## Data Viz I

Dr. Adrian Correndo

2025-01-31

#### Description

In this class, we will explore advanced visualization techniques using ggplot2. You'll learn how to fine-tune plots by modifying shapes, lines, legends, and adding custom titles and annotations. We will continue working with the agridat::lasrosas.corn dataset.



## 1 Learning Objectives

By the end of this session, you will: 1. Customize plot aesthetics, including point shapes, line types, and colors. 2. Modify legends and axis labels for better readability. 3. Use annotations and text elements to highlight key data points. 4. Adjust themes and layouts for professional presentation.

#### Required Packages

Install the necessary packages for this class:

```
library(pacman)
p_load("ggplot2", "gganimate", "agridat", "patchwork", "sf", "ggtext", "dplyr")
```

#### 2 Datasets: Las Rosas Corn Data

We'll use the agridat::lasrosas.corn dataset, focusing on yield, nitro levels, years, and spatial data.

```
corn_data <- agridat::lasrosas.corn %>%
  mutate(Year = as.factor(year))

glimpse(corn_data)
```

```
Rows: 3,443
Columns: 10
$ year <int> 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999~
                          <dbl> -33.05113, -33.05115, -33.05116, -33.05117, -33.05118, -33.05120~
$ lat
$ long <dbl> -63.84886, -63.84879, -63.84872, -63.84865, -63.84858, -63.84851~
$ yield <dbl> 72.14, 73.79, 77.25, 76.35, 75.55, 70.24, 76.17, 69.17, 69.77, 6~
$ nitro <dbl> 131.5, 131.5, 131.5, 131.5, 131.5, 131.5, 131.5, 131.5, 131.5, 1~
                          <dbl> 162.60, 170.49, 168.39, 176.68, 171.46, 170.56, 172.94, 171.86, ~
$ bv
$ rep
                          $ nf
                          <fct> 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1999, 1990, 1990, 1990, 1990, 1990, 1990, 1990, 1990, 1990, 1990, 
$ Year
```

```
nass_soybean <- agridat::nass.soybean %>%
  mutate(Year = as.factor(year))
select_states <- c("Kansas", "Iowa", "Illinois", "Missouri")
glimpse(nass_soybean)</pre>
```

```
Rows: 2,528
Columns: 5
$ year <int> 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924~
$ state <fct> Alabama, Arkansas, Delaware, Georgia, Illinois, Indiana, Iowa, K~
$ acres <dbl> 3000, 3000, 12000, 10000, 115000, 66000, 10000, 2000, 9000, 8000~
$ yield <dbl> 6.5, 6.5, 11.0, 5.5, 12.0, 9.9, 12.0, 11.0, 9.5, 8.0, 11.8, 13.0~
$ Year <fct> 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924, 1924,
```

# 3 1. Customizing ggplot2 Visualizations

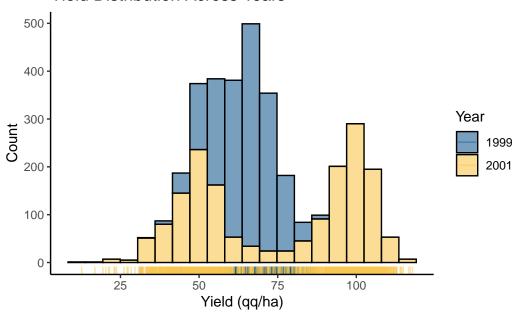
#### 3.1 Define custom palettes

```
my_colors <- c("#1e6091", "#f9c74f", "#9b2226", "#599999", "#8e5572")
```

#### 3.2 Histogram with Adjusted Transparency and Custom Bins

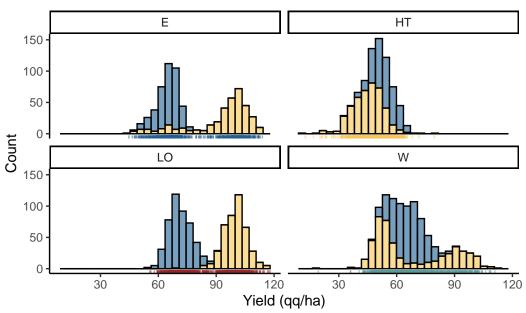
```
# Density
density_01 <-
corn_data %>%
ggplot(aes(x = yield)) +
  geom_density(aes(fill = Year), color = "grey15", alpha = 0.5)+
 labs(title = "Yield Distribution Across Years", x = "Yield (qq/ha)", y = "Count") +
  scale_fill_manual(values = my_colors)+
  theme_classic()
# Histogram
histo_01 <-
ggplot(corn_data, aes(x = yield)) +
  geom_histogram(aes(fill = Year), bins = 20, alpha = 0.6, color = "black") +
 labs(title = "Yield Distribution Across Years", x = "Yield (qq/ha)", y = "Count") +
  scale_fill_manual(values = my_colors)+
  geom_rug(aes(color = Year), alpha = 0.5)+
  scale_color_manual(values = my_colors)+
  theme_classic()
histo 01
```

