

In-Class Problem Set: Scatterplots and Association (R + GitHub)

Goal. Use NBA player data to practice visualizing relationships between **two quantitative variables** using scatterplots, smoothers, polynomial fits, and careful interpretation. You will pull the data from GitHub, clean variable types, generate required figures, interpret what they show, and submit via GitHub.

Dataset. `basketball` (569 rows; 23 variables). Key variables you may use:

- IDs and labels: `PLAYER_NAME`, `TEAM_ABBREVIATION`
- Player attributes: `AGE`, `PLAYER_HEIGHT_INCHES`, `PLAYER_WEIGHT`
- Games + performance: `GP`, `PTS`, `REB`, `AST`, `NET_RATING`
- Rates (0–1): `OREB_PCT`, `DREB_PCT`, `USG_PCT`, `TS_PCT`, `AST_PCT`
- Draft info: `DRAFT_YEAR` (contains “Undrafted” for some players)

Important data note. Many columns are stored as **character strings**. You must convert variables you analyze into appropriate numeric types before plotting.

What to submit (in your GitHub repo).

- A script file: `scripts/lab.R`
- A short write-up: `outputs/writeup.md`
- Saved figures in `figures/` (see requirements below)

Rules.

- Work inside an **R Project**.
- Use a **sequential, hard-coded workflow** (no user-defined functions).
- Save figures using `ggsave()` (no screenshots).
- Git commands must be run in the **Terminal tab**, not the R Console.
- Use `theme_classic()` unless you explicitly justify an alternative.
- Handle missing values defensibly (state what you did).

Questions

1. Pull the repo and confirm the dataset (proof required).

- (a) In the **Terminal tab**, run:

```
git status  
git pull
```

- (b) Confirm the dataset file exists in your repo (path posted in the course repository).

- (c) Create the standard folder structure (if missing): `scripts/`, `outputs/`, `figures/`.

- (d) **Proof (write-up):** In `outputs/writeup.md`, paste:

- the output of `getwd()`,
- the output of `list.files()` from the project root, and
- the output of `list.files("data")` showing the dataset file.

2. Load and clean basketball (proof required).

- (a) Load the dataset into an object named `basketball`.
- (b) Create a cleaned object named `basketball_clean` where you convert the following columns to numeric:

```
AGE, PLAYER_HEIGHT_INCHES, PLAYER_WEIGHT, GP, PTS, REB, AST, NET_RATING,
OREB_PCT, DREB_PCT, USG_PCT, TS_PCT, AST_PCT.
```

- (c) Create a simple draft indicator:

```
draft_status = "Undrafted" vs "Drafted"
```

where "Drafted" means `DRAFT_YEAR` is not "Undrafted".

- (d) **Proof (write-up):** Report:

- the number of rows in `basketball` and `basketball_clean`,
- a quick summary of at least three numeric columns (e.g., `AGE`, `PTS`, `USG_PCT`),
- one sentence describing how you handled missing or non-numeric values after conversion.

3. Relationship 1: usage and scoring (scatterplot baseline).

Make a scatterplot of `USG_PCT` (x) vs `PTS` (y).

- Use `geom_point()` with an overplotting fix (e.g., `alpha` and/or smaller `size`).
- Use clear labels. Since `USG_PCT` is a proportion, format the x-axis as percent if you can.
- Use `theme_classic()`.

Save as:

```
figures/usg_pts_scatter.png
```

4. Relationship 1 (extension): add a linear smoother with standard errors.

Using the same x/y pairing as the previous question, add a linear fit:

- `geom_smooth(method = "lm", se = TRUE)`
- Keep the points visible (do not remove them).

Write-up (3–5 sentences):

- What does the fitted line claim?
- What does the shaded band represent (in plain language)?
- Does the band make you more or less confident about the trend?

Save as:

```
figures/usg_pts_lm_se.png
```

5. Relationship 1B: usage and scoring efficiency.

Make a scatterplot of `USG_PCT` (x) vs `TS_PCT` (y).

- Use `geom_point()` with an overplotting fix (e.g., `alpha` and/or smaller `size`).
- Format `USG_PCT` and `TS_PCT` as percents if you can.
- Add a linear smoother with standard errors: `geom_smooth(method = "lm", se = TRUE)`.
- Use `theme_classic()`.

Write-up (3–5 sentences):

- Does the trend look linear, or do you suspect curvature?
- What does the SE ribbon suggest about uncertainty across usage levels?

Save as:

```
figures/usg_ts_eff_lm_se.png
```

6. Relationship 1C: assists (raw vs rate).

Make a scatterplot of `AST` (x) vs `AST_PCT` (y).

- Use `geom_point()` with an overplotting fix (e.g., `alpha` and/or smaller `size`).
- Format `AST_PCT` as a percent if you can.
- Add a linear smoother with standard errors: `geom_smooth(method = "lm", se = TRUE)`.
- Use `theme_classic()`.

Write-up (3–5 sentences):

- Are `AST` and `AST_PCT` close to a one-to-one relationship, or are there notable exceptions?
- Give one plausible basketball reason you might see players with similar `AST` but different `AST_PCT`.

Save as:

```
figures/ast_astpct_lm_se.png
```

7. Final interpretation (write-up required).

In `outputs/writeup.md`, write 12–16 sentences addressing:

- For each figure, what is the main pattern (direction + strength + shape)?
- Name one outlier or “surprising” point pattern and what it could imply (without claiming causality).
- Name one concrete plotting choice you made (`alpha`, axis formatting, legend placement, polynomial fit) and why it helped interpretability.

8. Git workflow and submission (proof required).

You must show evidence of both `pull` and `push`, plus at least two commits.

- (a) After you finish cleaning the data (Question 2), commit and push:

```
git status
git add .
git commit -m "NBA relationships: clean basketball types"
git push
```

- (b) Before your final push, run a fresh pull (to catch updates):

```
git pull
```

- (c) Commit and push your figures + write-up:

```
git status
git add .
git commit -m "NBA relationships: scatterplots + smoothers + writeup"
git push
```

- (d) **Proof (write-up):** Paste:

- the output of `git status` after the final push (clean working tree), and
- the output of `git log -2`.

Optional challenge (if you finish early)

Choose one:

- Create a small-multiple version of one relationship by faceting on `TEAM_ABBREVIATION` for **only** the teams with at least 12 players in the dataset. (State your rule and why you chose it.)
- Create a correlation heatmap for the numeric columns you used (and in 4–6 sentences, explain what the heatmap hides that a scatterplot reveals).

Checklist (before you leave)

- `scripts/lab.R` runs top-to-bottom
- Required figures exist in `figures/`:
 - `usg_pts_scatter.png`, `usg_pts_lm_se.png`
 - `usg_ts_eff_lm_se.png`, `ast_astpct_lm_se.png`
 - `size_reb_scatter.png`
 - `age_ts_poly2_se.png`
- `outputs/writeup.md` includes required proofs + interpretation
- At least two commits + pushed to GitHub