

The background is a solid green color with a subtle gradient. In the four corners, there are decorative white line-art patterns that resemble circuit traces or data paths, with small circles at the end of the lines.

TUTORIAL 4

Introductory Statistics with *R*

Two-sample *t*-test

I. Data & Hypothesis

- `ozone` ← `read_csv("ozone.csv")`
- **Question:** Ozone level differs between east/west?
- **Null Hypothesis** (H_0): No difference ($\mu_1 = \mu_2$)

```
glimpse(ozone)
```

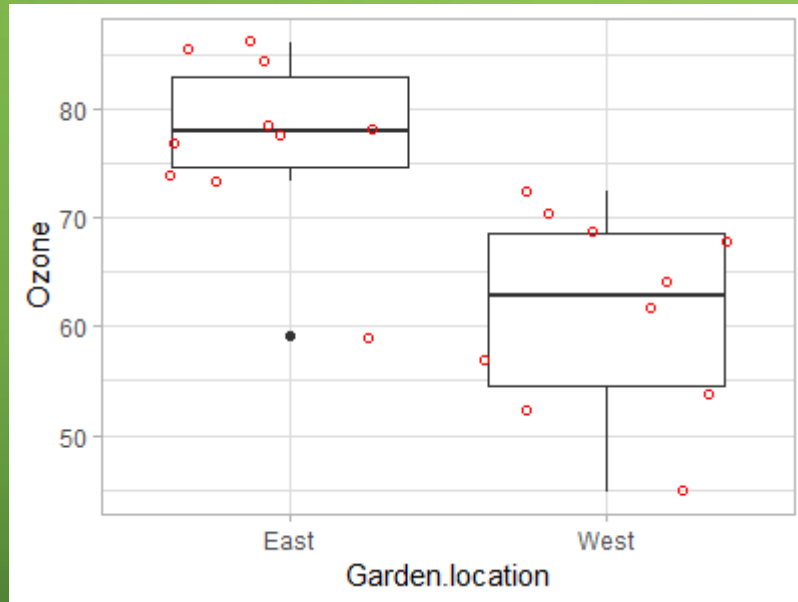
```
## Observations: 20
## Variables: 3
## $ Ozone          (dbl) 61.7, 64.0, 72.4, 56.8, 52.4, 4...
## $ Garden.location (fctr) West, West, West, West, West, ...
## $ Garden.ID      (fctr) G1, G2, G3, G4, G5, G6, G7, G8...
```

| | Ozone | Garden.location | Garden.ID |
|----|-------|-----------------|-----------|
| | <dbl> | <chr> | <chr> |
| 1 | 61.7 | West | G1 |
| 2 | 64 | West | G2 |
| 3 | 72.4 | West | G3 |
| 4 | 56.8 | West | G4 |
| 5 | 52.4 | West | G5 |
| 6 | 44.8 | West | G6 |
| 7 | 70.4 | West | G7 |
| 8 | 67.6 | West | G8 |
| 9 | 68.8 | West | G9 |
| 10 | 53.7 | West | G10 |
| 11 | 59.1 | East | G11 |
| 12 | 78.5 | East | G12 |
| 13 | 73.9 | East | G13 |
| 14 | 86.1 | East | G14 |
| 15 | 78 | East | G15 |
| 16 | 84.4 | East | G16 |
| 17 | 77.7 | East | G17 |
| 18 | 76.8 | East | G18 |
| 19 | 85.6 | East | G19 |
| 20 | 73.3 | East | G20 |

