## Rainfall Example

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## Objectives

Here, we work through the rainfall analysis

## Load in Data

## library (Sleuth3) <br> library (ggplot2) <br> data("case0301")

## Rainfall EDA

$$
\begin{aligned}
& \text { qplot }(\mathrm{x}=\text { Treatment, } \mathrm{y}=\text { Rainfall, } \\
& \text { data }=\text { case0301, geom = "boxplot") }
\end{aligned}
$$



## Rainfall EDA

$$
\begin{aligned}
& \text { qplot }(x=\text { Rainfall, facets }=. \sim \text { Treatment, } \\
& \text { data }=\text { case0301, geom = "histogram", bins }=10)
\end{aligned}
$$



## Rainfall EDA

$$
\begin{aligned}
& \text { qplot (sample = Rainfall, facets }=. \sim \text { Treatment, } \\
& \text { data }=\text { case0301, geom }=\text { "qq") }+ \\
& \text { geom_qq_line() }
\end{aligned}
$$



## Apply Transformation

case0301\$logRainfall <- log(case0301\$Rainfall)

## Rain EDA

> qplot $(\mathrm{x}=$ Treatment, $\mathrm{y}=$ logRainfall, data $=$ case0301, geom = "boxplot")


## Rainfall EDA

$$
\begin{aligned}
& \text { qplot }(x=\text { logRainfall, facets }=. \sim \text { Treatment, } \\
& \text { data }=\text { case0301, geom }=\text { "histogram", bins }=10)
\end{aligned}
$$



## Rainfall EDA

```
qplot(sample = logRainfall, facets = . ~ Treatment,
    data = case0301, geom = "qq") +
    geom_qq_line()
```



## Interpretation

## Posit a Model

- $Z_{i}=$ rainfall on unseeded days.
- $Y_{i}=\mathbf{l o g}$ rainfall on unseeded days.
- $Z_{i}^{*}=$ rainfall on seeded days.
- $Y_{i}^{*}=\log$ rainfall on seeded days.
- $Y_{i}^{*}=Y_{i}+\delta$.
- $Z_{i}^{*}=e^{\delta} Z_{i}$.
- $e^{\delta}$ is the multiplicative effect of seeding on rainfall.
- $e^{\delta}=2$ means rainfall is twice as large on seeded days.
- $e^{\delta}=3$ means rainfall is three times as large on seeded days.


## Posit Hypotheses

- $H_{0}: \delta=0$
- $H_{a}: \delta \neq 0$


## Run t-test

```
tout <- t.test(logRainfall ~ Treatment, data = case0301)
tout
```

\#\#
\#\# Welch Two Sample t-test
\#\#
\#\# data: logRainfall by Treatment
\#\# t $=2.5, \mathrm{df}=50, \mathrm{p}$-value $=0.01$
\#\# alternative hypothesis: true difference in means is not
\#\# 95 percent confidence interval:
\#\# 0.24082 .0467
\#\# sample estimates:
\#\# mean in group Seeded mean in group Unseeded
\#\#
5.134
3.990

## Estimate and Confidence Intervals on Original Scale

$\exp ($ tout\$estimate[1] - tout\$estimate[2])
\#\# mean in group Seeded
\#\# 3.139
exp(tout\$conf.int)
\#\# [1] 1.2727 .742
\#\# attr(."conf.level")
\#\# [1] 0.95

## Conclusion

- We estimate that that seeding results in a 3.1 factor increase in rainfall ( $p$-value $0.01,95 \%$ confidence interval of 1.3 to 7.7 ).
- Note the causal language because this is a randomized experiment. I will deduct many points if you use causal language in an observational study.

