognitive process in graph comprehension





Literature Review II

Preattentive Uses in InfoVis:

1. For Design

- Design elements manipulation to achieve different goals (*e.g., identifying relevant data*).

2. For Measurement

- Assess attention on portions of a graph or determine the impact of preattentive attributes.

Insights

- Highlighting and Data enhancement
- Implicit and Unconscious design decisions
- Graph designer oblivion
- Saliency Models

Research Development Process



InfoVis and Human Attention



Saliency Prediction Models

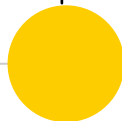
- Saliency Models Fundamentals
- Saliency Models + DataViz
- Saliency Model Validation



Saliency Prediction for DataViz
Design Process

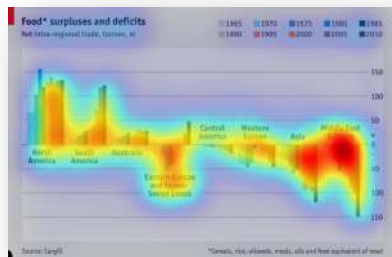


Conclusions

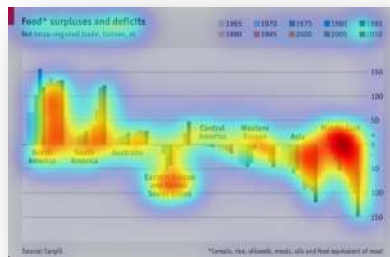




Saliency Models in DataViz

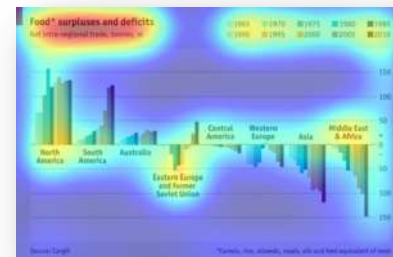


Itti-Koch
(2001)



Matzen
(2018)

Classical Models



Bylinskii
(2017)