Two-sample t -test

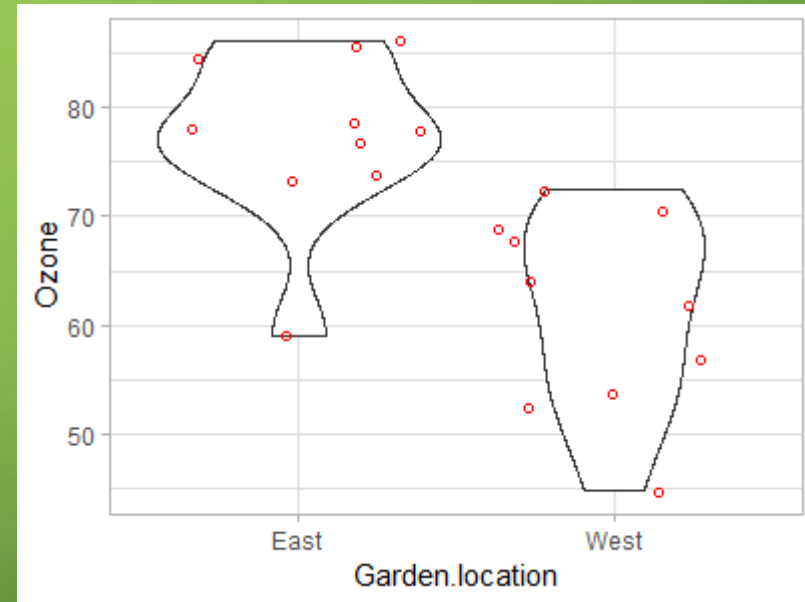
II. Data Visualization

Boxplot



```
ozone %>%  
  ggplot(data = ozone, aes(x =  
Garden.location, y = Ozone)) +  
  geom_boxplot() +  
  geom_jitter(shape=1, color="red") +  
  theme_bw()
```

Violin plot



```
Ozone %>%  
  ggplot(data = ozone, aes(x = Garden.location, y  
= Ozone)) +  
  geom_violin() +  
  geom_jitter(shape=1, color="red") +  
  theme_bw()
```

Two-sample *t*-test

III. Run *t*-test

```
# Do a t.test now....  
t.test(Ozone ~ Garden.location, data = ozone)
```

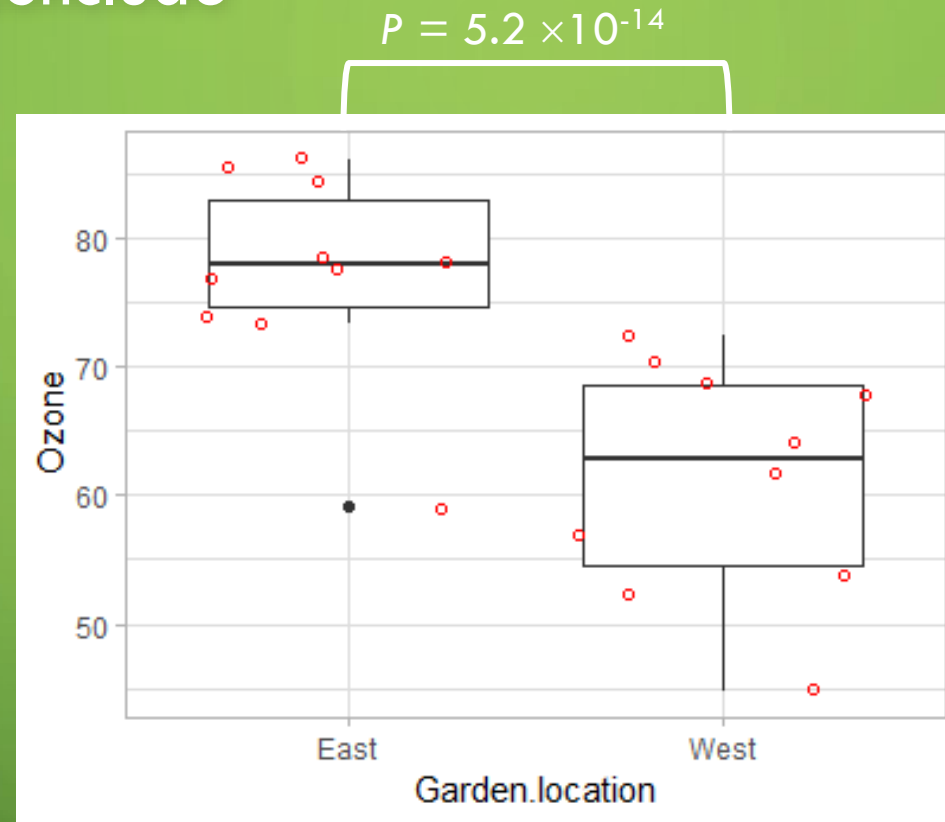
```
##  
## Welch Two Sample t-test  
##  
## data: Ozone by Garden.location  
## t = 4.2363, df = 17.656, p-value = 0.0005159  
## alternative hypothesis: true difference in means is not equal to 0  
## 95 percent confidence interval:  
## 8.094171 24.065829  
## sample estimates:  
## mean in group East mean in group West  
## 77.34 61.26
```

P value –

- Probability that observed difference is due to chance
- (more specifically) probability that $t \geq 4.2363$ under null hypothesis (H_0)

T-test

IV. Re-plot & Conclude



Conclusions:

- **Statistical conclusion:** The null hypothesis (same mean) is rejected at $p=5.2 \times 10^{-14}$
- **Biological conclusion:** The ozone level is significantly different between the east & west locations

Linear Regression

I. Data & Hypothesis

- Biological Question: Does soil moisture affect growth rate?
- Null Hypothesis (H_0): No correlation ($r=0$)

```
glimpse(plant_gr)
```

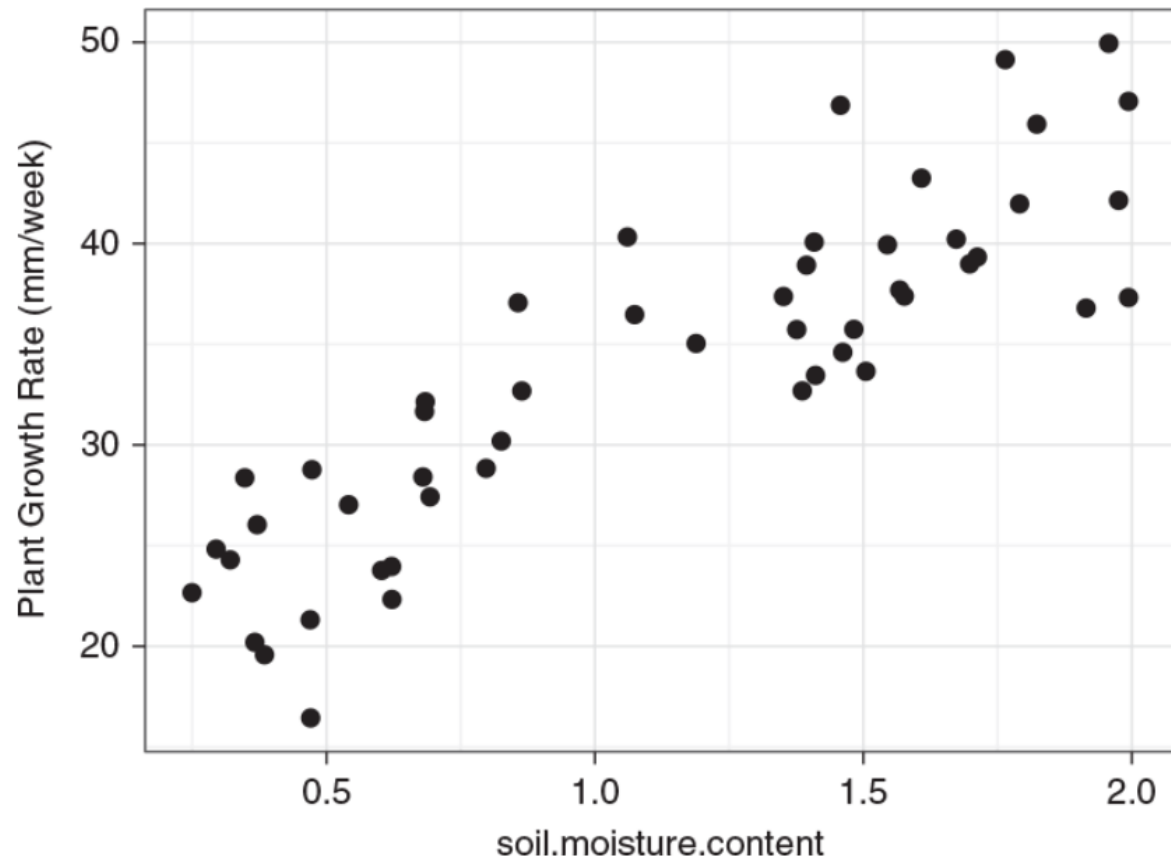
```
## Observations: 50
## Variables: 2
## $ soil.moisture.content (dbl) 0.4696876, 0.5413106, 1.6...
## $ plant.growth.rate (dbl) 21.31695, 27.03072, 38.98...
```

```
> plant_gr <- read_csv("plant.growth.rate.csv")
Parsed with column specification:
cols(
  soil.moisture.content = col_double(),
  plant.growth.rate = col_double()
)
> tbl_df(plant_gr)
# A tibble: 50 x 2
  soil.moisture.conte~ plant.growth.ra~
                <dbl>                <dbl>
1                0.470                 21.3
2                0.541                 27.0
3                1.70                  39.0
4                0.826                 30.2
5                0.857                 37.1
6                1.61                  43.2
7                0.250                 22.7
8                1.67                  40.2
9                1.46                  46.9
10               0.473                 28.8
# ... with 40 more rows
```

