

GGPLOT2 COURSE, AUFLAND CONFERENCE

Eduard Szöcs

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Institute for Environmental Sciences - University of Koblenz-Landau



INTRO

- PhD student @Quantitative Landscape Ecology
- Environmental Sciences & Ecotoxicology
- Research:
 - Statistical Ecology - Eco(toxico)logical Statistics
 - Effects and distribution of pesticides in freshwaters
- R-Programming:
 - R-user for 6years
 - Author/Co-Author of 3 CRAN packages (taxize, webchem, rspear)
 - Other packages on github (restax, esmisc)
 - Minor contributions to other pkgs (e.g. vegan)

edild.github.io

@EduardSzoebs

- Diploma in Environmental Sciences
- Diploma Thesis @Quantitative Landscape Ecology
- Topics: Ecology, Renewable Energies and Conservation
- Graduate Assistant @Environmental Sciences Lab: Renewable Energies
 - Automated display of Spatial Information on Renewable Energies in a WebGIS
- R-Programming:
 - R-User for 5 Years
 - Assistant in several Summerschools (@UFZ-Leipzig, @Landau)

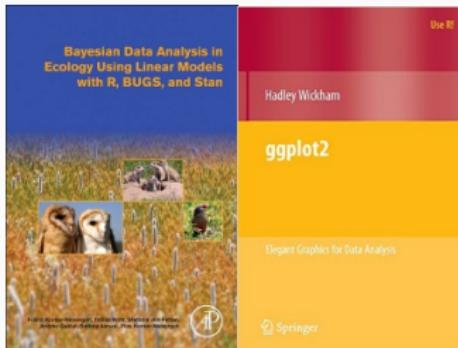
- Short intro & course organisation, Software preparation
- An introduction to ggplot2.
 - Basic ggplot2 usage
 - Customization of your plots
 - Free time for specific Q.
- After 4 hours of ggplot2 you should be an expert ;)

Course material: https://github.com/EDiLD/r_landau_2015

- Download the course repository.
- No formal R knowledge required to follow.
- (Install RStudio and packages needed).
- Just open the '.R' files in RStudio and execute the script line by line:
‘CTRL + ENTER’

We will use 2 data sets in this course:

1. Frog monitoring in Swiss (Slides).
2. Diamond prices (Exercises).



Tip:

Don't buy the old ggplot2 book - a new one is coming soon...

FROGS ABUNDANCE

```
library(blmeco)
data(frogs)
head(frogs)

##   count1 count2 elevation year fish waterarea vegetation pondid      x
## 1     16     12       380 2013     0     2500                 1 400301 649750
## 2     0      0       565 2009     0     300                  1 400411 647350
## 3     0      0       430 2012     0     450                  1 400603 650250
## 4     0      0       500 2012     0     348                  1 400608 649400
## 5     0      0       450 2012     0     200                  1 400701 646700
## 6     0      0       560 2010     0      42                  1 400802 646500
##
##           y
## 1 248850
## 2 255750
## 3 244600
## 4 243850
## 5 240750
## 6 253650
```

```
frogs$fish <- factor(frogs$fish)
frogs$vegetation <- factor(frogs$vegetation)
```

Questions:

What influences frog abundance? Can we predict frog abundance (ala SDM)?

DIAMONDS DATASET

```
library(ggplot2)
data(diamonds)
head(diamonds)

##   carat      cut color clarity depth table price     x     y     z
## 1 0.23    Ideal    E    SI2  61.5    55   326 3.95 3.98 2.43
## 2 0.21  Premium    E    SI1  59.8    61   326 3.89 3.84 2.31
## 3 0.23      Good    E    VS1  56.9    65   327 4.05 4.07 2.31
## 4 0.29  Premium    I    VS2  62.4    58   334 4.20 4.23 2.63
## 5 0.31      Good    J    SI2  63.3    58   335 4.34 4.35 2.75
## 6 0.24 Very Good    J   VVS2  62.8    57   336 3.94 3.96 2.48

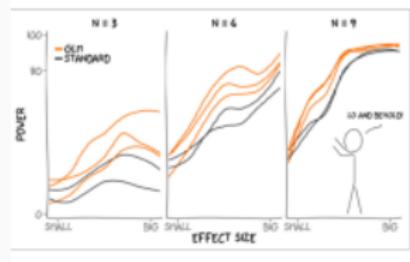
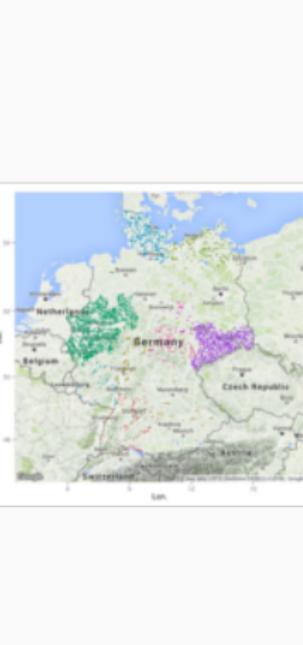
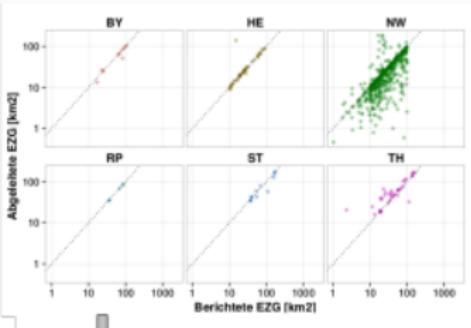
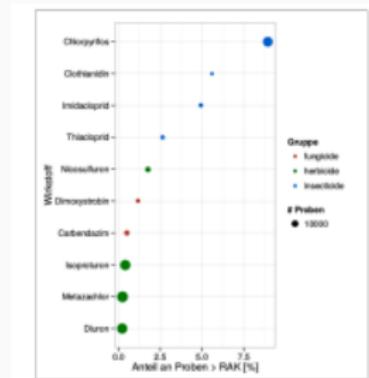
?diamonds
```