Deep Learning

Itti-Koch

2001

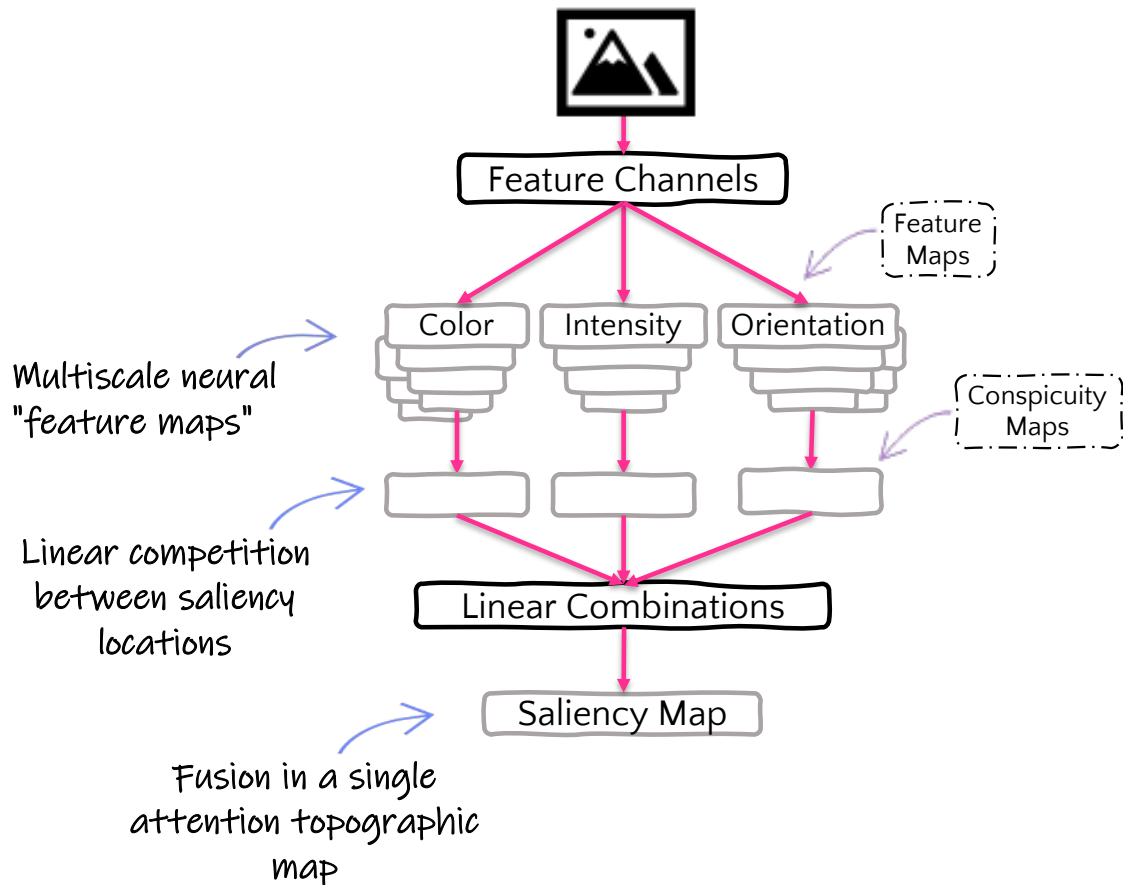
Biologically
conceivable

Bottom-up
Classical Model

Natural Images
Root Data Set

Markov Chains
Strategy

Matlab
Code Language



Matzen

2018

Biologically
conceivable

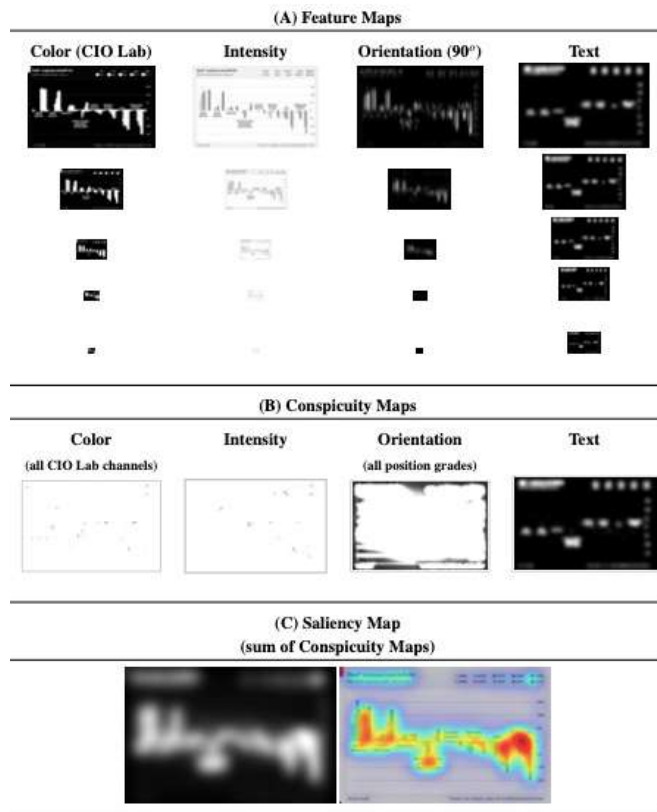
Bottom-up + Top-down
Classical Model

MASSVIS + Natural Images
Root Data Set

Modified Itti-Koch + Text
Detection
Strategy

Matlab
Code Language

Saliency Prediction Process



2017

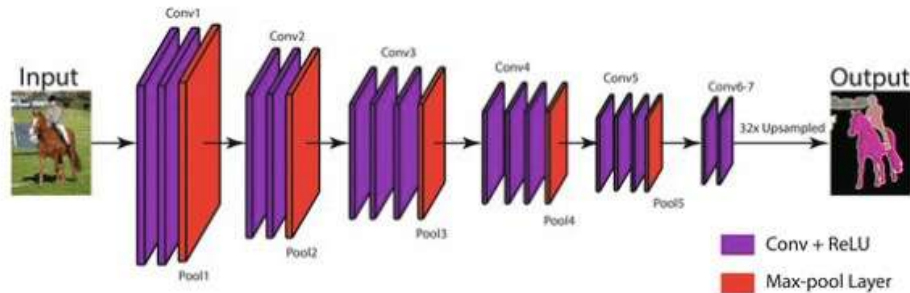
Deep Learning

Model Classification

Fully Convolutional Networks

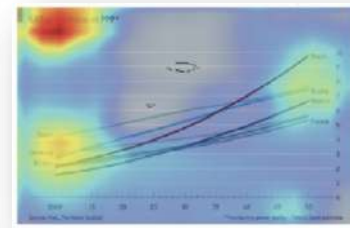
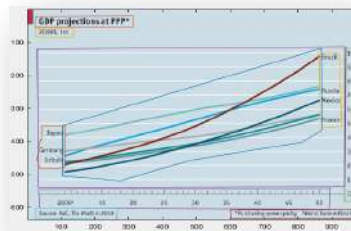
Strategy

Python + Caffe DL



Fully Convolutional Networks (FCN32)

Relative Importance
(Higher-level Features)





Saliency Models Analysis

Concerns about the Saliency Models' behavior:

1. Significant imbalance in the attention given to Text elements.
2. Suitability of the dataset for training DataViz models (e.g., several context elements).
3. Some representative features, such as position, are not considered in DataViz saliency models.

Which is the best-performing model?

Even if the graph's visual elements are varied (or removed).

Research Development Process



InfoVis and Human Attention



Saliency Prediction Models

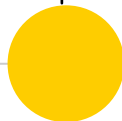
- Saliency Models Fundamentals
- Saliency Models + DataViz
- **Saliency Model Validation**



Saliency Prediction for DataViz
Design Process



Conclusions





First Experiment

Scopes

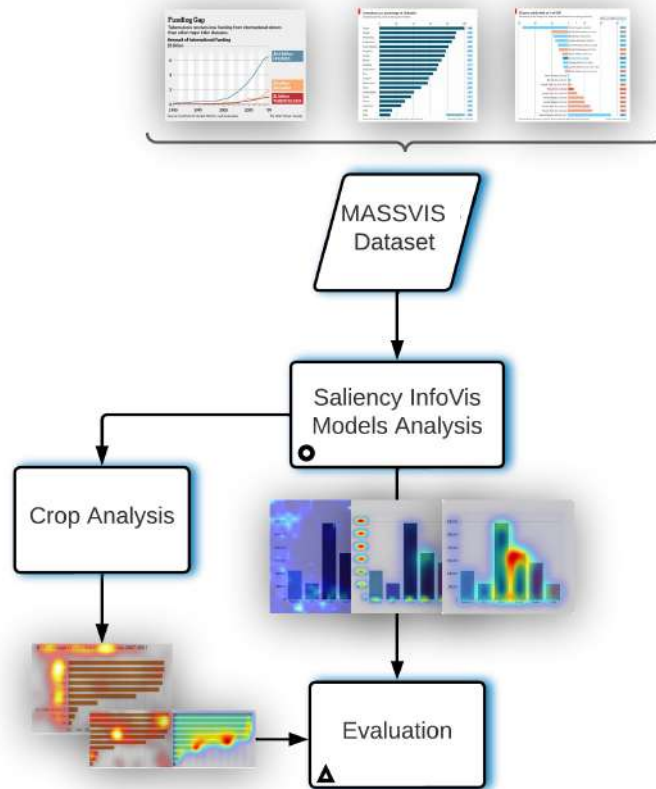
- Confirming the model's performance with the frequently used dataset.
- Detailing how the models behaved by predicting the graph saliency.

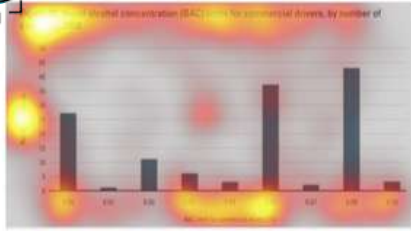
Dataset

- MASSVIS: Massachusetts (Massive) Visualization Dataset

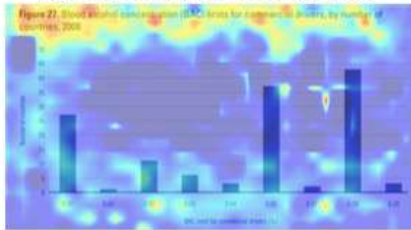
Metrics typically used in the Saliency Prediction:

- AUC (Area Under the ROC –Receive Operating Characteristics–)
- CC (Pearson's Correlation Coefficient)
- NSS (Normalized Scanpath Saliency)

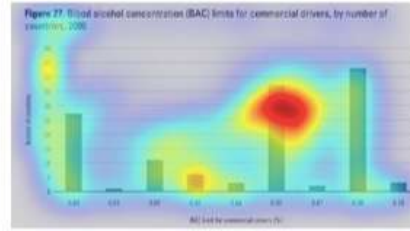




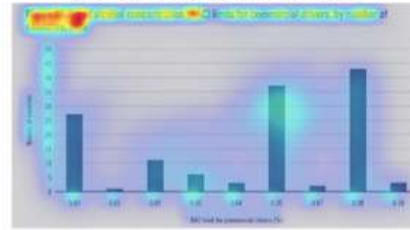
(a) Heatmap (eye-tracking data)



(c) Bylinskii Saliency Map



(b) Itti-Koch Saliency Map



(d) Matzen Saliency Map

Insights:

- The results were **consistent** with previous studies.
- Matzen model had the **highest** score of the three metrics.
- Most of the attention goes to the **textual elements**, and Matzen and Bylinskii models accurately address this.



