

# Data Visualization With R

Zack Treisman

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# Philosophy

Exploration<sup>1</sup> ← **Visualizations** → Publication

## Graphics systems in R

- ▶ Base R: intuitive, perhaps limited.
- ▶ `ggplot2`: robust and widely used.
- ▶ `lattice`: also nice, often older.

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<sup>1</sup>Be wary of inference based on purely exploratory data analysis. If you look at your data until you find a pattern, and then test for that pattern, the significance levels of that test are inflated.

# Loading data

Step 0 of visualizing your data with R is loading it.

- ▶ Clean your data spreadsheet:
  - ▶ Remove non-data (summaries, etc.)
  - ▶ Fix typos
  - ▶ Make good variable names
    - ▶ meaningful
    - ▶ not too long
    - ▶ no spaces - use under\_score or camelCaps instead
    - ▶ don't start with a number
  - ▶ More good advice: [Data Carpentry](#)
- ▶ csv (comma separated variable)
- ▶ working directory
  - ▶ possibly *data* subdirectory.
- ▶ `read.csv` or `read_csv`.

## Check the data loaded correctly

- ▶ `str()` Are variables coded correctly? (factors, dates)
- ▶ `head()` or `View()`

```
str(ReedfrogPred)
```

```
## 'data.frame': 48 obs. of 5 variables:  
## $ density : int 10 10 10 10 10 10 10 10 10 10 ...  
## $ pred    : Factor w/ 2 levels "no","pred": 1 1 1 1 1 1 1 1 1 2 2 ...  
## $ size    : Factor w/ 2 levels "small","big": 2 2 2 2 1 1 1 1 1 2 2 ...  
## $ surv    : num 9 10 7 10 9 9 10 9 4 9 ...  
## $ propsurv: num 0.9 1 0.7 1 0.9 0.9 1 0.9 0.4 0.9 ...
```

```
head(SeedPred)
```

	station	dist	species	date	seeds	tcum	tint	taken	available
## 1	1	10	psd	1999-03-25	5	0	NA	NA	NA
## 2	1	10	psd	1999-03-28	5	3	3	0	5
## 3	1	10	psd	1999-04-04	5	10	7	0	5
## 4	1	10	psd	1999-04-11	5	17	7	0	5
## 5	1	10	psd	1999-04-18	0	24	7	5	5
## 6	1	10	psd	1999-04-25	0	31	7	0	0

*# data from R package emdboook*

# Exploration

Data are in R, now what?

- ▶ Check numerical summaries.

```
summary(ReedfrogPred)
```

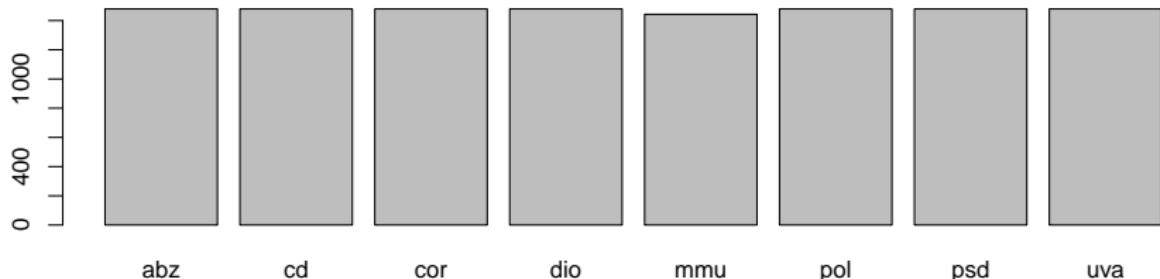
```
##      density      pred      size      surv      propsurv
##  Min.   :10.00  no    :24  small:24  Min.   : 4.00  Min.   :0.1143
##  1st Qu.:10.00  pred:24  big   :24  1st Qu.: 9.00  1st Qu.:0.4964
##  Median :25.00                               Median :12.50  Median :0.8857
##  Mean   :23.33                               Mean   :16.31  Mean   :0.7216
##  3rd Qu.:35.00                               3rd Qu.:23.00  3rd Qu.:0.9200
##  Max.   :35.00                               Max.   :35.00  Max.   :1.0000
```

- ▶ Make some graphics!
  - ▶ patterns; expected/unexpected?
  - ▶ Data issues?