Questions:

What factors are crucial for the diamond prices?

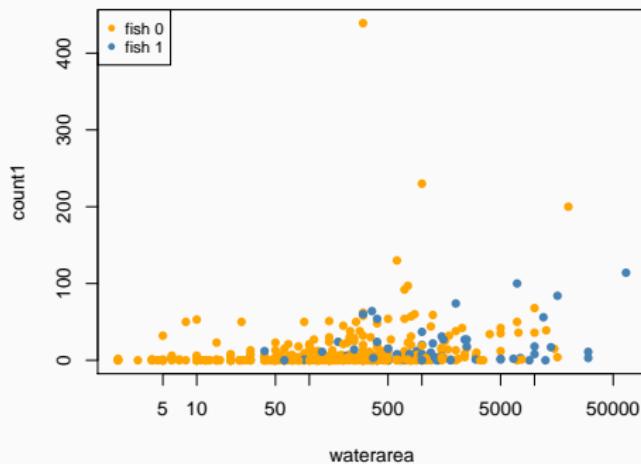
GGPLOT2 BASICS

RECENT PLOTS I CREATED WITH GGPLOT2



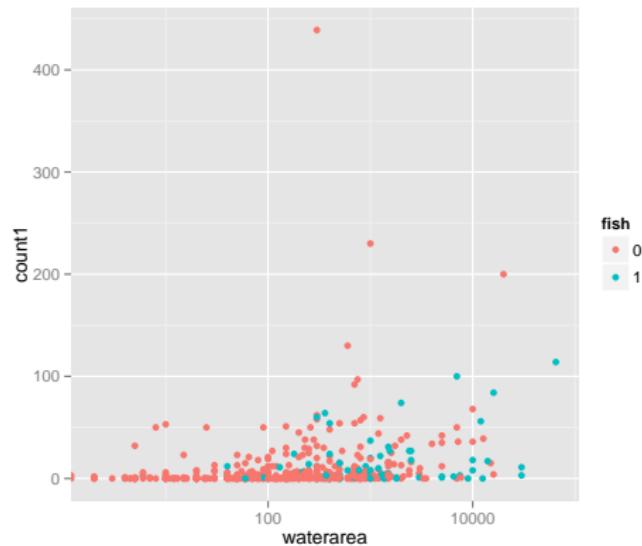
WITH BASE R

```
cols <- c('orange', 'steelblue')
plot(count1 ~ waterarea, data = frogs, type = 'n', log = 'x')
with(frogs, points(waterarea, count1, col = cols[fish], pch = 16))
legend('topleft', legend = c('fish 0', 'fish 1'), pch = 16,
       col = cols,
       cex = 0.8)
```



WITH GGPLOT2

```
ggplot(frogs) +  
  geom_point(aes(x = waterarea, y = count1, col = fish)) +  
  scale_x_log10()
```



WHY GGPLOT2?

base graphics

- more work / code
- legends, colors?
- defaults ok
- multipanel plots?
- can do anything
- have to know the *tricks*

ggplot2

- quick-and-dirty and complex
- Automatic legends, colors!
- nice defaults
- easy multipanel plots
- restrictions (e.g. 2nd y-axis)
- intuitive syntax

WHY GGPLOT2?