Second Experiment

Scope

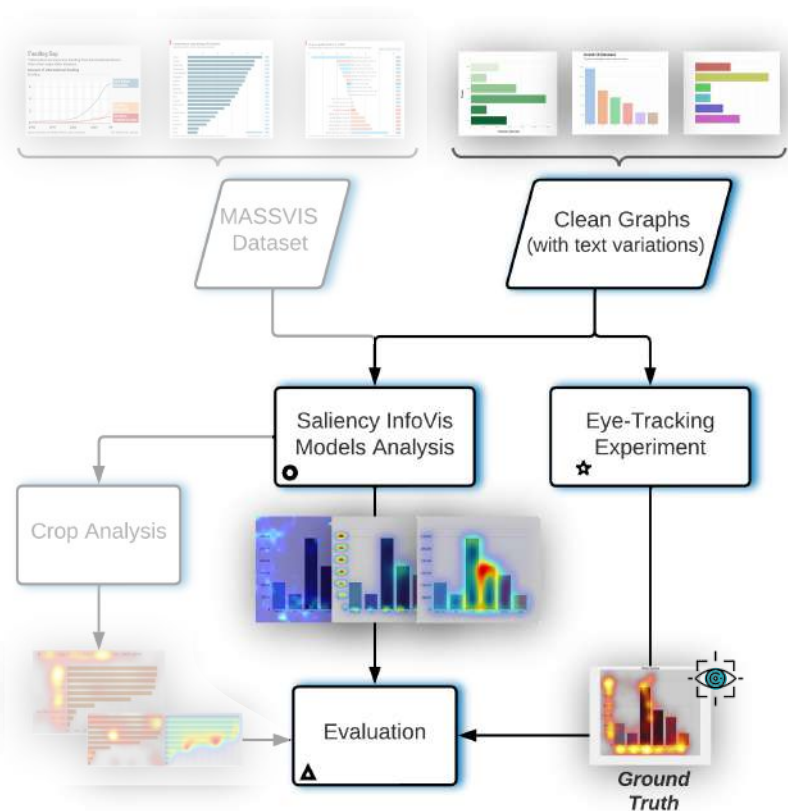
- Validate the model's performance with graphs **without design elements distractions** (e.g., logos, icons) and **varying** some design elements.

Images

- Clean Graphs dataset

Metrics for Saliency

- AUC , CC and NSS



Clean Graphs Dataset

30 images

Bar Charts

DataViz Technique

Covid-19

Dataset

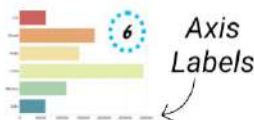
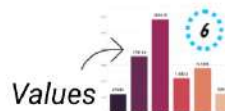
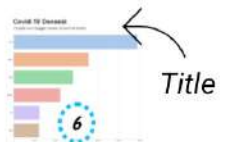
Position, Color, Orientation
and Textual Elements.

Variations

Python

Seaborn library

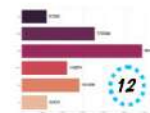
Textual Elements



Position



Color Palettes



Orientation



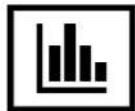


Eye-Tracking Data Acquisition



Clean Graphs Dataset

General Data



30

Images



62

Observers



14 - 50

Age rate



Exploratory

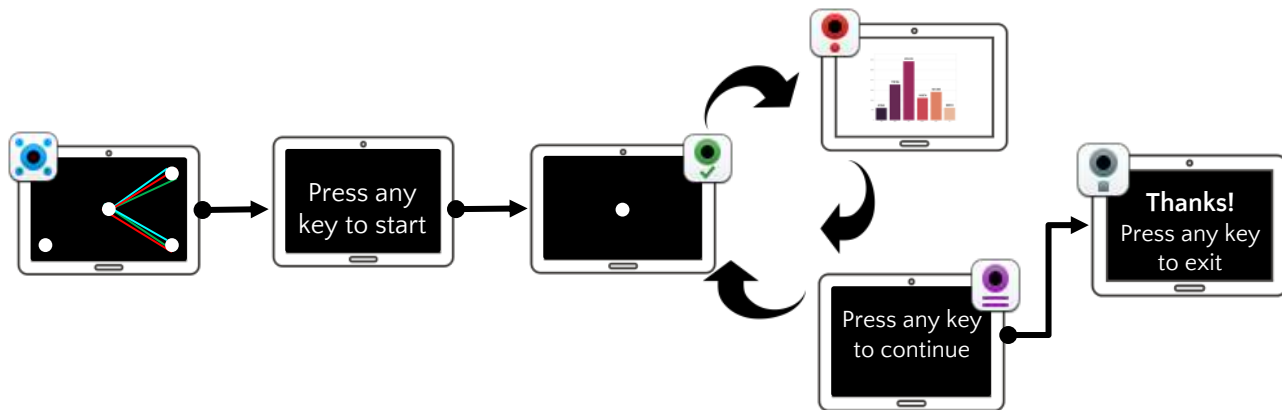
Visual observer task



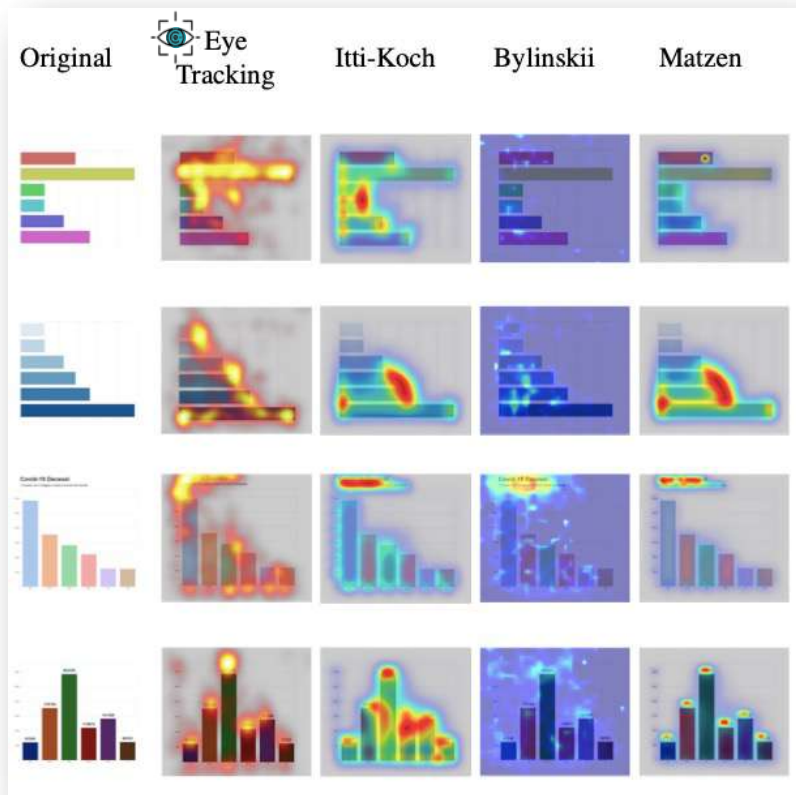
5 sec.

Data recording

Data Collection Process



Example Saliency Maps



Insights

- Text is a visual component that is **highly influential**, for better or worse, in all three model's performance.
- Orientation, position and Color **influence the model's** accuracy, but less than text elements.
- Blank graph (no text), was where the models had the **lowest scores**.