# Barplots and Histograms - one variable

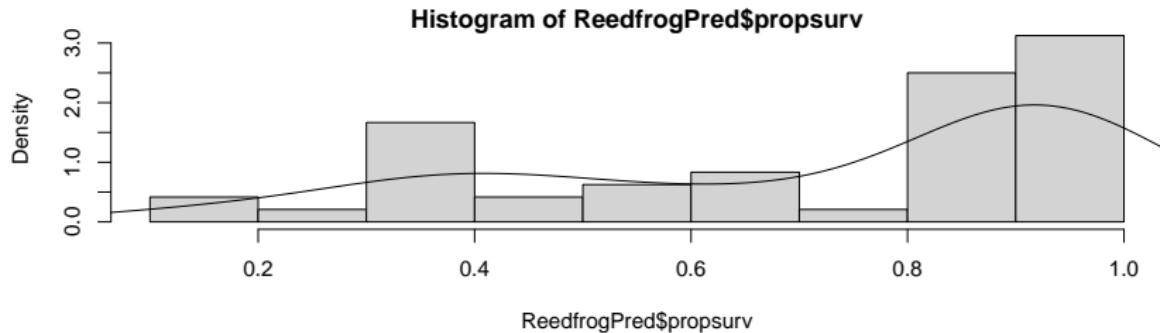
- ▶ Barplots  $\leftrightarrow$  categorical variables.

```
barplot(table(SeedPred$species))
```



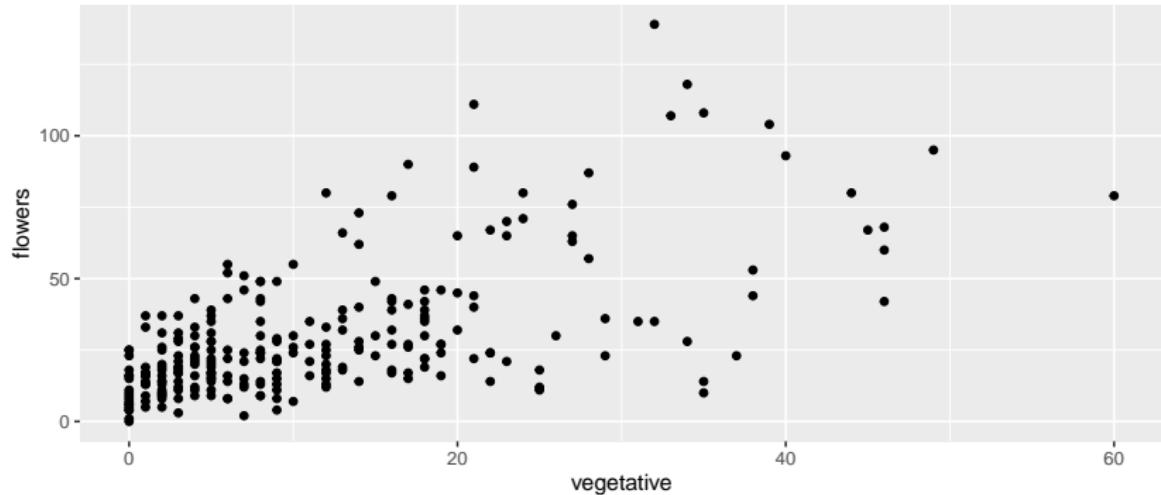
- ▶ Histograms, density estimates  $\leftrightarrow$  numeric variables.

```
hist(ReedfrogPred$propsurv, freq=F); lines(density(ReedfrogPred$propsurv))
```



