

# Five Design Sheet – Design study of your data: Analysis of the ATP Match Statistics.

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## Introduction

The dataset used for this visualisation design study originates from Jeff Sackmann's *tennis\_atp* repository, one of the most complete open-source collections of professional men's tennis data. It contains detailed historical information on player biographical attributes, match results, ranking histories, and match-level performance statistics. The dataset spans over five decades of the Open Era, with weekly ATP rankings mostly complete from 1985 onward and match statistics consistently available for tour-level events from 1991.

This assignment builds upon the data analysis performed in Assessment 1 and now focuses on designing appropriate visualisation concepts to communicate meaningful patterns within the dataset. Using the Five Design Sheet (FdS) methodology, several alternative visualisation ideas were developed to explore different aspects of the data, including global participation, player-level performance profiles, and the relationship between match statistics and outcomes. The aim of this design study is to select one suitable visualisation concept that effectively communicates an engaging and insightful tennis data story, and to document the design reasoning, comparisons, and evaluations that lead to the final choice.

## Brief Outline of the data

The dataset comprises multiple interconnected CSV files providing structured information about ATP players, tournaments, rankings, and match statistics. Key components include:

### Player Dataset

This file contains biographical information for each player, including first and last name, player ID, handedness (right/left), birth date, nationality (country code), and height in centimetres. These attributes support analysis

related to demographics, player development, and cross-country comparisons.

### Ranking Dataset

Weekly ATP ranking records include ranking date, ranking position, player ID, and ranking points (when available). Rankings are consistently recorded from 1985 onward, enabling longitudinal analysis of player progression, dominance, and career trajectories across multiple decades.

### Match Results & Statistics

Match datasets (e.g., *atp\_matches\_2023.csv*) include tournament details, surface type, round, player IDs, scores, match winners, and match-level statistical variables such as aces, double faults, first serves in, break points saved, and total points won. These statistics are recorded most reliably for top-level ATP events from 1991 onward, with expanded coverage for Challengers and qualifiers in later years. Many match rows also store redundant biographical and ranking information for both players to simplify analysis.

### Dataset Purpose

The dataset supports a wide range of analytical tasks, from evaluating player performance factors, studying match outcomes, and analysing ranking dynamics, to exploring country-level dominance and comparing statistical strengths between players. Because the data includes both numerical and categorical variables across many years, it is highly suitable for visual storytelling, statistical graphics, and interactive dashboards.

## Alternative Designs:

Three alternative visualisation concepts were developed using the FdS methodology. Each design focuses on a different perspective of the dataset and offers unique storytelling possibilities.

### 1. World Map Choropleth – Country Dominance (Sheet 2)

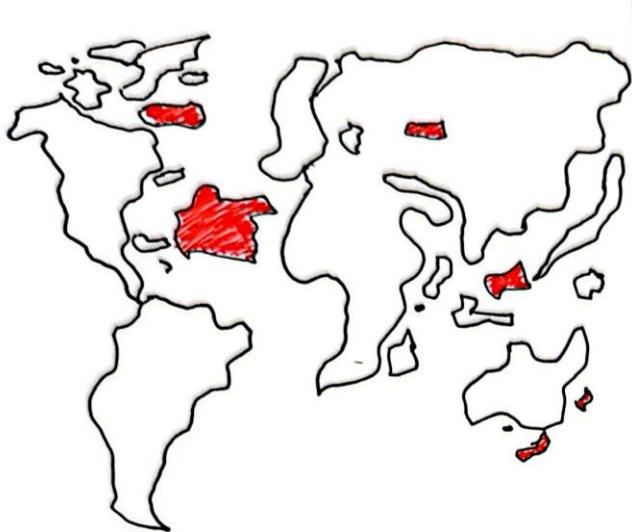


Figure 1: FDS World Map Choropleth

This concept visualises global tennis dominance by aggregating player-level data at the country level. The world map uses colour intensity to represent metrics such as number of top-100 players, total ranking points accumulated by each nation, or the proportion of match wins by players from that country. Countries with stronger professional tennis activity appear in darker shades, producing an intuitive and visually compelling representation of global participation and performance.