Third Experiment

Scope

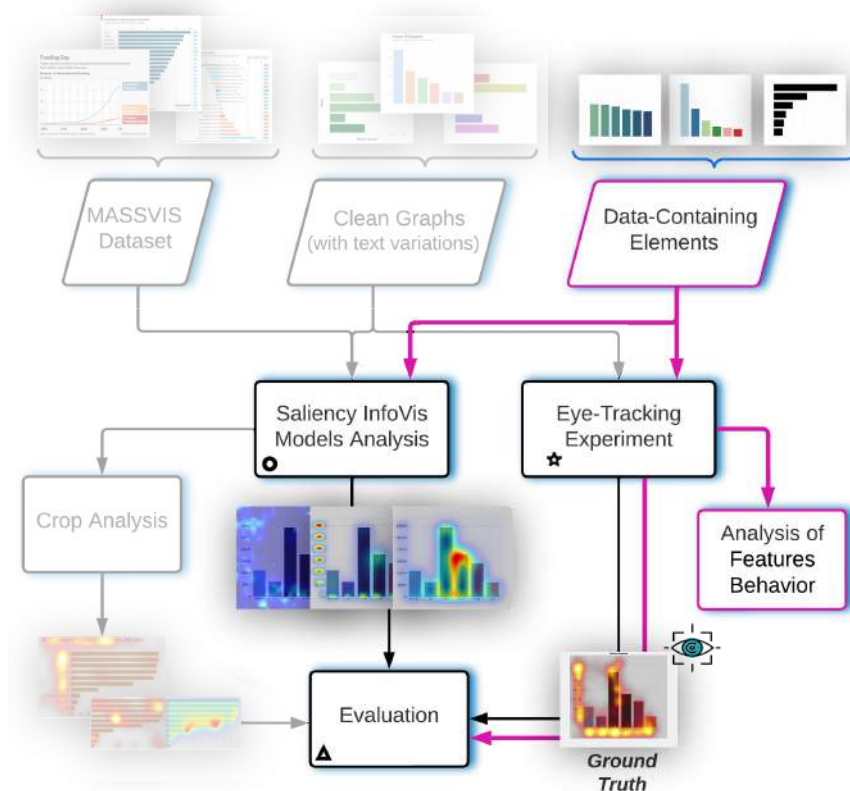
- Deepen the behavior of the models only in the elements that represent the data (no textual elements).

Images

- Clean Graph (without text)

Metrics for Saliency

- AUC, CC and NSS





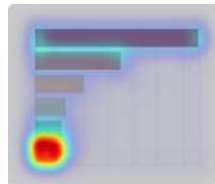
Third Experiment Results



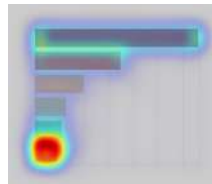
Orientation



Itti-Koch



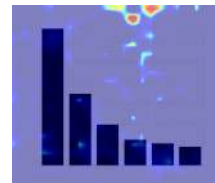
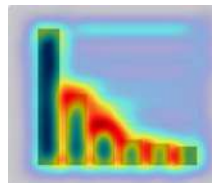
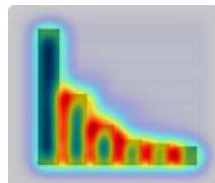
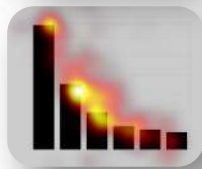
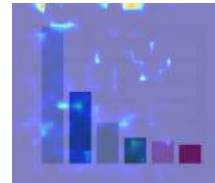
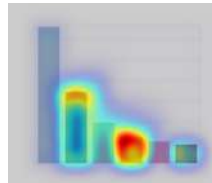
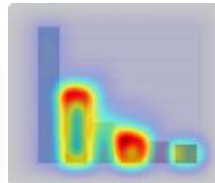
Matzen



Bylinskii



Color Palette








Models Comparison

| Model | Experiments | CC | | NSS | | AUC | |
|-----------|--------------------------------|-------|----------|-------|----------|-------|----------|
| | | t | p | t | p | t | p |
| Bylinskii | MASSVIS vs. Clean Graph | 7.50 | 1.06e-09 | 5.26 | 2.30e-06 | 6.13 | 9.81e-08 |
| | Clean Graph vs. Data-Contained | 6.28 | 9.31e-08 | 5.94 | 7.44e-07 | 5.60 | 3.08e-06 |
| Matzen | MASSVIS vs. Clean Graph | 0.66 | 0.51 | 0.49 | 0.63 | 2.74 | 0.008 |
| | Clean Graph vs. Data-Contained | 0.097 | 0.92 | 0.74 | 0.46 | -1.24 | 0.23 |
| Itti-Koch | MASSVIS vs. Clean Graph | -1.90 | 0.063 | -2.40 | 0.020 | -2.70 | 0.01 |
| | Clean Graph vs. Data-Contained | -3.12 | 0.003 | -2.03 | 0.052 | -1.05 | 0.30 |

t-test

- Bylinskii is strongly attached to the training images. 
- Itti-Koch, the less context information (distractors), the better the saliency prediction. 
- Matzen is the model that demonstrated the most stable behavior during the three experiments. 

Research Development Process



InfoVis and Human Attention



Saliency Prediction Models

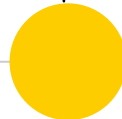


**Saliency Prediction for DataViz
Design Process**

- Saliency as Design Tool
- Saliency as Measurement Tool
- Experts Validation



Conclusions





Human Sensory Capabilities in Graph Design

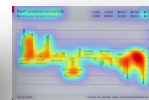
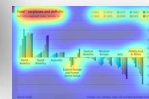
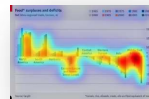


For Design



For Measure

Literature Review



Saliency Models for DataViz



Matzen Model

How do we integrate **saliency models** into
the **DataViz design process** to support
Graph Designer decision-making?



Development Tools



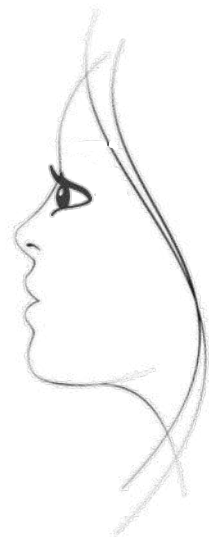
Graph Designers

Design Tool

Systematic variation of Visual Elements to highlight relevant data.

Measurement Tool

Saliency Maps into graph design commonly process to validate design decisions.