## Scatterplots - two numeric variables

```
ggplot(Lily_sum, aes(vegetative, flowers)) + geom_point()
```

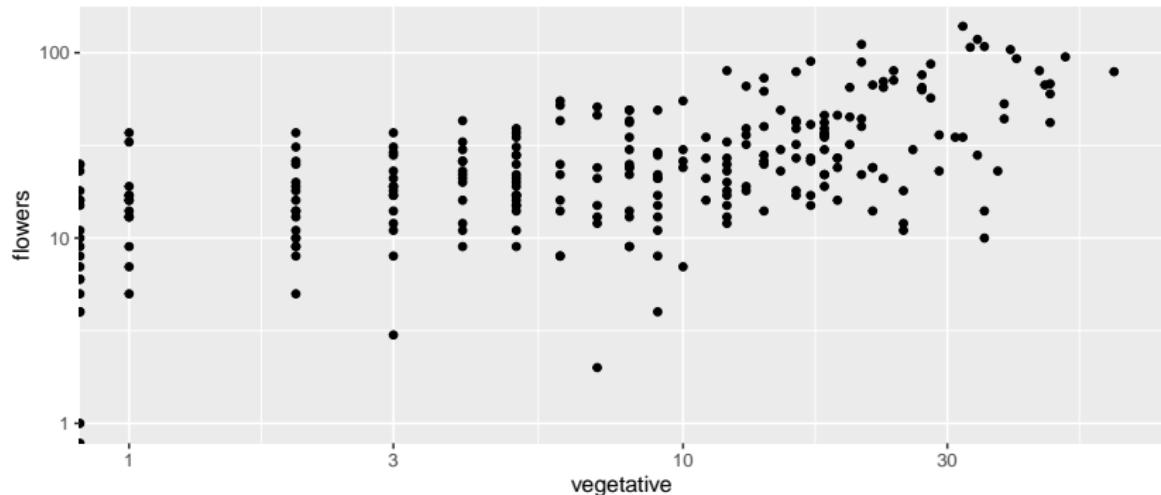


```
# data in emdbook
```

## Log scales

- ▶ Right skew
- ▶ Counts
- ▶ Dimensional data

```
ggplot(Lily_sum, aes(vegetative, flowers)) +  
  geom_point() +  
  scale_x_log10() +  
  scale_y_log10()
```

