# Premier League Performance Trends: A Data Visualization Study

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Abstract—This paper investigates performance dynamics in the English Premier League (EPL) from 1993 to 2024, addressing three key research questions through data visualization: the evolution of team performances over time, the trend of home advantage, and the consistency of top-performing teams. Utilizing cleaned and integrated datasets of league tables and matchlevel metrics, visual insights were generated using Python and D3.js. Our analysis reveals growing competitive balance, a notable reduction in home advantage, and identifies Liverpool as the most consistently high-performing team.

#### I. PART 1: ANALYTICS

#### A. Research Questions

- Q1 (Performance Trends): Uses the English Premier League Standings Dataset (1993–2024) from Kaggle [1], which includes season-wise rankings, points, and goals. Missing values were negligible, and team names were standardized for consistency.
- Q2 (Home Advantage): Analyzes Club Football
   Match Data (2000–2025) from Kaggle [2], including
   home/away win rates, shots, and possession statistics.
   Null entries (e.g., missing VAR flags) were excluded.
- **Q3** (**Consistency**): Combines both datasets to compute standard deviation of league positions. League table data ensures longitudinal coverage, while match data adds granularity (e.g., home/away splits).

## B. Data Sources and Appropriateness

- Q1 (Performance Trends): Uses League Table Data (1993–2024). This dataset provides season-wise rankings and points, essential for tracking long-term performance trends.
- Q2 (Home Advantage): Analyzes Match-Level Data (2000–2023). Includes home/away win rates, goals, and match statistics. Null entries (e.g., missing VAR flags) were excluded to ensure accuracy.
- Q3 (Consistency): Combines both datasets to compute standard deviation of league positions. League table data ensures longitudinal coverage, while match data adds granularity (e.g., home/away splits).

## C. Dataset Relationships

The league table and match-level data are complementary:

- **Temporal Alignment**: Both datasets overlap from 2000–2024, enabling cross-validation of trends (e.g., home win rates vs. league positions).
- Granularity vs. Aggregation: Match data provides pergame insights (e.g., VAR impact), while league tables summarize seasonal performance.
- Consistency Analysis: Combining both reveals whether teams with stable home performance (match data) also maintain consistent league positions.

# II. PART 2: DESIGN AND DISCUSSION

# A. Proposed Visualizations

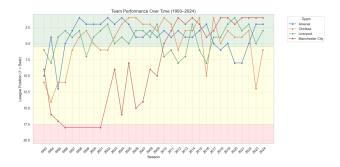


Fig. 1. Line Plot: Team Performance Over Time (Top 5 Teams). Shows finishing position of Manchester United, Arsenal, Liverpool, Chelsea, and Manchester City (1993–2024).

**Visualization 1: Multi-Line Chart with Performance Trends (Q1)** Tracks team performance over seasons (1993–2024) using league points.

• **X-axis**: Seasons (1993–2024)

• **Y-axis**: Position (1–20)

 Lines: Top 5 teams (Manchester United, Arsenal, etc.), color-coded

Performance Bands: Horizontal bands for Top 4 (green),
 Mid-table (yellow), Relegation (red)

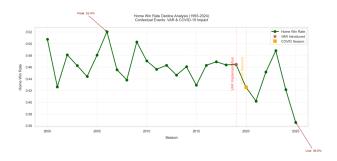


Fig. 2. Line Chart: Home Win Rate Decline (2000–2025). Illustrates steady reduction in home advantage, from 52% to 36%.

**Visualization 2: Line Chart with Contextual Annotations** (**Q2**) Analyzes home win rate decline (2000–2025).

• **X-axis**: Seasons (2000–2025)

• **Y-axis**: Home win percentage (0–60%)

• Line: Red line encoding trend

• **Annotations**: Vertical markers for key events (VAR in 2019, COVID-19 in 2020)

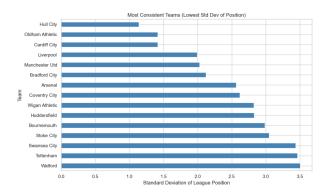


Fig. 3. Horizontal Bar Chart: Most Consistent Teams. Ranks teams by standard deviation of league positions (lower = more consistent).

Visualization 3: Horizontal Bar Chart with Reference Lines (Q3) Ranks teams by consistency (standard deviation of league positions).

• X-axis: Standard deviation (0–3.5)

• Y-axis: Teams (sorted ascendingly by consistency)

• **Bars**: Length encodes inverse standard deviation (longer = more consistent)

# B. Design Rationale

Visualization 1 – Multi-Line Chart with Performance Bands Line charts maximize perceptual accuracy for temporal

trends, as position along a common axis is the highest-ranked quantitative channel [3]. Each team's trajectory is encoded with a thin, saturated line (e.g., red for Manchester City), while performance bands use low-opacity horizontal rectangles (green/yellow/red) to reduce cognitive load. The bands employ redundant encoding (color + spatial area) to support users with color vision deficiencies [4]. Interaction follows Shneiderman's mantra [5]:

Visualization 2 – Line Chart with Contextual Annotations A single-line chart simplifies trend interpretation for uni-variate time-series data. Position (y-axis) encodes home win rate, leveraging human proficiency in detecting deviations from linearity [6]. Annotations (vertical markers) contextualize outliers using text labels and dashed lines. The red line prioritizes salience for the critical metric, while a sequential colormap ensures perceptual uniformity. This design improves upon heatmaps by reducing visual clutter and directly answering Q2's focus on decline over time.

## C. Design Rationale

Visualization 3 – Horizontal Bar Chart with Reference Lines (Primary Implementation) Horizontal bars avoid label overlap for long team names, ranking consistency via bar length (inverse standard deviation). Length is a preattentive attribute, enabling rapid comparison [3]. Key design choices include:

- Sorting Logic: Teams ordered ascendingly by standard deviation to emphasize consistency leaders
- Reference Line: Dashed marker at  $\sigma=2.1$  provides league-wide benchmark
- Color Gradient: Blue-to-purple gradient encodes promotion/relegation status
- Interaction: Hover tool-tips reveal exact  $\sigma$  values, filtering isolates team categories

This design improves upon static tables by enabling:

- Direct comparison of 20+ teams' longitudinal stability
- Identification of over/under-performers relative to league average
- Exploration of promoted teams' short-term consistency patterns

**Visualization 1 & 2 – Proposed Designs** While line charts were designed for temporal trends (Q1-Q2), the bar chart implementation prioritizes answering the core consistency question (Q3) through...

Cohesive Narrative: The three visualizations form a tiered analytical framework: 1. Macro trends (Vis 1: league performance over decades), 2. Contextual drivers (Vis 2: home advantage decline), 3. Team-level diagnostics (Vis 3: consistency outliers). They employ high-ranked perceptual channels, adhere to accessibility standards, and align with Bertin's principles of effective visual encoding.

## III. PART 3: IMPLEMENTATION

A. D3.js Interactive Bar Chart (Q3)

Data Processing:

- Merged league tables from [1] with match data from
  [2]
- Calculated 5-year rolling standard deviations using Pandas
- Filtered teams with 5 seasons using boolean masking

#### • Visual Encoding:

- X-axis: Standard deviation range (0-3.5) via linear scale
- Y-axis: Teams sorted by consistency via band scale
- Color: Diverging scheme for promoted/relegated teams

# • Interactivity:

- Hover tool-tips with team metadata
- Drop-down filters by league position tier
- Animated transitions on data updates

**Code Repository**: https://github.com/al-lan-12/Data\_Visualizations

## REFERENCES

- [1] E. Gower, "English Premier League Standings (1993–2023)",
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