#### Strengths

- Provides an immediate high-level global overview
- Visually intuitive and appealing for poster presentation
- Highlights geographical patterns and national disparities
- Supports multiple dominance measurements (players, rankings, wins)

#### Limitations

- Geographic distortion may misrepresent small countries
- Countries with few but elite players (e.g., Serbia, Switzerland) may appear underrated
- Static choropleth cannot easily show temporal changes unless animated

### 2. Player Profile Performance Dashboard (Sheet 3)

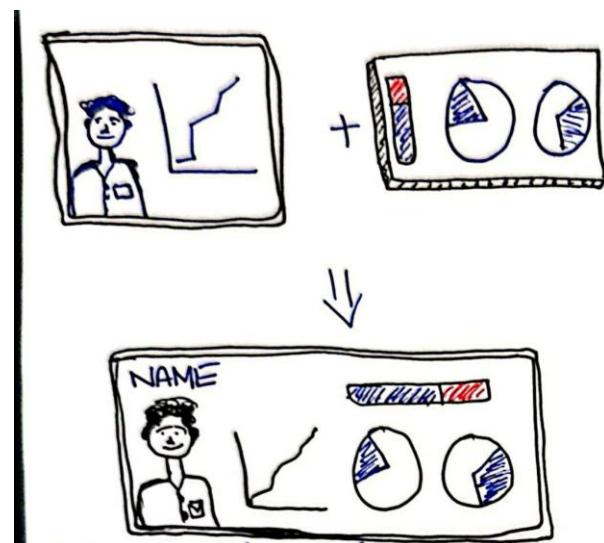


Figure 2: Player profile Performance Dashboard

This detailed dashboard focuses on a single player. It includes a player photograph, biographical information, ranking timeline, yearly performance bars, surface-wise pie charts, and mini-graphs of match statistics such as aces, double faults, and break-point performance. The aim is to create a magazine-style portrait of a player's entire career.

#### Strengths

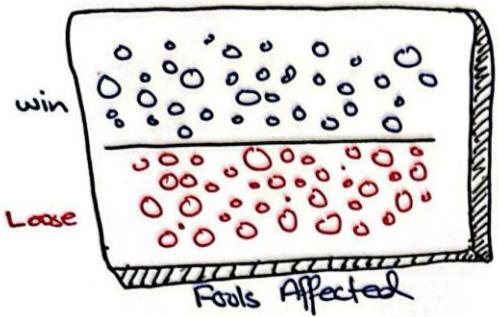
- Very rich storytelling at the individual player level
- Combines multiple visualisation types for a complete profile
- Makes ranking evolution and performance trends easy to understand
- Suitable for comparing top players if extended to multiple dashboards

#### Limitations

- Too detailed for country-level or global storytelling
- Requires clean, complete player statistics

- May overwhelm the viewer if not carefully arranged
- Limited usefulness for analysing broader patterns across many players

### 3. Scatterplot of Fouls vs Match Winning (Sheet 4)



*Figure 3: Scatterplot of Fouls vs match winning*

This design examines the relationship between double faults and match outcomes. A scatterplot is used where a separating diagonal line divides the space: points above the line represent matches won, while points below represent matches lost. The plot can use colour to differentiate surfaces or tournament levels.

#### Strengths

- Directly tests a performance-related hypothesis
- Simple and analytically strong visualisation
- Highlights outliers (e.g., players winning despite many double faults)
- Easy to interpret in academic or analytical contexts

#### Limitations

- Focused on only one statistic
- Does not capture multidimensional factors influencing match outcomes
- Not visually rich enough for a poster-level story
- Lacks the global/narrative appeal of other designs

## Critical Reflection Using CDS (Critical Design Strategy)

Using the CDS approach, each visualisation design was evaluated based on clarity, effectiveness, accessibility, novelty, data-mapping quality, and suitability for the final poster.

#### Clarity & Storytelling

- The **choropleth map** provides a very clear, instantly recognisable story about global tennis dominance.
- The **player dashboard** conveys deep insights but risks density and visual overload.
- The **scatterplot** is analytically strong but lacks narrative depth.