Observer



Design Tool



1. Inputs:



Dataset



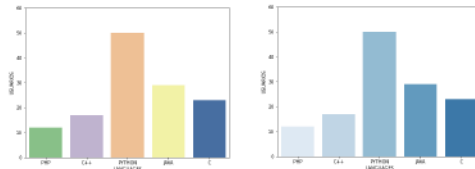
Focus Data



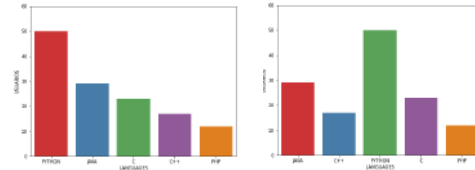
Preferred
Color Palettes

2. Systematic Variations:

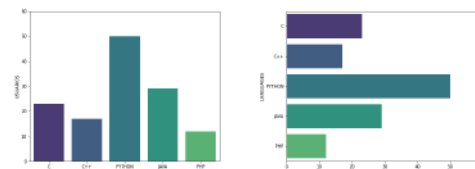
Color Palette



Data Position

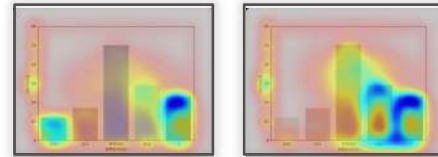


Data Orientation

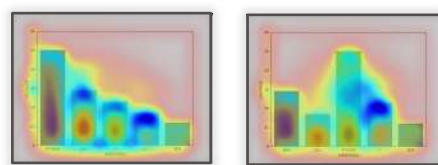


3. Saliency Map Generation:

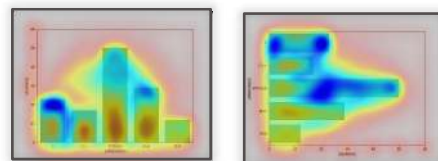
Color Palette



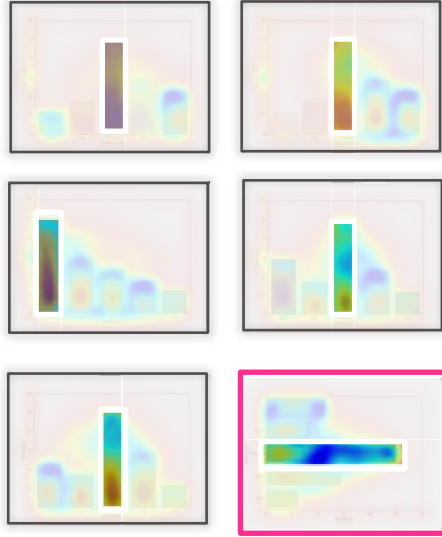
Data Position



Data Orientation

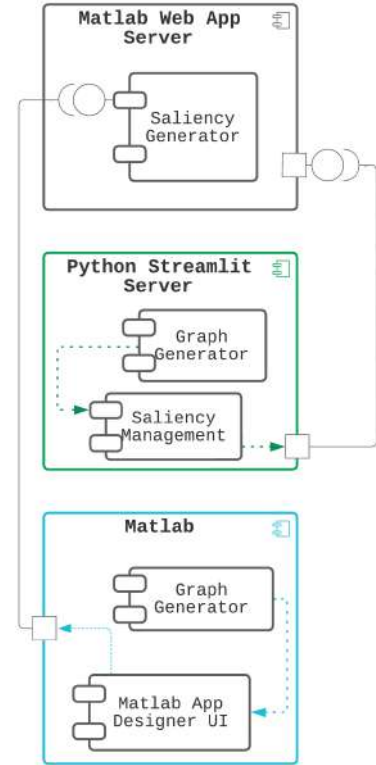
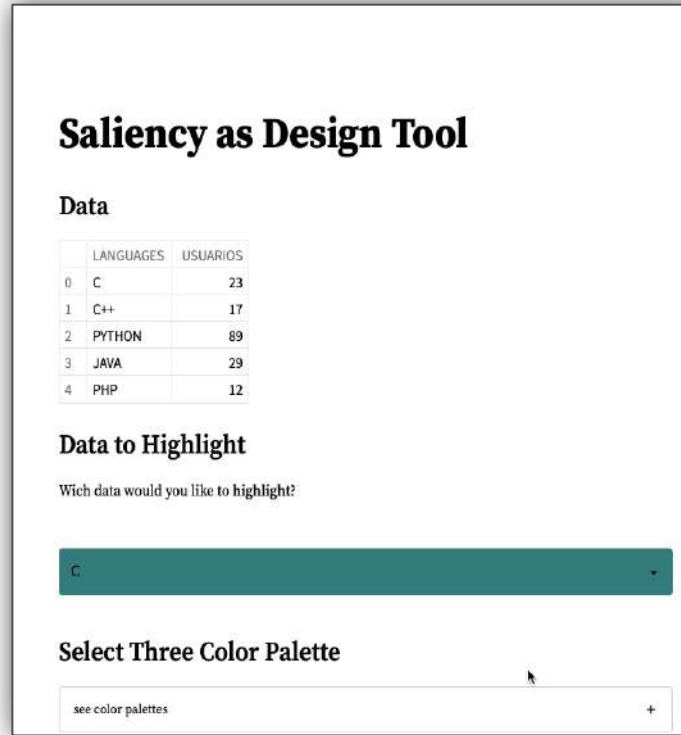


3. Graph Selection



Winner

4. Show Results



Components Diagram



Development Tools



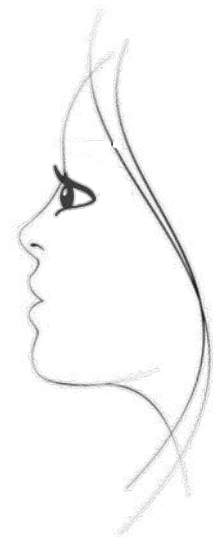
Graph Designers

Design Tool

Systematic variation of Visual Elements to highlight relevant data.

Measurement Tool

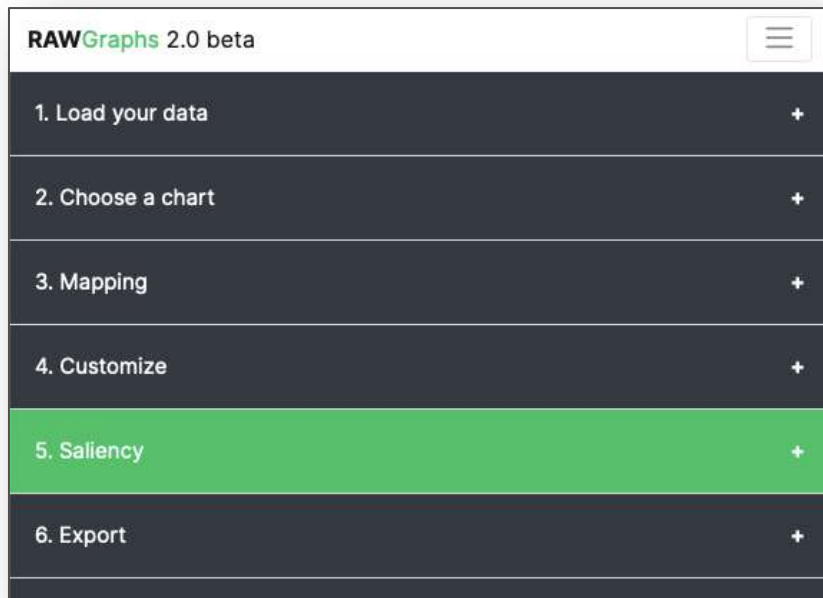
Saliency Maps into graph design common process to validate design decisions.