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ggplot2

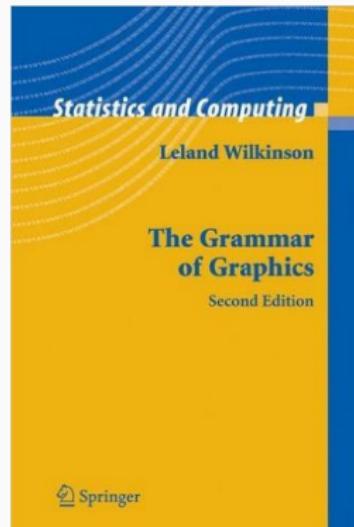
- quick-and-dirty and complex
- Automatic legends, colors!
- nice defaults
- easy multipanel plots
- restrictions (e.g. 2nd y-axis)
- intuitive syntax

Tip:

If a plot is too complicated to draw with ggplot2, reconsider if it's a good representation of your data!

WHAT IS GGPLOT2?

- A graphic system for R
- gg = Grammar of Graphics (Wilkinson 2005)
- *Grammar*: A set of components that define a sentence
- `ggplot` defines a grammar to create plots
- Consistent, intuitive, easy to learn



THE GRAMMAR

```
ggplot(frogs) +  
  geom_point(  
    aes(x = waterarea, y = count1, col = fish)) +  
    scale_x_log10() +  
    theme_bw()
```

ggplot() The main function. Can specify the data set and variables **globally**.

geom A geometric object: `geom_point`, `geom_line`,
`geom_text`, `geom_violin`, ...

aes aesthetics: maps a variable to the properties of a geom:
`shape`, `color`, `fill`, `linetype`, transparency (`alpha`), ...

scale How are data mapped visually (`log`, `date`, `colors`, `sizes`,
...)

theme The general plot appearance (`background`, `axes`, `labels`, ...)

THE GRAMMAR

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...)

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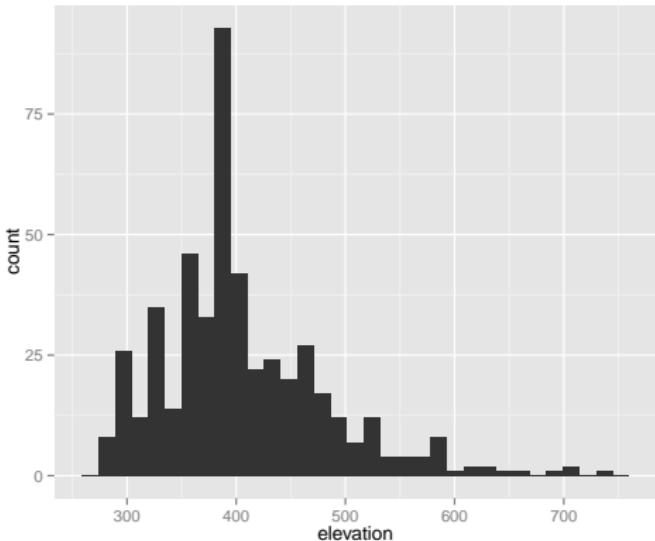
Tip:

Never use `qplot()` (=quick plot)! - You won't learn the grammar...

UNIVARIATE DATA

GEOM: HISTOGRAM

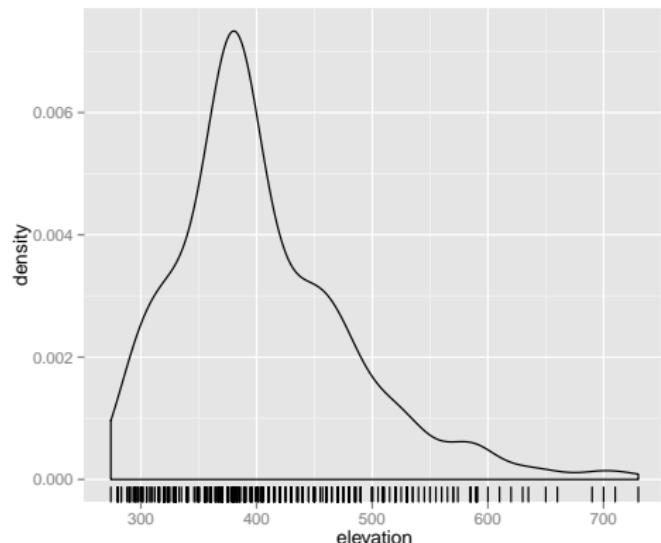
```
ggplot(frogs,           # use the frogs  
       aes(x = elevation))+ # take the variable 'elevation' from the dataset)  
  geom_histogram()        # display a histogramm  
  
## stat_bin: binwidth defaulted to range/30. Use 'binwidth = x' to adjust  
this.
```