#### **Yield Distribution Across Years**



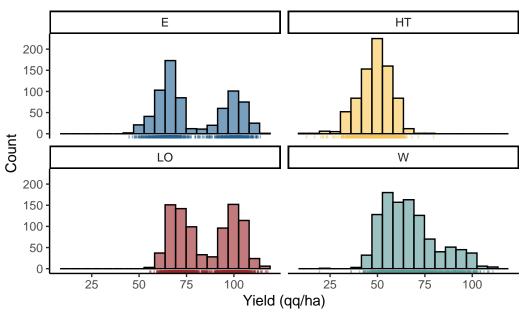
```
# Faceting by Year
histo_02 <-
corn_data %>%
ggplot(aes(x = yield)) +
    geom_histogram(aes(fill = Year), bins = 30, alpha = 0.6, color = "black") +
    labs(title = "Yield Distribution Across Years", x = "Yield (qq/ha)", y = "Count") +
    scale_fill_manual(values = my_colors)+
    geom_rug(aes(color = topo), alpha = 0.5)+
    scale_color_manual(values = my_colors)+
    facet_wrap(~topo)+
    theme_classic()+
    theme(legend.position = "none")
histo_02
```

#### Yield Distribution Across Years



```
corn_data %>%
ggplot(aes(x = yield)) +
  geom_histogram(aes(fill = topo), bins = 20, alpha = 0.6, color = "black") +
  labs(title = "Yield Distribution Across Years", x = "Yield (qq/ha)", y = "Count") +
  scale_fill_manual(values = my_colors)+
  geom_rug(aes(color = topo), alpha = 0.5)+
  scale_color_manual(values = my_colors)+
  facet_wrap(~topo)+
  theme_classic()+
  theme(legend.position = "none")
```

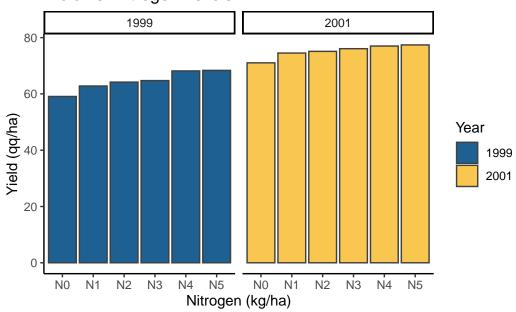
#### **Yield Distribution Across Years**



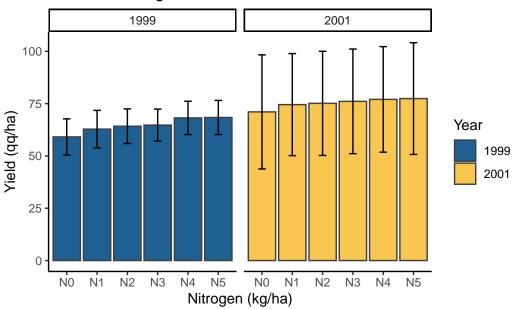
#### 3.3 Column/Bar plot

`summarise()` has grouped output by 'Year', 'year', 'nitro'. You can override using the `.groups` argument.