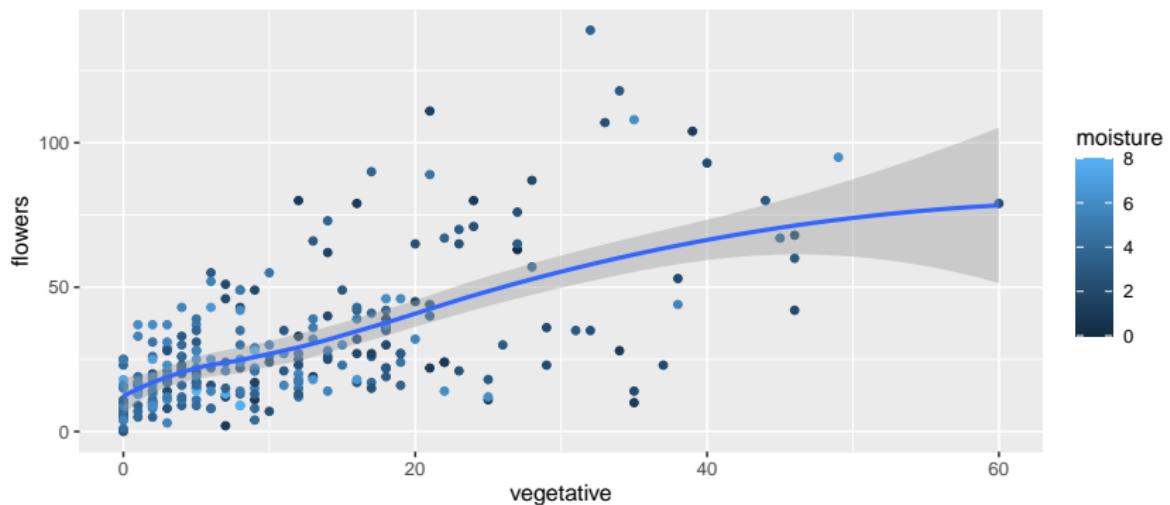


$\log(0)$  is undefined

## Additional aesthetics

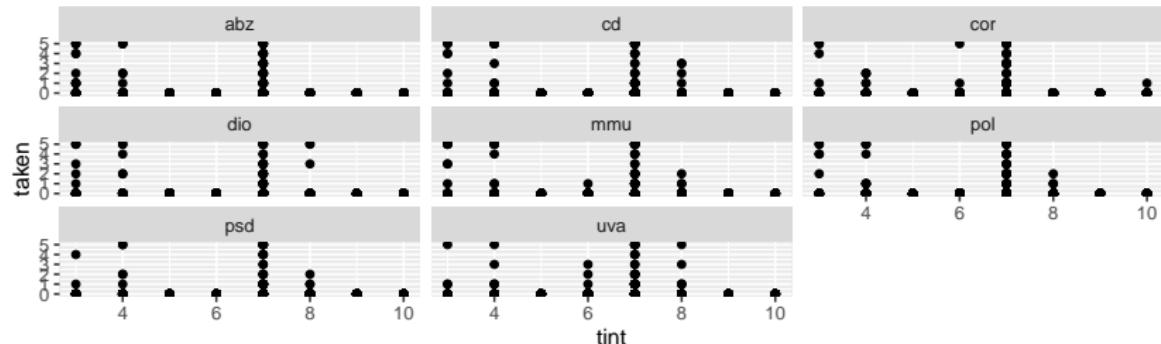
- ▶ color
- ▶ shape (categorical)
- ▶ size (numeric)
- ▶ trendlines or other model graphs.

```
ggplot(Lily_sum, aes(vegetative, flowers, color = moisture)) +  
  geom_point() +  
  geom_smooth()
```