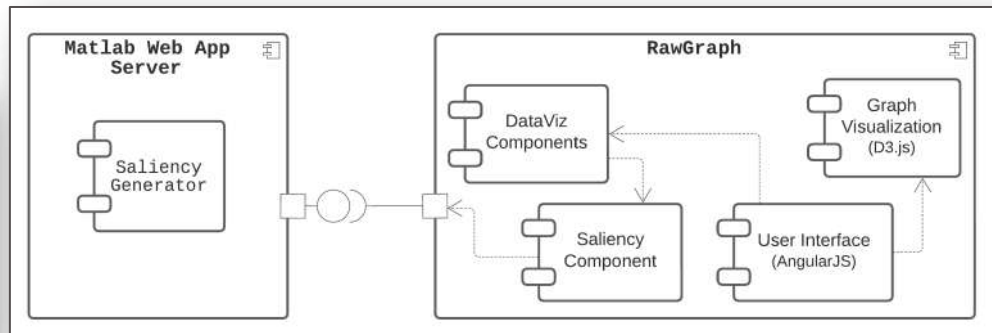
Observer



Measurement Tool



Components Diagram



New Function



Measurement Tool

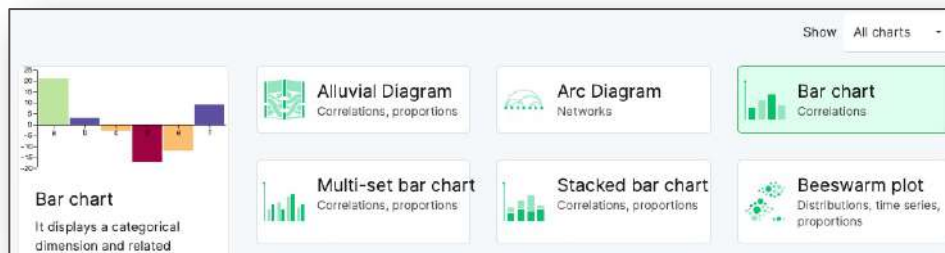


1. Input



Dataset

2. Graph Selection



3. Mapping





Measurement Tool

4. Customize

ARTBOARD

Width (px) 805

Height (px) 600

Background ☐ #FFFFFF

Margin (top) 20

Margin (right) 10

Margin (bottom) 20

Margin (left) 50

Show legend ☐ No

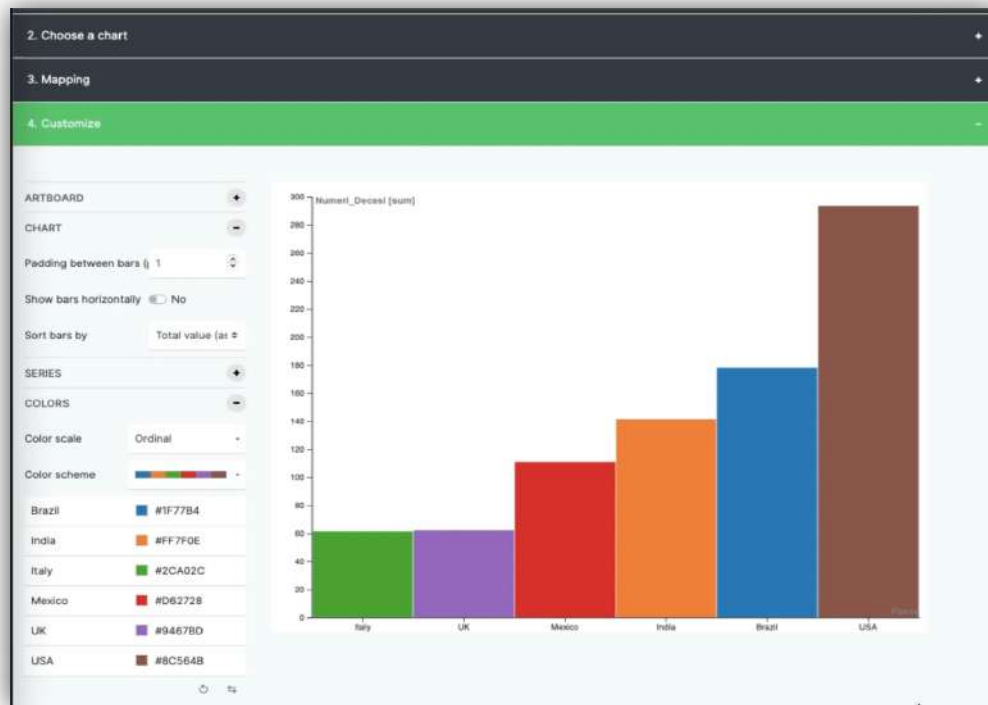
Legend width 200

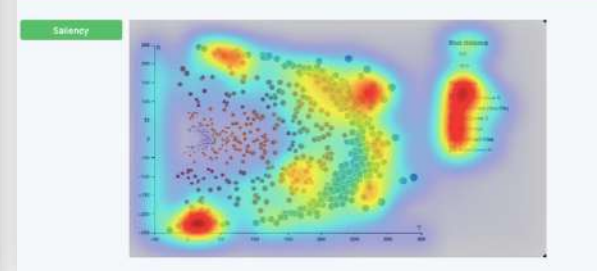
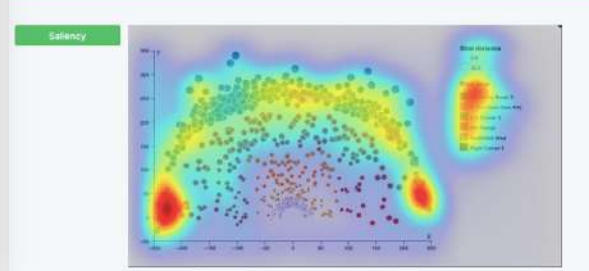
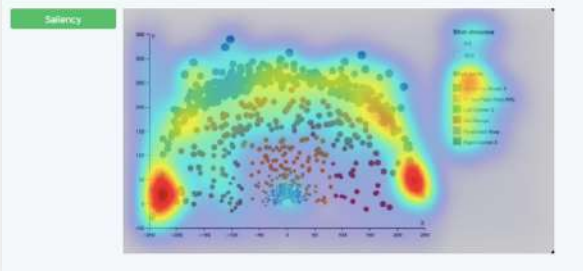
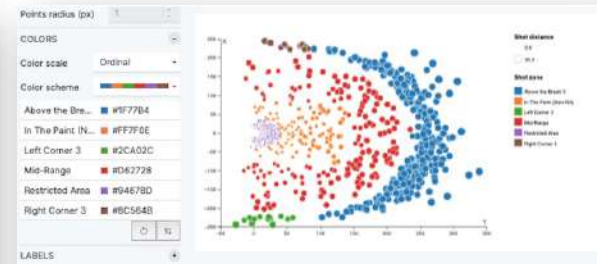
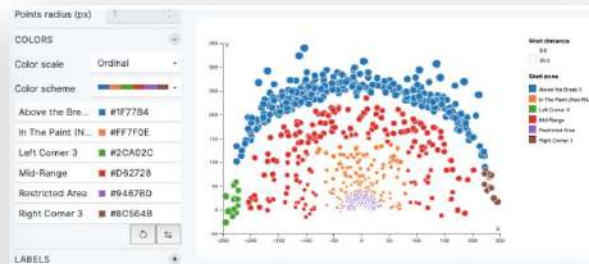
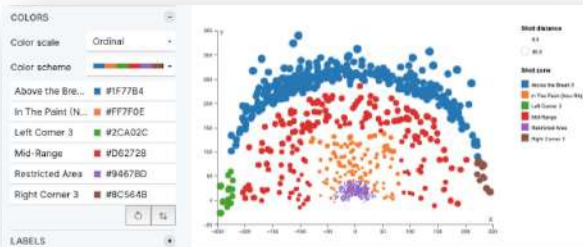
CHART

SERIES

COLORS

5. Saliency Map Generation



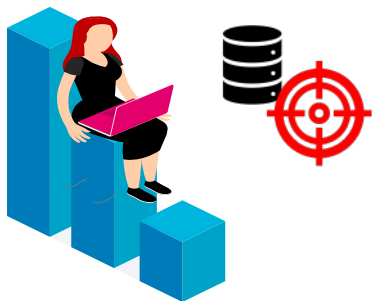


Examples



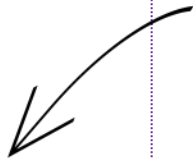


Development Tools



Graph Designers

Experts
Validation

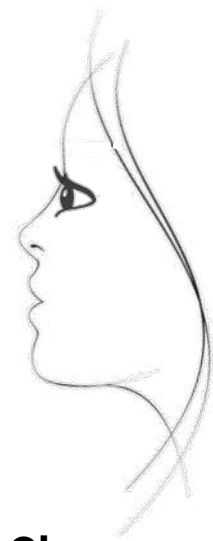


Design Tool

Systematic variation of Visual Elements to highlight relevant data.