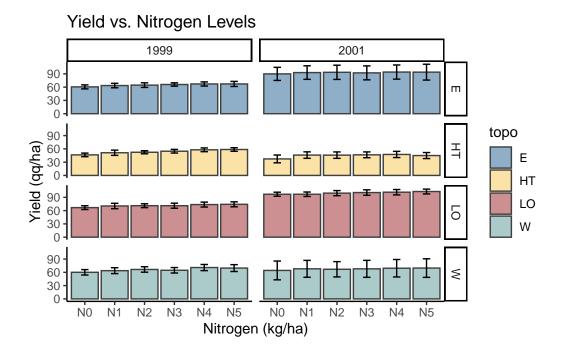
## Yield vs. Nitrogen Levels



#### Yield vs. Nitrogen Levels



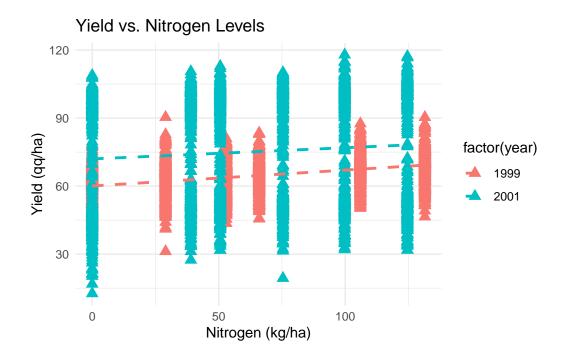
`summarise()` has grouped output by 'Year', 'year', 'nitro', 'nf'. You can override using the `.groups` argument.



#### 3.4 Scatter Plot with Custom Shapes and Line Types

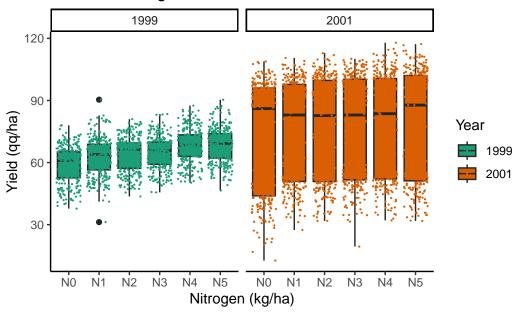
```
corn_data %>%
ggplot(aes(x = nitro, y = yield, color = factor(year))) +
  geom_point(size = 3, shape = 17) + # Triangle shape
  geom_smooth(method = "lm", se = FALSE, linetype = "dashed") +
  labs(title = "Yield vs. Nitrogen Levels", x = "Nitrogen (kg/ha)", y = "Yield (qq/ha)") +
  theme_minimal()
```

<sup>`</sup>geom\_smooth()` using formula = 'y ~ x'



#### 3.5 BoxPlot with Custom Shapes and Line Types

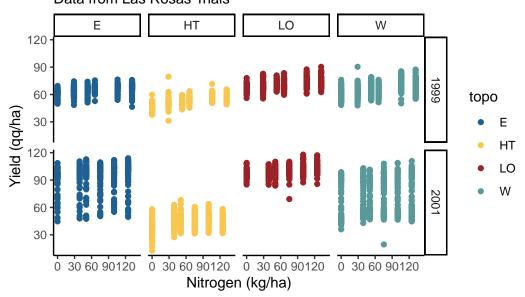
#### Yield vs. Nitrogen Levels



#### 4 2. Axis & Titles

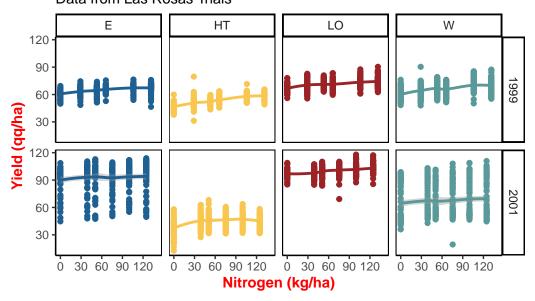
```
scatter_plot <-
corn_data %>%
ggplot(aes(x = nitro, y = yield, color = topo)) +
 geom_point()+
 # Add labels
 labs(title = "Yield vs. Nitrogen",
      subtitle = "Data from Las Rosas Trials",
      x = "Nitrogen (kg/ha)",
      y = "Yield (qq/ha)") +
 # Add manual color scale
 scale_color_manual(values = my_colors)+
 # Modify scale of x-axis
 scale_x_continuous(limits = c(0,140), breaks = seq(0,150, by=30))+
 facet_grid(Year~topo)+
 theme_classic()
scatter_plot
```

# Yield vs. Nitrogen Data from Las Rosas Trials



 $geom_smooth()$  using method = 'loess' and formula = 'y ~ x'