GEOM: DENSITY + RUG

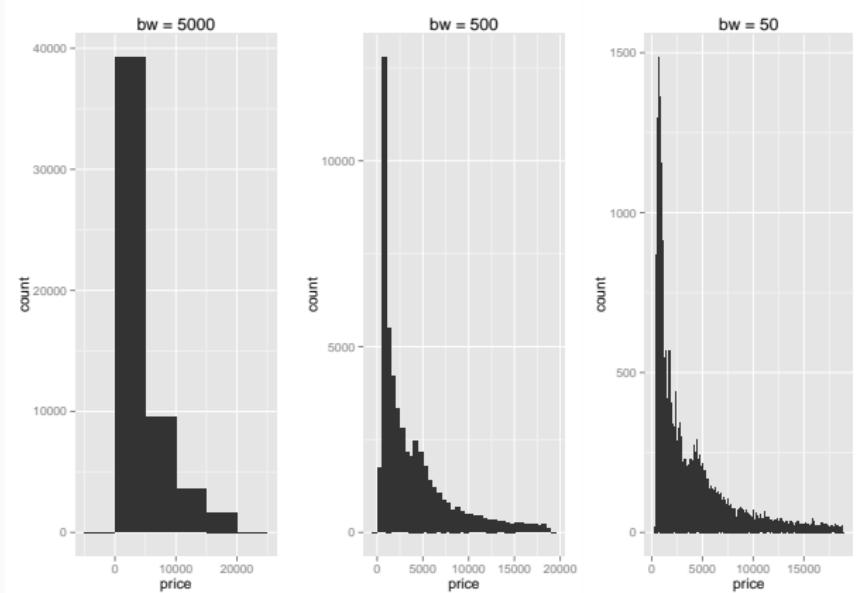
We can add multiple geoms of the same variable to the plot

```
ggplot(frogs, aes(x = elevation)) +      # plot the 'elevation' from the frogs data
  geom_density() +                         # display a density
  geom_rug()                                # display a rug
```



EXERCISE 1:
PLOT A HISTOGRAMM OF DIAMOND PRICES AT DIFFERENT BINWIDTHS
(50, 500, 5000).
HOW DOES THIS AFFECT THE PERCEPTION OF THE DATA?

EXERCISE

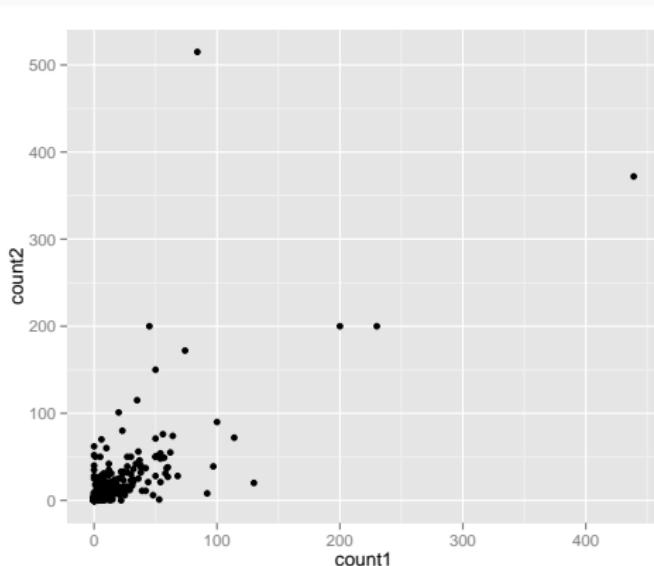


BIVARIATE DATA
(CONT. X CONT.)

GEOM: POINT (=CONT. X CONT.)

The mother of all plots.

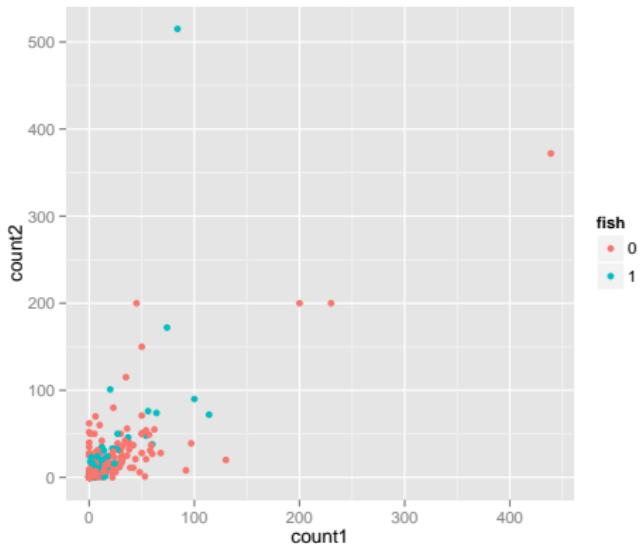
```
ggplot(frogs, aes(x = count1, y = count2)) +  
  geom_point()
```



AESTHETIC: COLOR (DISCRETE)

Differentiate between fish and no-fish:

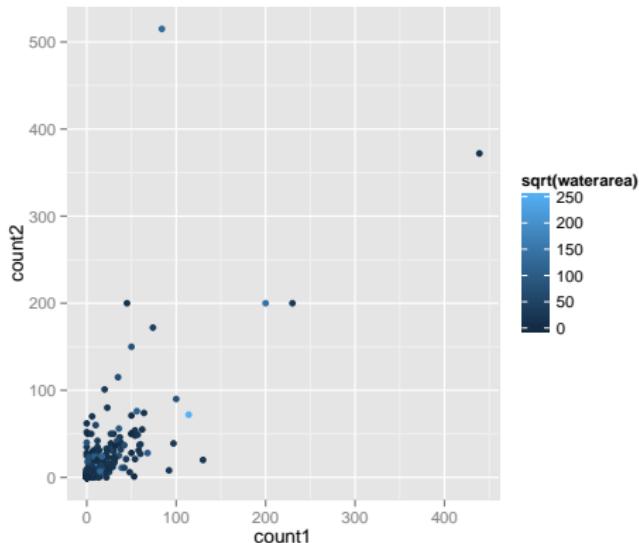
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point()
```



AESTHETIC: COLOR (CONTINUOUS)

Continuous color scale:

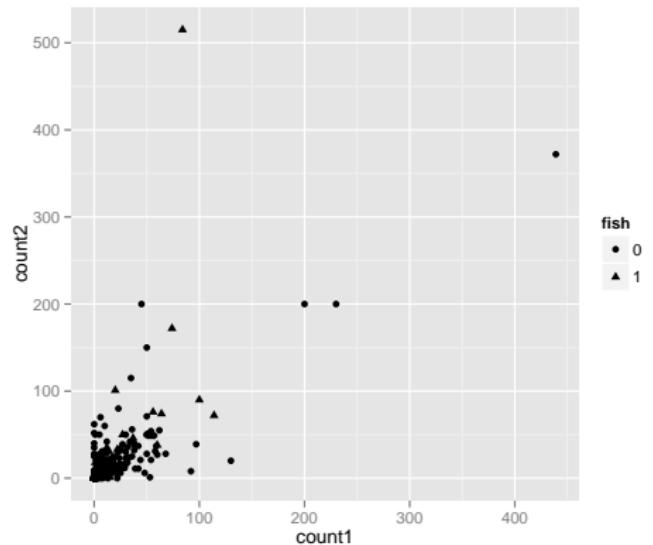
```
ggplot(frogs, aes(x = count1, y = count2, col = sqrt(waterarea))) +  
  geom_point()
```



AESTHETICS: SHAPE

Can also differentiate by shape:

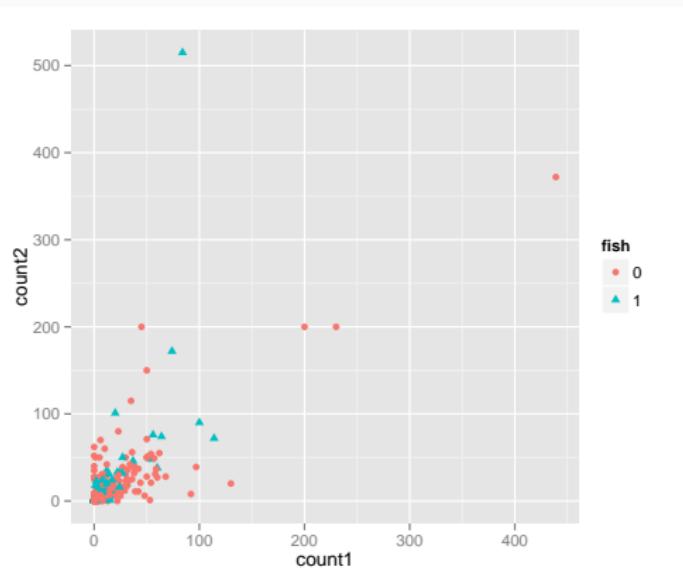
```
ggplot(frogs, aes(x = count1, y = count2, shape = fish)) +  
  geom_point()
```



AESTHETICS: SHAPE + COLOR

Can also differentiate by shape:

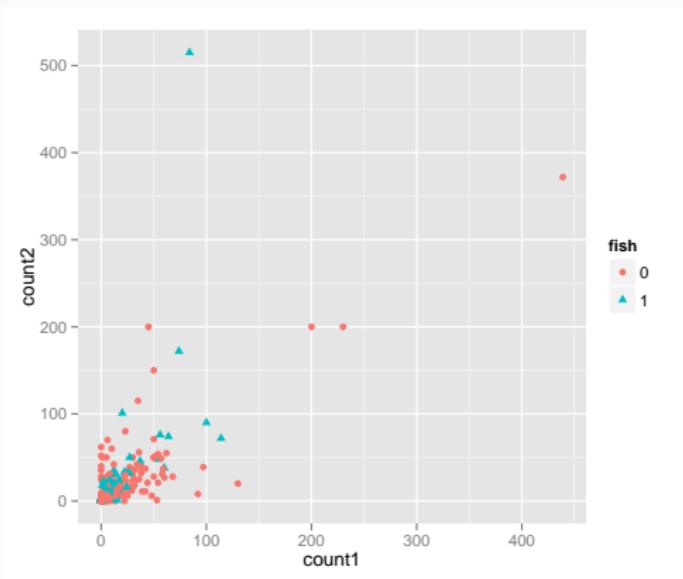
```
ggplot(frogs, aes(x = count1, y = count2, shape = fish, col = fish)) +  
  geom_point()
```



AESTHETICS: SHAPE + COLOR

Can also differentiate by shape:

```
ggplot(frogs, aes(x = count1, y = count2, shape = fish, col = fish)) +  
  geom_point()
```

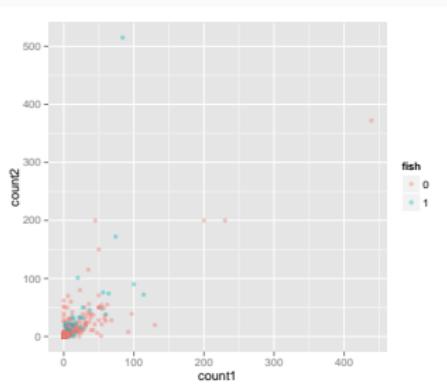


Tip:

Do not use redundant aesthetics.

AESTHETICS: ALPHA (=TRANSPARENCY)

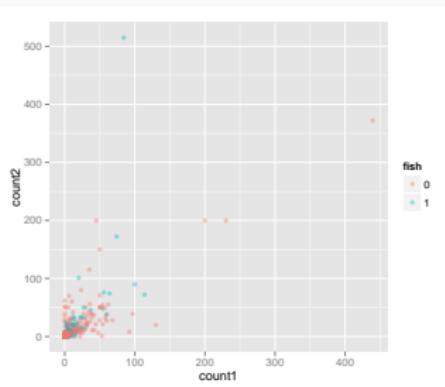
```
ggplot(frogs) +  
  geom_point(aes(x = count1, y = count2, col = fish), alpha = 0.4)
```



Q: alpha is not within aes(), why?

AESTHETICS: ALPHA (=TRANSPARENCY)

```
ggplot(frogs) +  
  geom_point(aes(x = count1, y = count2, col = fish), alpha = 0.4)
```

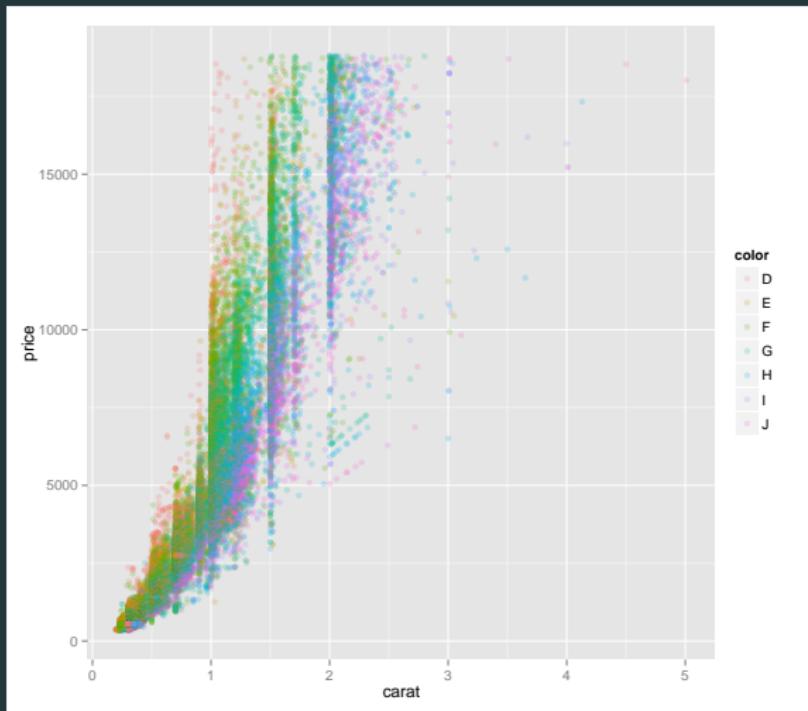


Q: alpha is not within aes(), why?

A:

Within `aes()` the **aesthetic** changes with the specified variable (e.g. `col = fish`). Outside of `aes` we set it to a specific value.

EXERCISE 2: CREATE THIS PLOT:

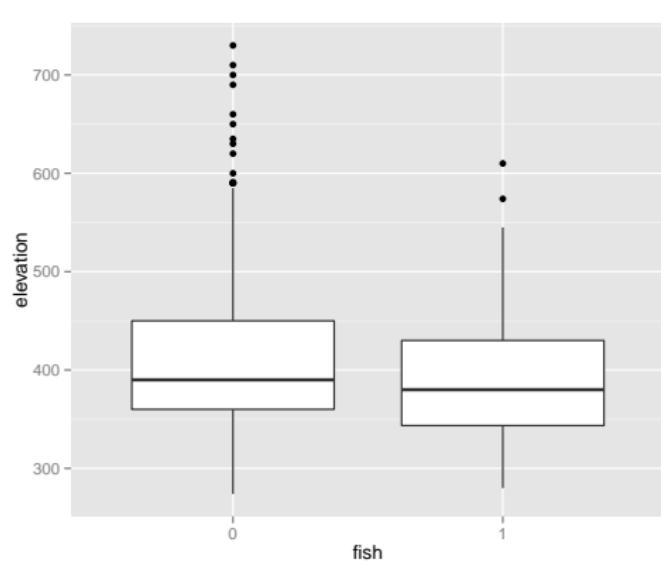


BIVARIATE DATA
(DISCRETE X CONTINUOUS)

GEOM: BOXPLOT

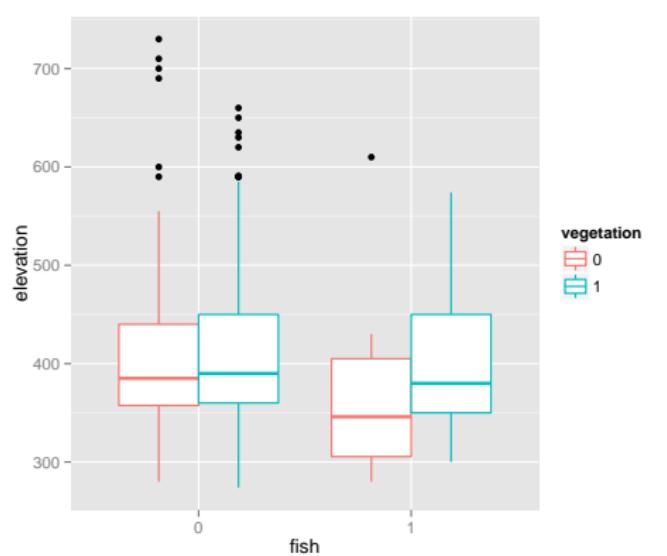
- More than 45 years (Tukey 1970)
- For discrete scales (here fish)