#### Data Mapping Quality

- The choropleth uses colour intensity effectively to show dominance, but requires careful colour-brewer palettes to avoid rainbow scales.
- The player dashboard uses multiple encodings (size, colour, shape), but could confuse inexperienced viewers.
- The scatterplot's binary win/loss division is clear but too simplistic for multi-factor tennis analysis.

#### Audience & Purpose

- The choropleth suits spectators, students, and analysts wanting broad comparisons.
- The player dashboard is ideal for detailed analyst-level investigations.
- The scatterplot is best for statistical audiences.

#### Novelty & Creativity

- The choropleth map offers creative potential, especially when enhanced with dominance metrics.
- The player dashboard is visually rich but less original.
- The scatterplot is traditional and expected.

## Feasibility

- The choropleth is straightforward to implement using libraries such as GeoPandas, Plotly, or D3.js.
- The player dashboard requires more time and multiple coordinated charts.
- The scatterplot is the easiest but least impactful.

## Overall Conclusion

The **choropleth world map** offers the strongest blend of clarity, storytelling power, visual appeal, feasibility, and relevance to the broad ATP dataset. It best represents the wide geographical spread of tennis talent and aligns most closely with the goals of Assessment 3.

## Final Chosen Design - World Map Choropleth (Sheet 5)

The final chosen design builds on the choropleth concept from Sheet 2 and incorporates implementation-level details required for the realisation sheet.

### Description of the Visualization

The final visualisation is a world map choropleth representing **country dominance in ATP tennis**, using aggregated metrics such as:

- Number of ATP players in the dataset
- Number of Top 100 or Top 10 players historically
- Total ranking points generated by players from each country
- Win percentages at ATP-level events

The map uses a colour-brewer-compliant sequential colour scale where darker colours indicate stronger dominance.

### Algorithms & Processing

- Aggregation algorithm:
  - Group data by country\_code
  - Compute dominance metrics (e.g., total ranking points, player counts)
  - Normalise values between 0–1 for consistent colouring
- Choropleth rendering:

- Use geographical shapefiles and merge with aggregated data
  - Map normalised dominance values to colour scale
- Map projection:
    - Robinson projection for balanced global display

## Dependencies & Libraries

Possible implementation tools:

- **Python:** Pandas, Matplotlib/Plotly, Numpy
- **Javascript:** D3.js,
- **Tableau (optional)**
- Colour palettes: ColorBrewer (per assignment rules)

## Interaction (Optional for Poster)

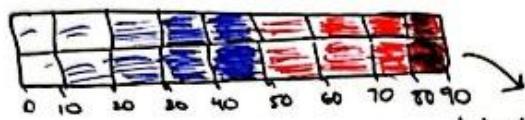
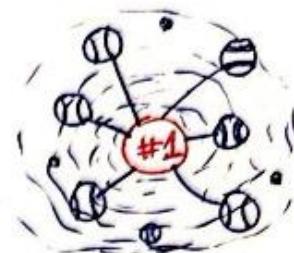
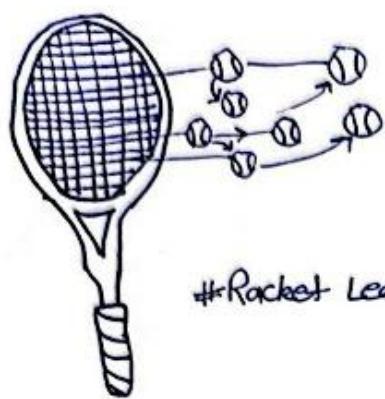
- Hover labels with country stats
- Highlight of top countries
- Tooltip with player names and ranking summaries

## Technical Requirements

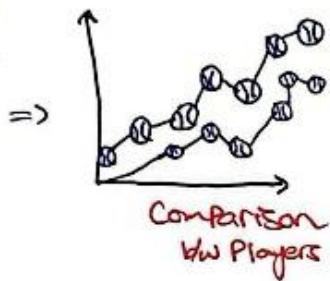
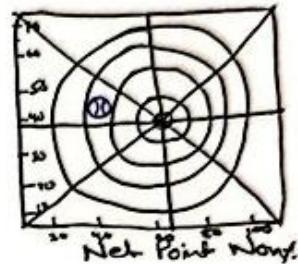
- Minimum display resolution:
  - 2500 px width for poster clarity