Linear Regression

II. Visualization

```
ggplot(plant_gr,  
  aes(x = soil.moisture.content, y = plant.growth.rate)) +  
  geom_point() +  
  ylab("Plant Growth Rate (mm/week)") +  
  theme_bw()
```



Linear Regression

III. Run linear model

```
model_pgr <- lm(plant.growth.rate ~ soil.moisture.content,  
               data = plant_gr)
```

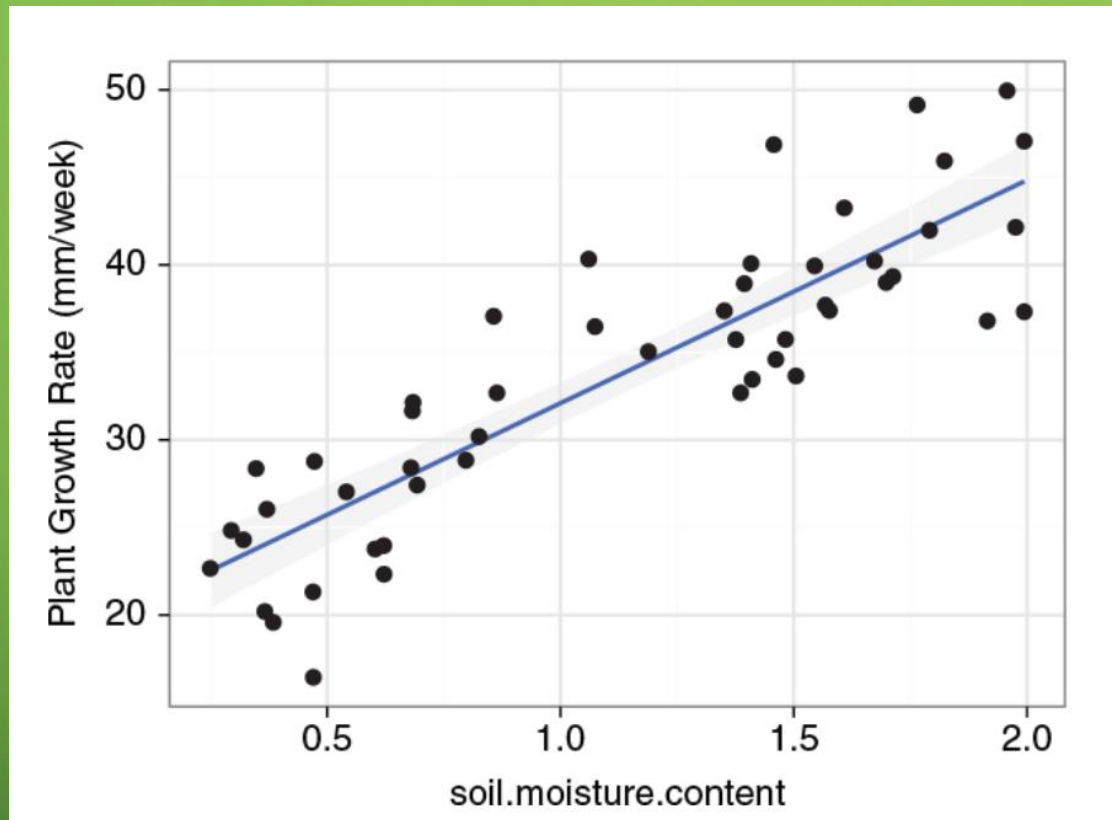
```
summary(model_pgr)  
  
##  
## Call:  
## lm(formula = plant.growth.rate ~ soil.moisture.content,  
      data = plant_gr)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -8.9089 -3.0747  0.2261  2.6567  8.9406   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)      19.348      1.283   15.08 <2e-16     
## soil.moisture.content  12.750      1.021   12.49 <2e-16     
##  
## (Intercept)          ***  
## soil.moisture.content ***  
## ---  
## Signif. codes:  
## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 4.019 on 48 degrees of freedom  
## Multiple R-squared:  0.7648, Adjusted R-squared:  0.7599  
## F-statistic: 156.1 on 1 and 48 DF, p-value: < 2.2e-16
```

Conclusions:

- The null hypothesis (no correlation) is rejected at $p < 2.2e-16$
- The plant growth rate is significantly correlated with soil moisture with $R^2 = 0.7599$