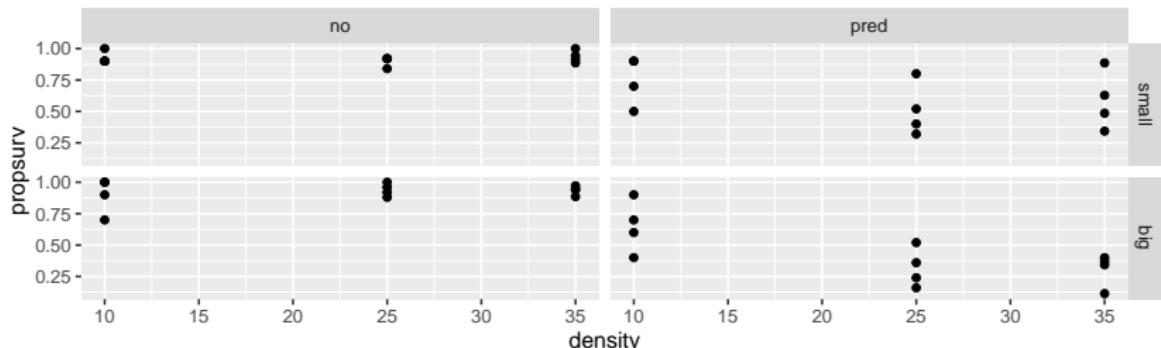


## Categorical variables → facets

```
ggplot(SeedPred, aes(tint, taken)) + geom_point() + facet_wrap(~species)
```

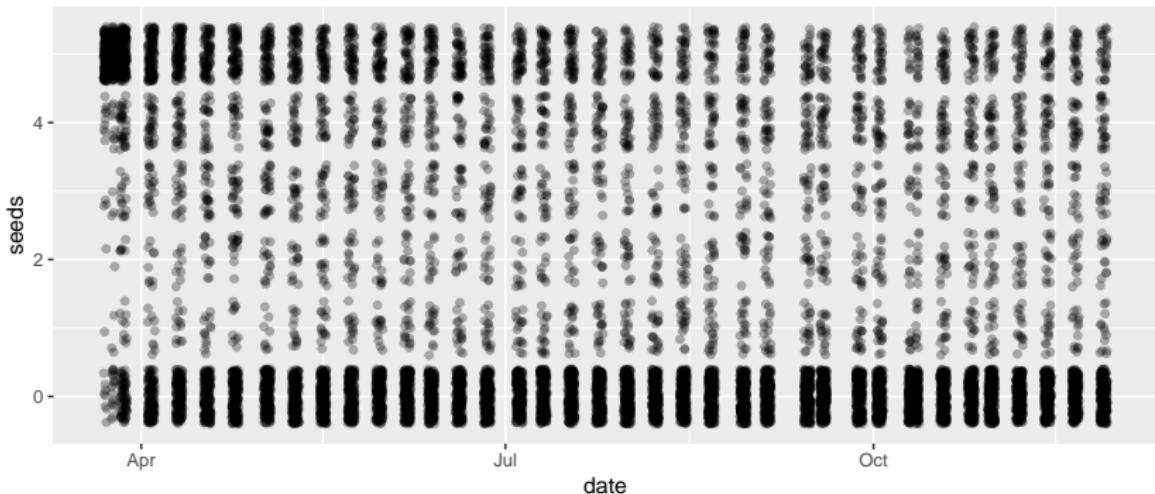


```
ggplot(ReedfrogPred, aes(density, propsurv)) + geom_point() + facet_grid(size~pred)
```



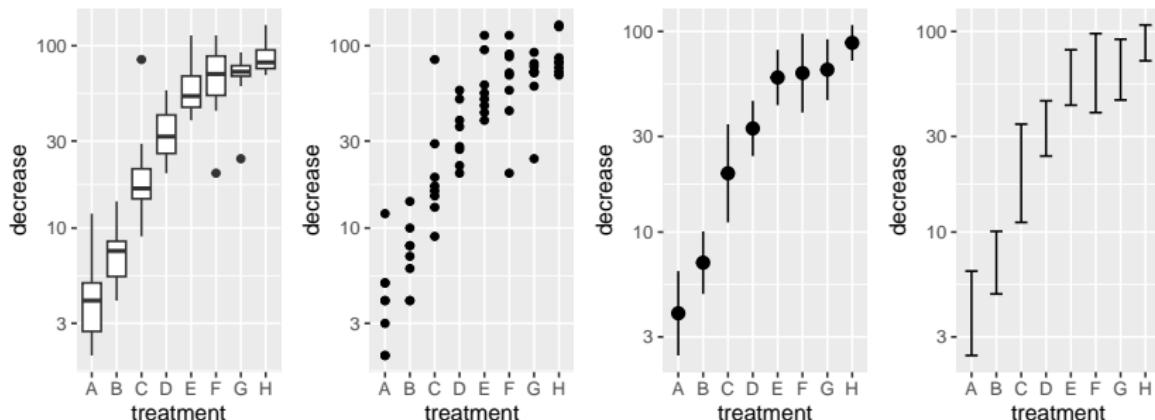
# Jittering and transparency

```
ggplot(SeedPred, aes(date, seeds))+
  geom_jitter(alpha= 0.3)
```