## Why This Design Was Selected

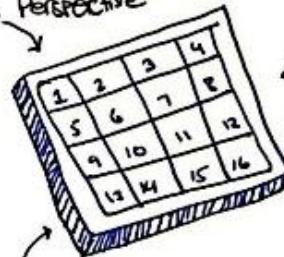
- Best balance of global storytelling and visual clarity
- Strong aesthetic impact suitable for an academic poster
- Easily understood by non-experts
- Represents the dataset at a macro level
- Allows identification of meaningful patterns such as dominance of Spain, USA, France, Serbia, and Switzerland
- Flexible enough for additional supporting charts in Assessment 3



Highest Number  
of Streaks.



Player's Perspective



← Yearly Calendar

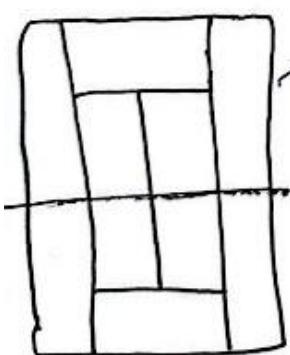
Colour density



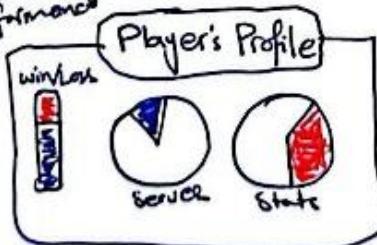
Do Age affect Player's Ranking?



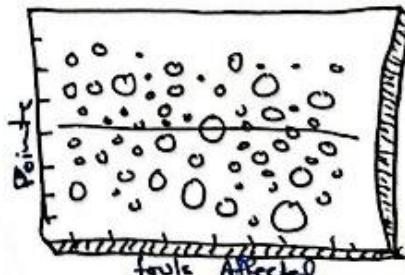
Country dominance?



show surface  
affect Player  
Performance



face  
analysis  
gas vs hard  
which Player is  
best in which surface.



## Big Picture/Layout



### WORLD MAP CHOROPLETH

This design explores which country dominates ATP tennis ranking data.

#### Focus

#### Key Question:-

- ① Which countries produced the highest-ranked players over time?
- ② Which countries consistently appear in the top 10/top 50/top 100?
- ③ Are certain countries improving or declining in tennis performance?
- ④ Do smaller countries produce elite players?
- ⑤ How do big nations (USA, Spain, France, Germany, UK) compare?

## Sheet 2

HASSAN YOUNAS

World Map Visualization

02.12.2025

#### COMPONENTS

##### A. Country Representation

- A world map, with intensity = dominance score

##### B. Data Input

- Total Player in dataset from each Country
- Number of Players in TOP 100, TOP 50 or TOP 10
- Total Ranking Points accumulated by Country
- Grand Slams won
- Avg ranking per country

##### C. Visual Encoding

- Colour intensity = Strength/Dominance
- Highlight size by winning

##### D. Time slider year by year

#### PROS & CONS

##### Pros:-

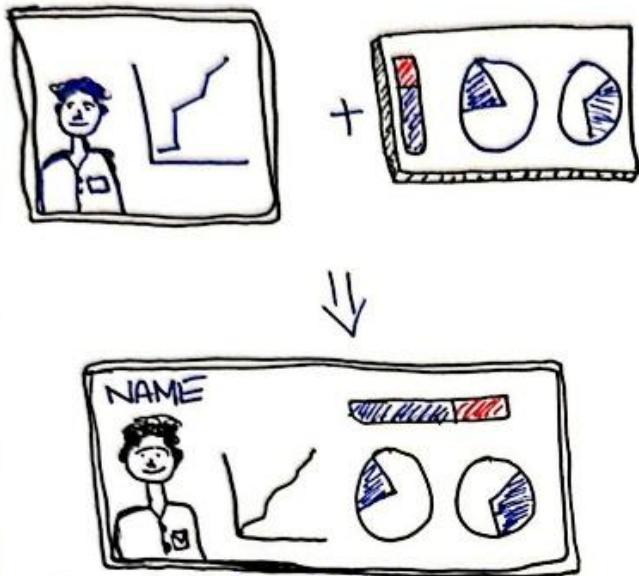
- Global Perspective
- Easy to understand
- Make Ranking Pattern obvious
- Strong Storytelling
- Visually attractive