Measurement Tool

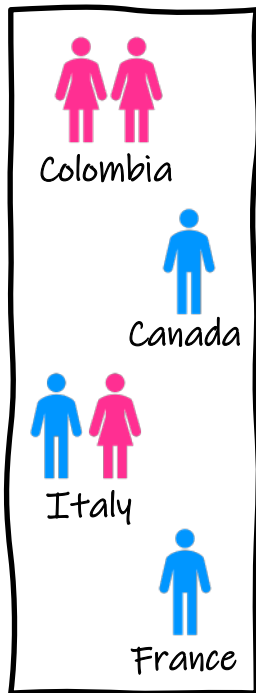
Saliency Maps into graph design commonly process to validate design decisions.



Observer

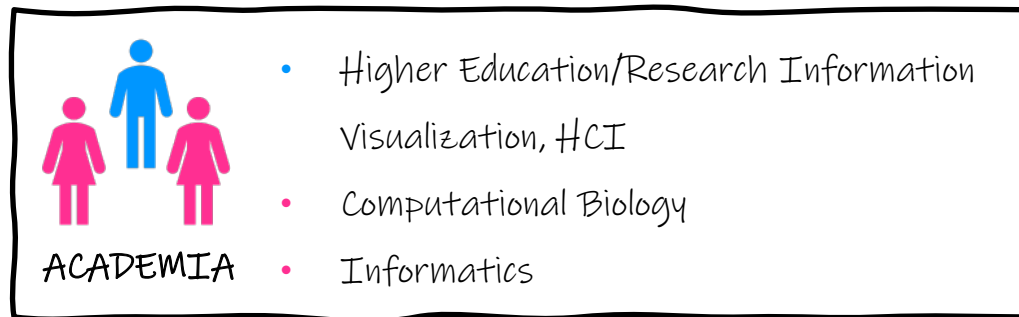


Experts Profiles



6
Experts

30 - 45
Age Range





Evaluation Protocol

Design Tool

1. Open Questions (relevant data highlighting)
2. Tool Demo (both versions)
3. Evaluation using Heuristics instruments:
 - **USE** (Usefulness, Satisfaction, and Ease of use)
 - **QUIS** (Questionnaire for User Interface Satisfaction)

Measurement Tool

1. Open Questions (area attention detection)
2. Tool Demo
3. Evaluation using Heuristics instrument:
 - **TAM** (Perceived Usefulness and Ease of Use)

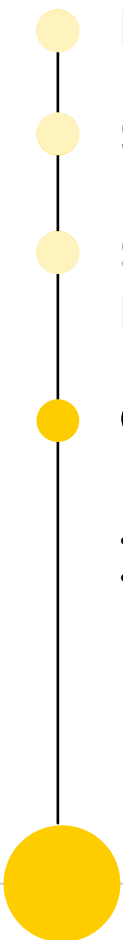


Experts Evaluation - Discussion

- "Tools with much **potential** and would be very useful"
- Attention behavior (patterns) could be **learned** if the tools are used for an extended period.
- The second benefit is expert-oriented since there may be an acquired data **visualization bias** over time
- Measurement Tool should provide **more information** about saliency
- Extend Design Tool with **other visualization** types.

*Results demonstrated that both tools are a **relevant** and **valuable** approaches.*

Research Development Process

- 
- InfoVis and Human Attention
 - Saliency Prediction Models
 - Saliency Prediction for DataViz Design Process
 - Conclusions**
 - Contributions
 - Future Works



Contributions



Saliency Prediction Models

Matzen

Design and Measurement

Literature Review

Integrate visual-cognitive concepts into the Data Visualization Design Process and tools. We intend to bring those concepts to the graph designer's context and provide insights into how her design choices might affect the observer's perception.

Measurement Tool.

Design Tool

Consciousness about visual impact.

Oblivion in research

"Tools with much potential and would be very useful"



Future Works



More
observation
Tasks

1. Improve saliency Prediction Models

Local Saliency for
data-contained
elements

2. Extend graph types

Graph
profile

3. Saliency Details for graph designer

Saliency
percentage
per by zones



Thanks!



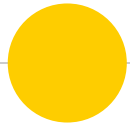
Luisa Barrera-León
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Fulvio Corno
Advisor