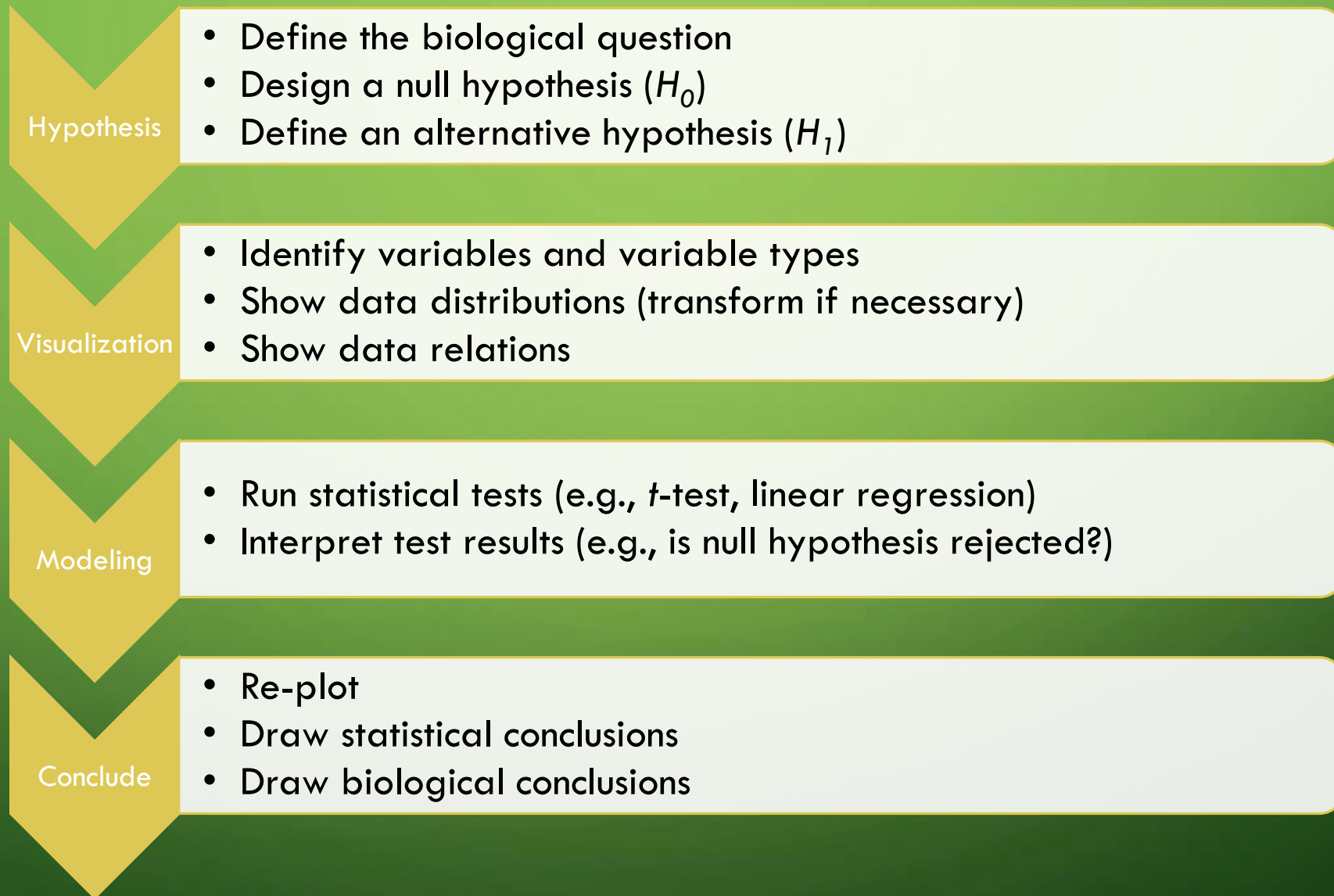
Linear Regression

IV. Re-plot (add regression line & confidence band)



```
ggplot(plant_gr, aes(x = soil.moisture.content,  
  y = plant.growth.rate)) +  
  geom_point() +  
  geom_smooth(method = 'lm') +  
  ylab("Plant Growth Rate (mm/week)") +  
  theme_bw()
```

Data Analysis Workflow



PRACTICE #4

- Does the “Sepal.Length” differ between the two species “virginica” & “vesicolor”?
Perform a *t*-test and include all 4 steps
- How about the “Sepal Width”? Perform a *t*-test and include all 4 steps
- Are the “Sepal.Width” and “Sepal.Length” correlated in the species “setosa”? Show all 4 steps.
- How about in the other two species?
- Batch testing the above correlation in all 3 species at once
- Save all commands to a file “**practice-4.R**”