##### Cons:-

- Country Population bias
- Map clutter
- Combining metric will be tricky
- Geographic Maps can instead act as area ≠ tennis strength
- A time-based component may be needed, but maps don't show time naturally.



## Big Picture



A Player-centric Performance dashboard that shows an individual tennis Player's Picture, ranking, match stats & career summary.

## Focus

- ↳ To showcase how a single ATP Player has performed throughout their career.
- ↳ To show metrics such as ranking over time, matches won/lost, aces, double faults, Surface Performance, and yearly summary.
- ↳ To Present an engaging, magazine-style Profile combining Player Photograph + visual Analytics

## Sheet 3

HASSAN YOUNAS

Player Profile Performance Dashboard

02.12.2025

### COMPONENTS

#### A. Player Profile Section:-

- ↳ Player Picture
- ↳ Name
- ↳ Country flag
- ↳ Career high ranking

#### B. Ranking Timeline (Main chart)

#### C. Yearly Performance Bars (Bar chart)

#### D. Surface Performance (Pie chart)

#### E. Visual Encoding

- ↳ Colour brewer safe Palette
- ↳ Photo - humanizes the data
- ↳ Icons

### Pros & Cons:-

#### Pros:-

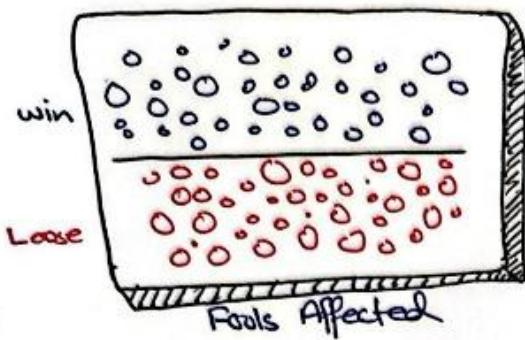
- very engaging & visually appealing
- Combines Identity + Stats
- Storytelling is strong
- Shows full career Progression
- Easy to compare Players
- Uses multiple visualisation types

#### Cons:-

- Too much information may feel dense
- Not suitable for comparing many Players at once.
- requires clean, complete stats per Player.
- May overwhelm viewers if not designed carefully.



## Big Picture layout



This design examines whether there is a relationship b/w foul-related statistics and match outcomes.

A scatterplot is used to visually compare foul counts to match results. A central reference line divides the plot. Points above the line indicate match won & vice versa.

### Focus

- ↳ To understand whether high foul counts reduce the chance of winning.
- ↳ To compare match outcomes visually against a foul metric
- ↳ To compare Patterns across multiple matches and Potentially across many players.
- ↳ To show correlation b/w fouls and the outcome clearly.
- ↳ To analyse Player discipline and its effect on Performance.

Sheet 4

HASSAN YOUNAS

Relationship b/w Fouls & Match Winning

03.12.2025

### Components

- A. ScatterPlot Layout
  - X-axis: Foul counts
  - Y-axis: Match outcome
- B. Central Decision Line
- C. Colour Encoding
  - Blue Point → Win
  - Red Point → Loose
- D. Size of Points, based on no. of fouls per match