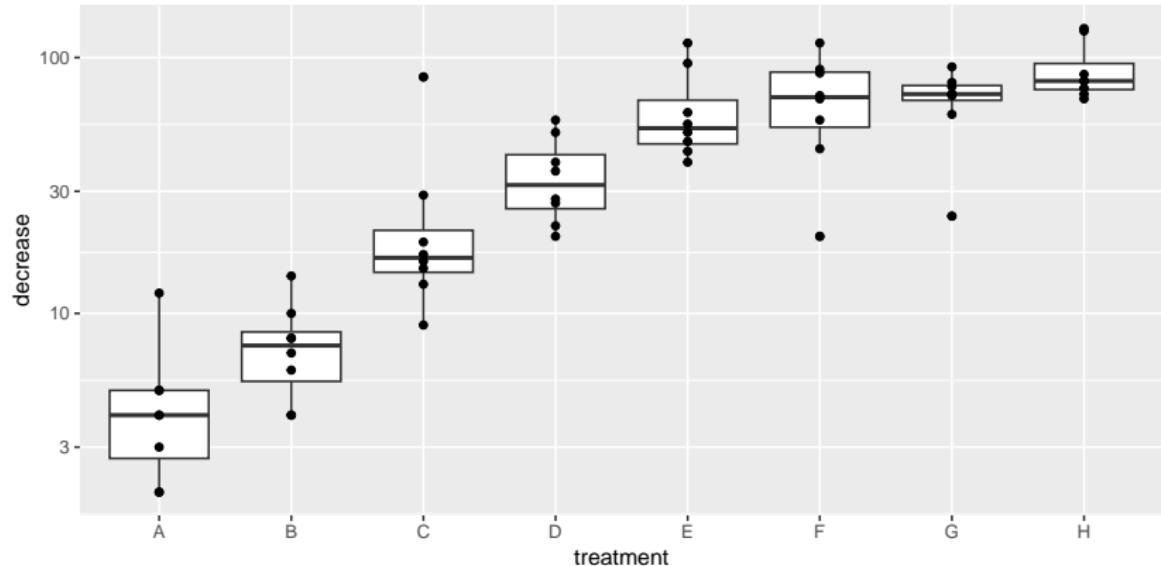
## Boxplots - numeric response, categorical predictors

```
g0 <- ggplot(OrchardSprays,aes(x=treatment,y=decrease)) + # data in MASS
  scale_y_log10()
g_boxplot <- g0 + geom_boxplot()
g_point <- g0 + geom_point()
g_ptrng <- g0 + stat_summary(fun.data=mean_cl_normal,geom="pointrange")
g_errbar <- g0 +
  stat_summary(fun.data=mean_cl_normal,geom="errorbar",width=0.5)
grid.arrange(g_boxplot,g_point,g_ptrng,g_errbar, nrow=1)
```



# Combining layers

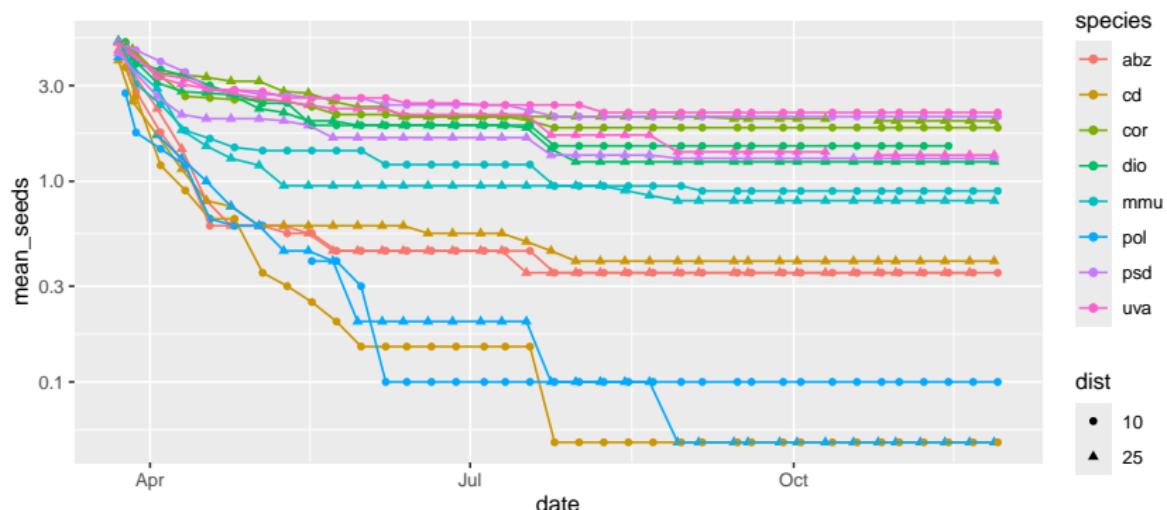
```
g0 + geom_boxplot() + geom_point()
```