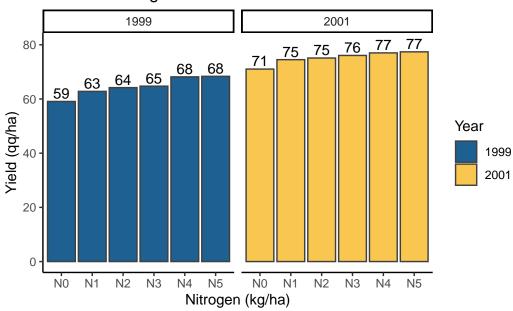
Yield vs. Nitrogen
Data from Las Rosas Trials



# 5 3. Annotating text

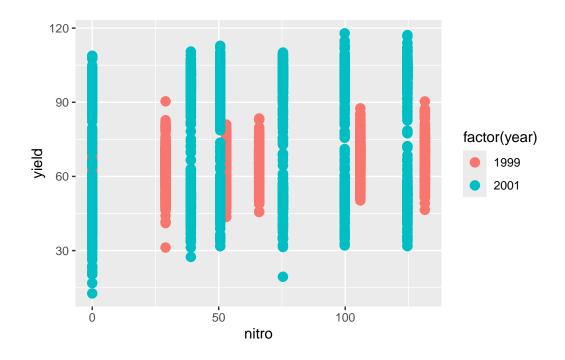
```
colplot_01 +
  geom_text(aes(label = round(yield_mean,0), x = nf, y = yield_mean + 3))+
  scale_color_manual(values=my_colors)
```



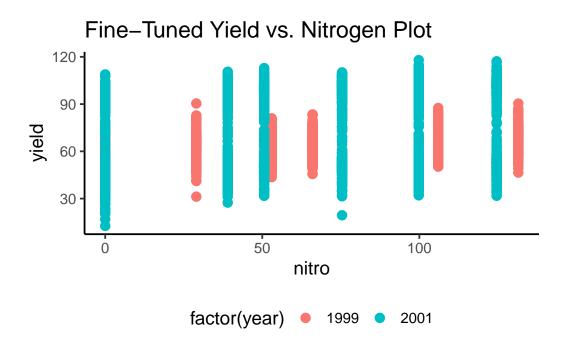


# 6 4. Adjusting Themes for Presentation

```
base_plot <- ggplot(corn_data, aes(x = nitro, y = yield, color = factor(year))) +
   geom_point(size = 3)
base_plot</pre>
```



```
base_plot +
  labs(title = "Fine-Tuned Yield vs. Nitrogen Plot") +
  theme_classic(base_size = 14) + # Larger base font
  theme(legend.position = "bottom")
```

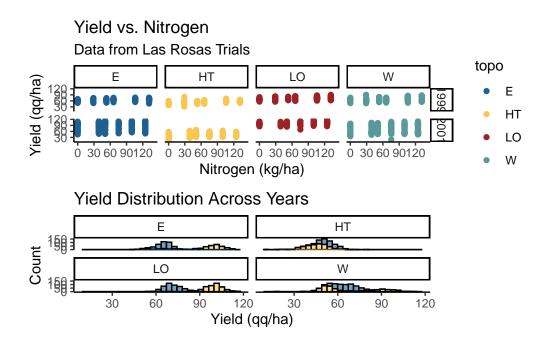


# 7 5. Combining Multiple Plots with Custom Layouts

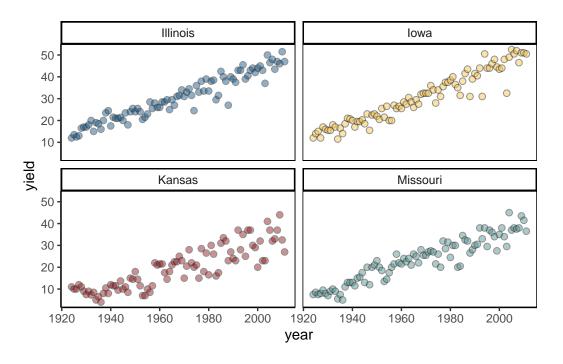
```
p1 <- scatter_plot

p2 <- histo_02

p1 + p2 + plot_layout(ncol = 1)</pre>
```



### 8 6. Animated plots



#### 9 Conclusion

This expanded tutorial provides various ggplot2 customization techniques relevant to agricultural research. Experiment with different styling options to enhance the clarity and impact of your visualizations!