### Pros & Cons

#### Pros:

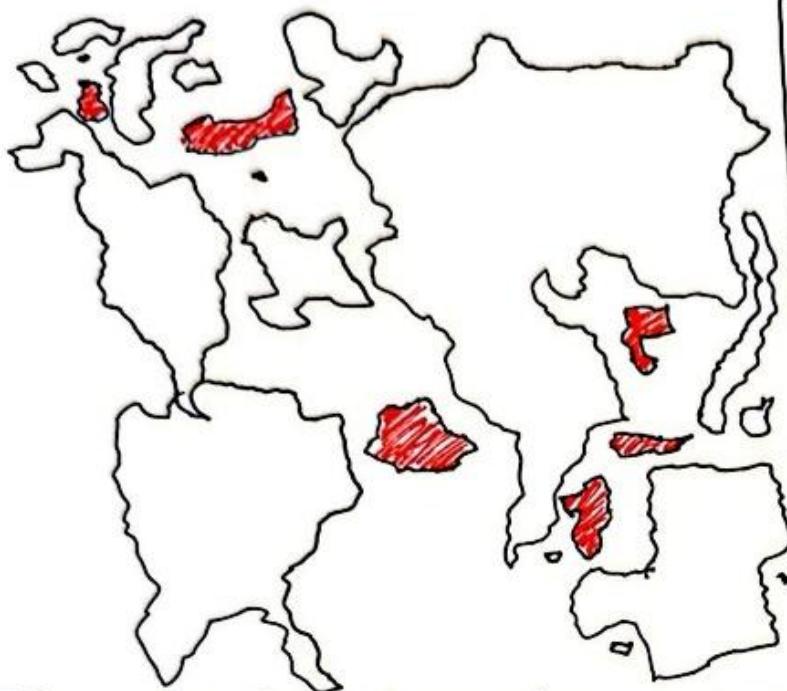
- Very clear statistical relations visualised.
- Scatterplot is ideal for spotting correlation
- Easy to identify outliers
- Shows actionable insight.
- Clean, simple, and interpretable visualisation.

#### Cons:

- Not Good for storytelling
- Only shows two stats
- Doesn't capture long trends (only Permatch)
- May clutter.



## Big Picture



This is the final chosen design: a choropleth world map showing tennis dominance by country.

## Focus

### Key Questions:-

- ① Which countries produced the highest-ranked players over time?
- ② Which countries consistently appear in the top 10/top 50/top 100?
- ③ Are certain countries improving or declining in tennis performance?
- ④ Do smaller countries produce elite players?
- ⑤ How do big nations (USA, UK, Spain, France, Germany) compare?
- ⑥ Geographic pattern of tennis strength

## SHEET 5

HASSAN YOUNAS

ATP Country Dominance  
Choropleth

03.12.2025

### Components

#### A. Choropleth Map (center)

- Countries shaded based on dominance score
- Dark, Light, Grey colour

#### B. Tooltip Box

- When hovering a country it will show country name, number of players, Avg ranking, #Top 10, highest ranking.

#### C. Visual Encoding

- Color Intensity = Strength
- Highlight size by winning

#### D. Time slider year by year

### Details

#### ① Algorithm:

↳ Aggregation Algorithm and Normalization will be done.

#### ② Dominance Score:-

$$DS = \frac{\text{Total Ranking Point} + \# \text{Top 100}}{\text{Number of Players}}$$

#### ③ Dependencies:-

- Python
- Pandas
- Matplotlib
- Numpy
- Tableau (optional)

#### ④ Estimated Time:-

- Data Cleaning 3-4 hrs
- Map generation 2-3 hrs



## Critical Design Survey (CDS)

## Step 1

### Assign a name to the design:

### **Summarise essence:**

### **Circle 5 (first impression) words:**

clear confusing sensible indifferent clever reliable pointless  
indistinctive complex organised moderate spectacular useless  
average bad fulfilling useful fair vague beautiful