# Reshape and Summarize

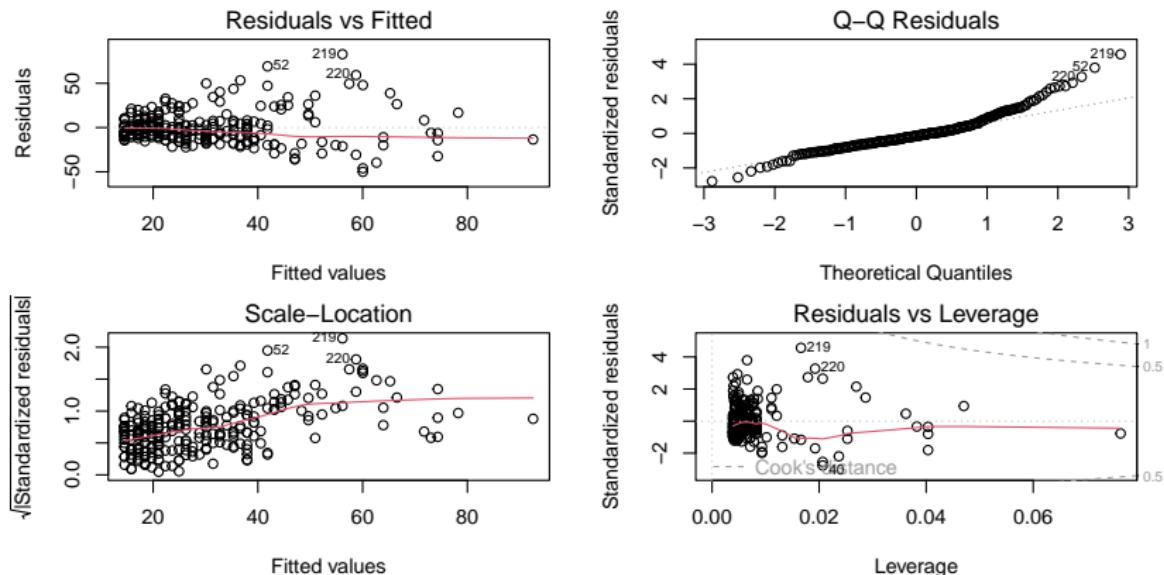
Figure 2.1 from Bolker (2008):

```
daily_avgs <- SeedPred %>%
  group_by(date, species, dist) %>%
  summarise(mean_seeds = mean(seeds))
ggplot(daily_avgs, aes(date, mean_seeds, color=species, shape=dist)) +
  geom_point() + geom_line() + scale_y_log10()
```



# Diagnostics, assessment of model validity.

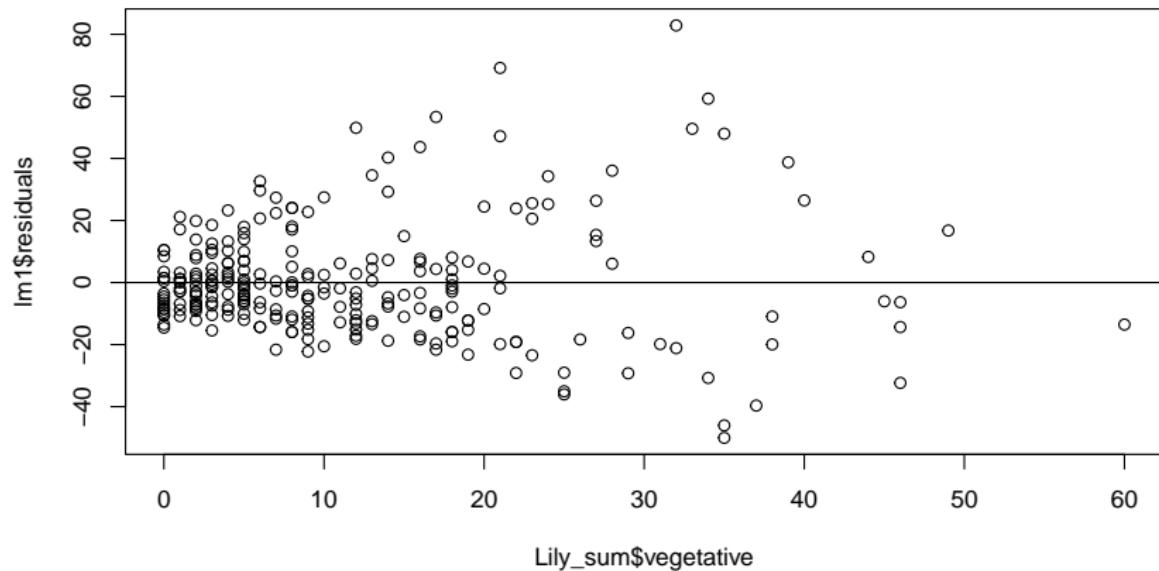
```
lm1 <- lm(flowers~vegetative, data = Lily_sum)  
par(mfrow=c(2, 2), mar = c(4, 4, 2, 2)) # see all 4 plots at once  
plot(lm1)
```



```
par(mfrow=c(1, 1), mar = c(4, 4, 0.75, 0.5)) # restore graphics parameters
```