fulvio.corno@polito.it





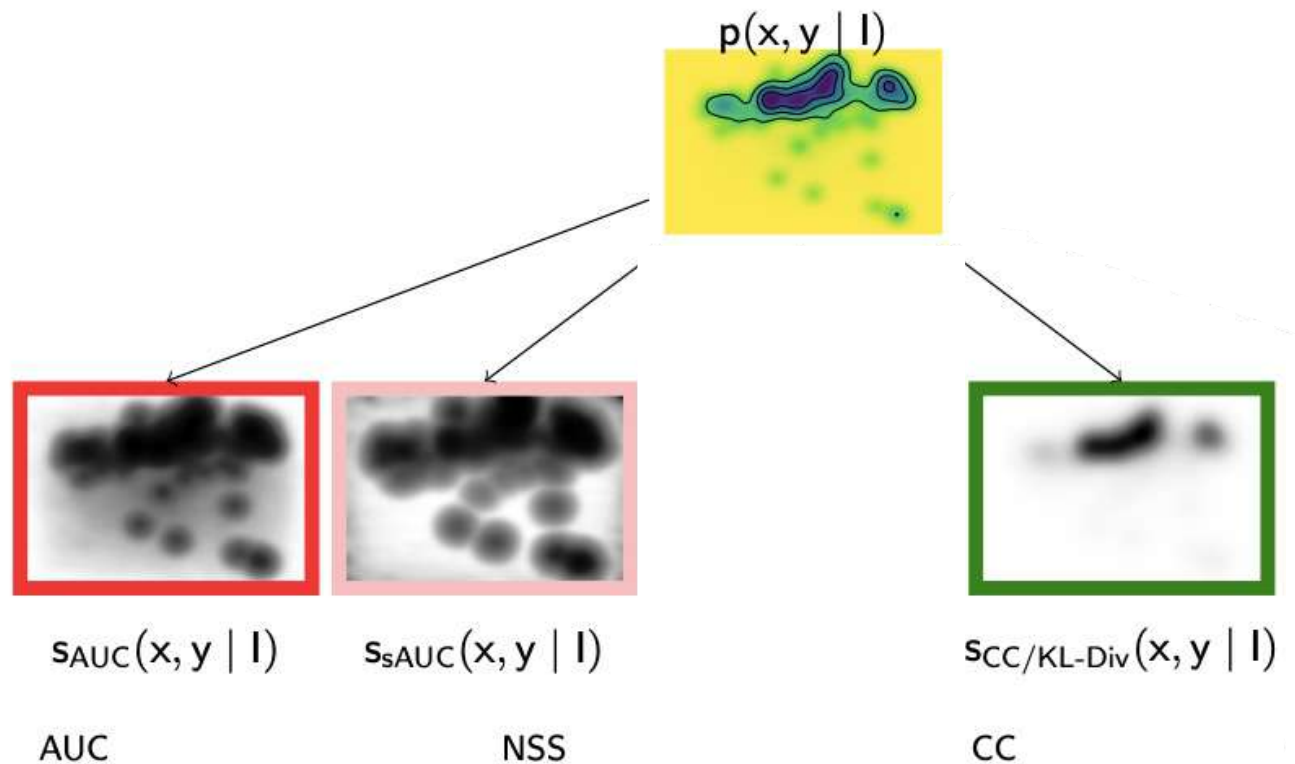
Supplemental Material

Feature Integration Model

Evaluation Metrics



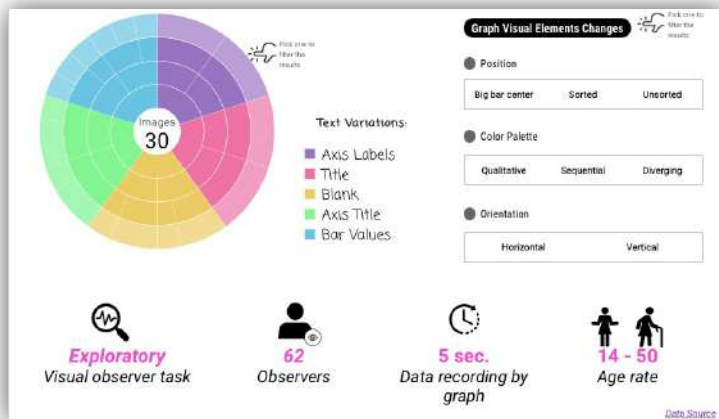
Metrics



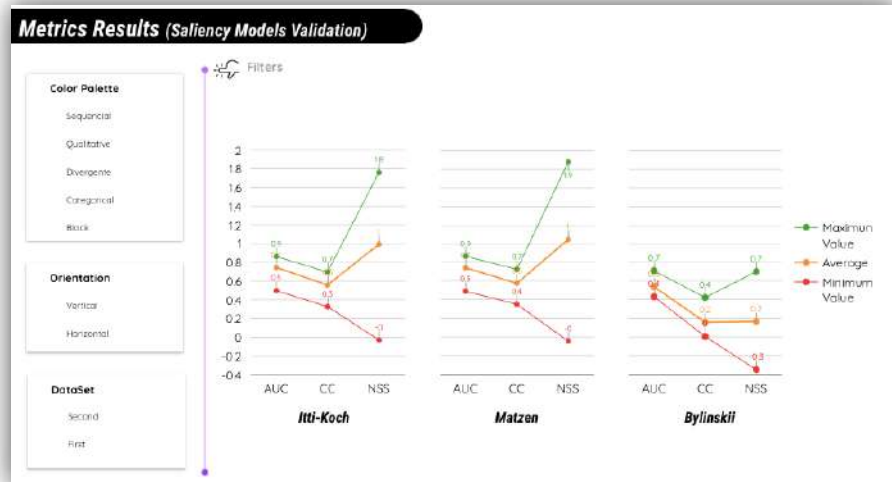


Dashboards

<https://bit.ly/2sdExpe> Dashboard

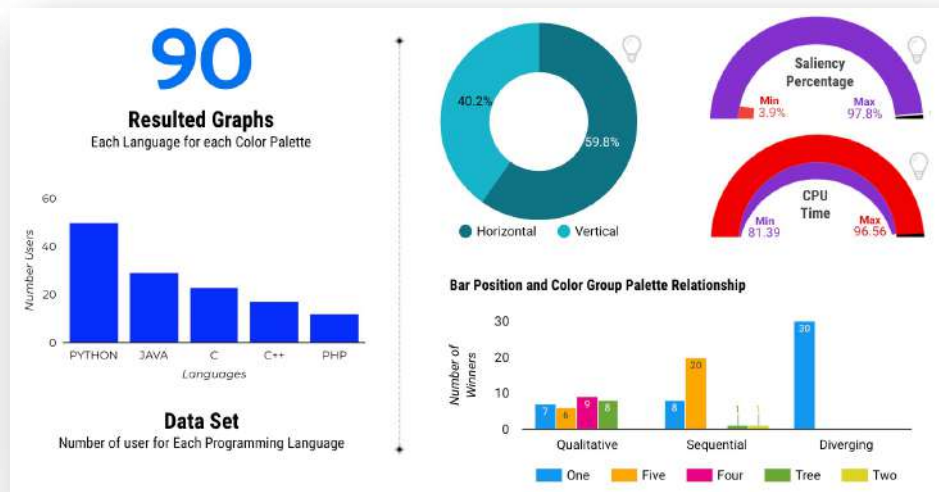


https://bit.ly/data-elements_results





https://bit.ly/DesignTool_CovidResutls



https://bit.ly/DesignTool_LanguResutls



Usability Heuristics

Design Tool

3. Evaluation using Heuristics instruments:

- **USE** (Usefulness, Satisfaction, and Ease of use): measures the subjective usability of a product or service. Questions: Perceived Usefulness, Perceived Easy to Use, and Perceived Satisfaction.
- **QUIS** (Questionnaire for User Interface Satisfaction): focuses on how an interface is evaluated, and it was created based on Shneiderman's list of five different types of dependent measures. Questions: Overall Reaction to the Software.

Measurement Tool

3. Evaluation using Heuristics instrument:

- **TAM** (Perceived Usefulness and Ease of Use): the model intends to forecast the future use of a product (expected usefulness and ease-of-use as viewed before any use) rather than to rate the actual user experience.



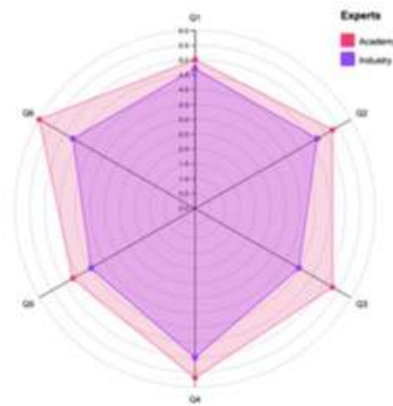
Experts Validation - Results

Table 8.2 QUIS Academia Experts

| Scale | Average Rating | Median Score |
|--------------------------------|----------------|--------------|
| Terrible 0 ... 9 Wonderful | 8,3 | 8,0 |
| Difficult 0 ... 9 Easy | 7,3 | 7,0 |
| Frustrating 0 ... 9 Satisfying | 8,3 | 8,0 |
| Inadequate 0 ... 9 Adequate | 7 | 8,0 |
| Dull 0 ... 9 Stimulating | 7,7 | 9,0 |
| Rigid 0 ... 9 Flexible | 5,3 | 8,0 |

Table 8.3 QUIS Industrial Experts results for Design Tool validation

| Scale | Average Rating | Median Score |
|--------------------------------|----------------|--------------|
| Terrible 0 ... 9 Wonderful | 7,3 | 8 |
| Difficult 0 ... 9 Easy | 6 | 7 |
| Frustrating 0 ... 9 Satisfying | 6,3 | 8 |
| Inadequate 0 ... 9 Adequate | 6,3 | 8 |
| Dull 0 ... 9 Stimulating | 7 | 9 |
| Rigid 0 ... 9 Flexible | 7,3 | 8 |



(a) TAM Perceived Usefulness (PU)



(b) TAM Perceived Ease of Use (PEU)