## Step 2

- 1 Is suitable for the user and task
  - 2 Is understandable for user and task to hand
  - 3 It doesn't require guesswork
  - 4 Is trustworthy
  - 5 Would be useful  
  - 6 It would fit in with other technologies
  - 7 Uses suitable technology
  - 8 Has appropriate interaction
  - 9 Its sizing is correct
  - 10 Gives a positive ambience  
  - 11 Suitable user interface
  - 12 Ergonomic interface
  - 13 Facets are sized suitably
  - 14 Interface suitably spaced
  - 15 Suitable quantity of interface parts  
  - 16 Has all necessary components
  - 17 Has all suitable output/view types
  - 18 Clear relationships between parts
  - 19 Task can be easily performed
  - 20 Suitable organisation of components  
  - 21 Inspiring design
  - 22 Aesthetic and visually attractive
  - 23 Good composition and space utilisation
  - 24 Suitable coverage of data/underpinning facts
  - 25 Clear instructions, labels, legends to give context  
  - 26 Right choice of channels to communicate the message
  - 27 Communicates appropriate relationships/messages
  - 28 The types of marks used, communicate things clearly
  - 29 Components are shown at the right level of abstraction
  - 30 Nothing is hidden that shouldn't be hidden

	-2	-1	0	1	2	
Perception	0	0	0	0	●	Unsuitable → Suitable
Environment	0	0	0	0	●	Incomprehensible → Understandable
Interface	0	0	0	●	○	Requires guesswork → Clear assumptions
Components	0	0	0	0	●	Distrustful → Trustful
Design	0	0	0	0	●	Useless → Useful
Visual marks	0	0	0	0	●	Wrong setting → Right setting
	0	0	0	0	●	Unsuitable technology → Right technology
	0	0	0	●	○	Unsuitable interaction → Appropriate interaction
	0	0	0	0	●	Unsuitable size → Suitable physical size
	0	0	0	0	●	Poor vibe/ambience → Positive ambience
	0	0	0	0	●	Unsuitable GUI → Suitable GUI
	0	0	0	●	○	Uncomfortable → Ergonomic
	0	0	0	0	●	Poorly proportioned → Suitable sized facets
	0	0	0	0	●	Poor facet spacing → Relevant spacing
	0	0	0	0	●	Unsuitable facet quantity → Suitable facet quantity
	0	0	0	0	●	Missing components → All necessary components
	0	0	0	0	●	Unsuitable types → Suitable view types
	0	0	0	0	●	Unclear correspondences → Clear view relationships
	0	0	0	●	○	Task unfulfilled → Task easily performed
	0	0	0	0	●	Poor component layout → Good component layout
	0	0	0	●	○	Uninspiring → Inspiring
	0	0	0	0	●	Unattractive → Visually attractive (aesthetic)
	0	0	0	0	●	Poor layout → Good composition
	0	0	0	0	●	Unsuitable coverage → Suitable coverage
	0	0	0	0	●	Poor labels/legends → Suitable legends/labels
	0	0	0	0	●	Poor choice of channels → Good channel choices
	0	0	0	0	●	Inappropriate mappings → Appropriate mappings
	0	0	0	0	●	Inappropriate mark types → Suitable mark types
	0	0	0	0	●	Poor scale/zoom → Good scale/zoom
	0	0	0	0	●	Overplotting → Clear display, easy read

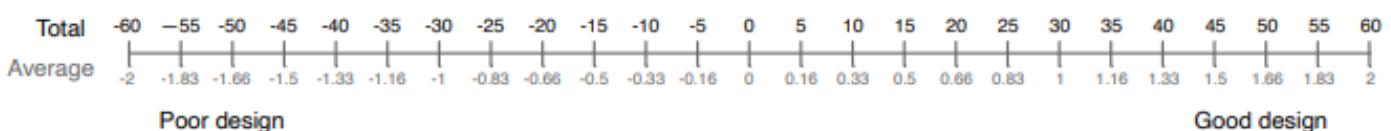
### Step 3

## Reflect

Sum values

Total 54

## Improvements



## **ACKNOWLEDGMENTS**

I wish to express my deepest gratitude to Jeff Sackmann for providing the ATP tennis dataset, which served as the cornerstone of this analysis. I am also profoundly grateful to my supervisor, Jonathan C. Roberts, for his expert guidance, constructive feedback, and unwavering support throughout the development of this project.

## **REFERENCES**

- [1] J. Sackmann, “tennis\_atp”, GitHub.[Online]. Available:  
[https://github.com/JeffSackmann/tennis\\_atp/blob/master/README.md](https://github.com/JeffSackmann/tennis_atp/blob/master/README.md).  
[Accessed: Nov. 06, 2025].