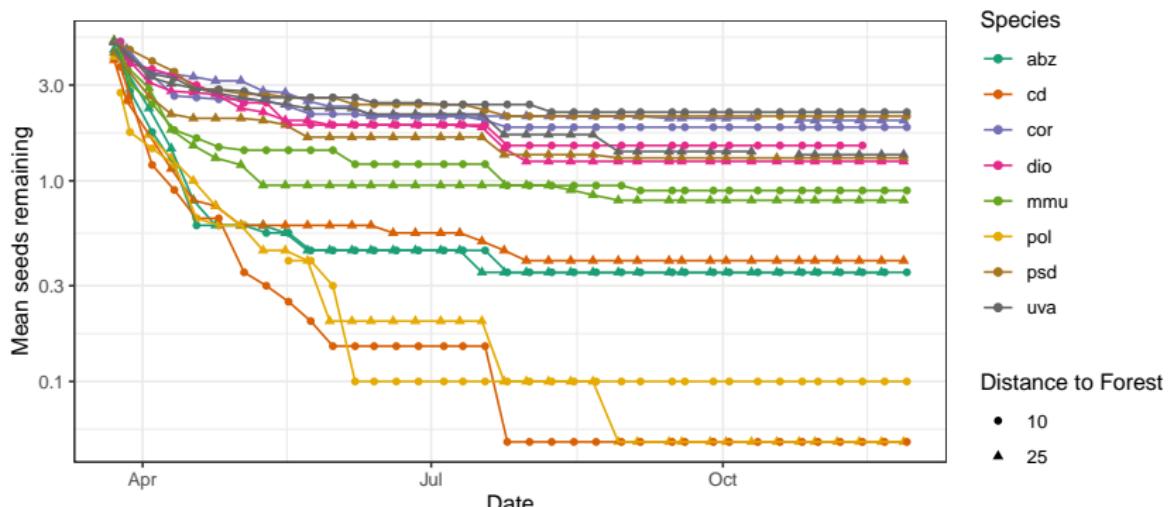
## Residuals v. predictors

```
plot(lm1$residuals~Lily_sum$vegetative)
abline(h=0)
```



# Fine tune and save graphics for presentation

```
emd2.1<-ggplot(daily_avgs,aes(date,mean_seeds,color=species,shape=dist))+  
  geom_point() + geom_line() + scale_y_log10() +  
  labs(y="Mean seeds remaining", x = "Date",  
       color = "Species", shape = "Distance to Forest") +  
  scale_color_brewer(palette = "Dark2") +  
  theme_bw()  
emd2.1
```



```
ggsave("figures/BolkerFig2.1.tiff", plot=emd2.1,  
       width = 10, height = 4, units = "cm", dpi = 800)
```

## Opinions on graphical style

Plenty of people with good ideas about style.

- ▶ Leland Wilkinson
- ▶ Edward Tufte
- ▶ William Cleveland
- ▶ Andrew Gelman

Some graph types are controversial. That doesn't mean never use them, but if you do, be aware of the criticisms.

- ▶ Pie charts, dynamite plots, dual-axes plots

## References

Bolker, Benjamin M. 2008. *Ecological Models and Data in R*. Princeton University Press.