```
ggplot(frogs, aes(x = fish, y = elevation)) +  
  geom_boxplot()
```



AESTHETIC: COLOR (DISCRETE)

```
ggplot(frogs, aes(x = fish, y = elevation, col = vegetation)) +  
  geom_boxplot()
```

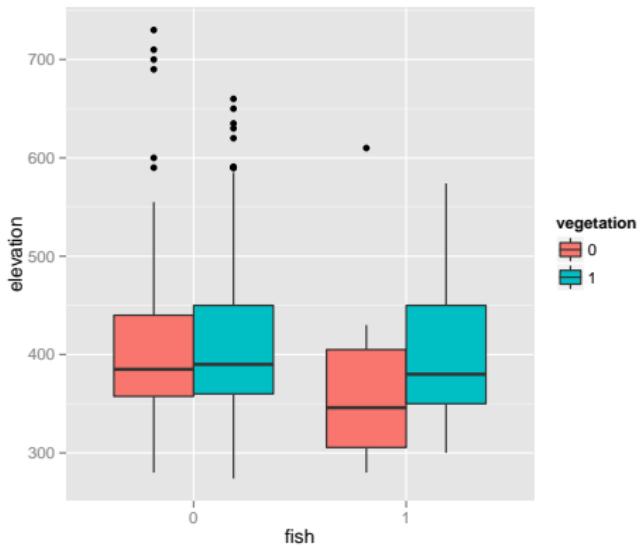


Tip:

Nice for factorial designs (e.g. easily spot interactions).

AESTHETIC: FILL (DISCRETE)

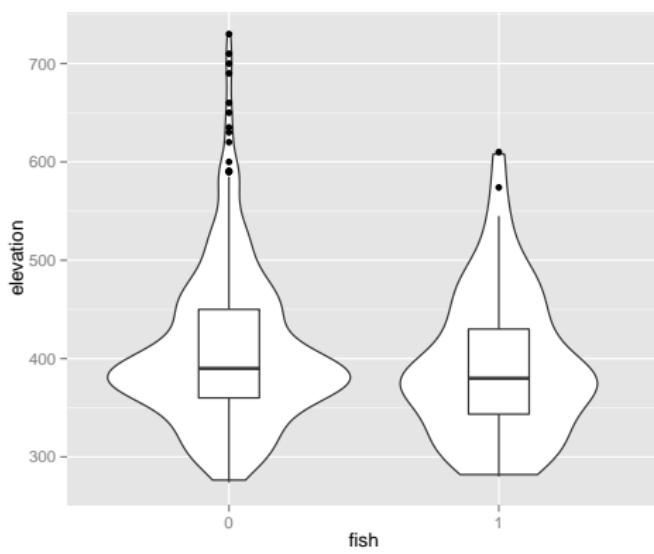
```
ggplot(frogs, aes(x = fish, y = elevation, fill = vegetation)) +  
  geom_boxplot()
```



GEOM: VIOLIN

The Violin plot is a powerful alterative/addition:

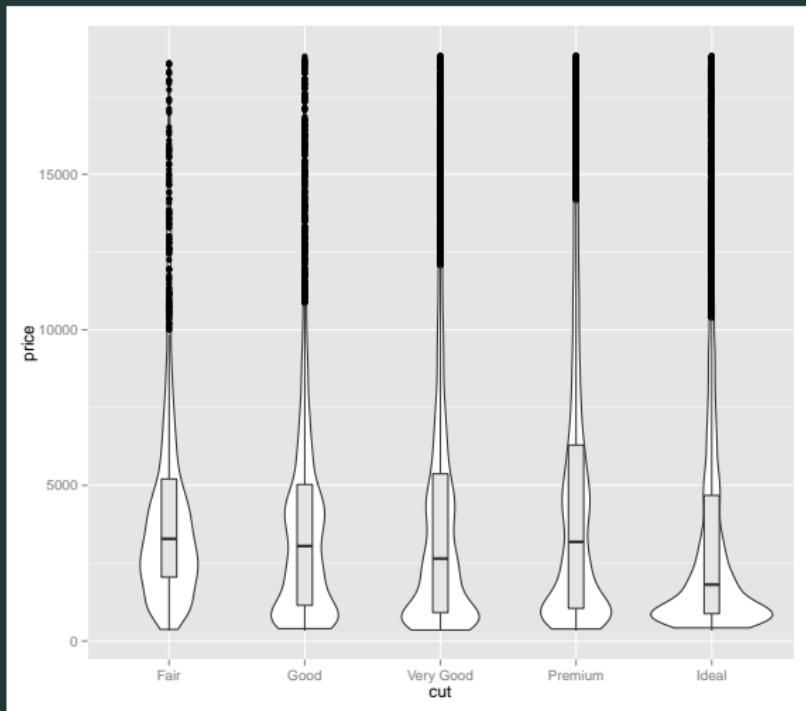
```
ggplot(frogs, aes(x = fish, y = elevation)) +  
  geom_violin() +  
  geom_boxplot(width = 0.3)
```



Tip:

The order of geoms is important! Uppermost layer is the last **geom** specified.

EXERCISE 3: CREATE THIS PLOT:

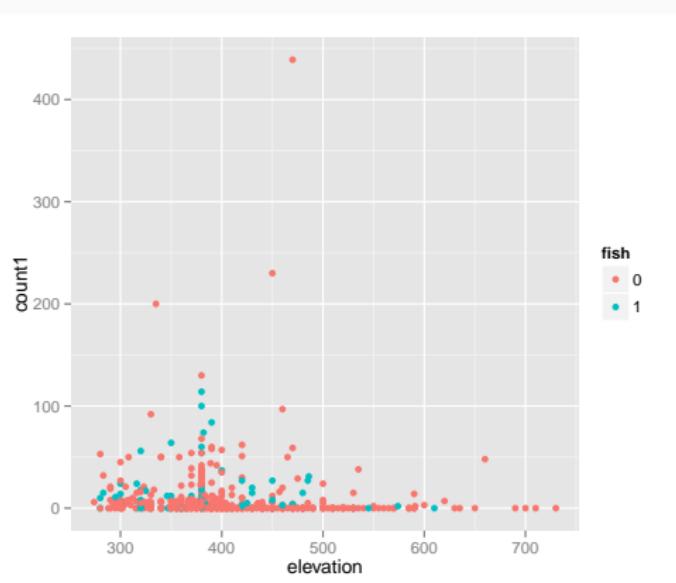


MULTIVARIATE DATA (FACETS)

SUBGROUPS

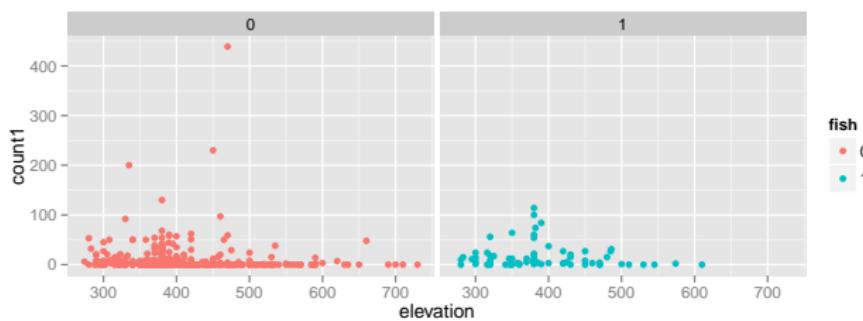
The different relationships between *fish* and *no fish* are hard to spot:

```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point()
```



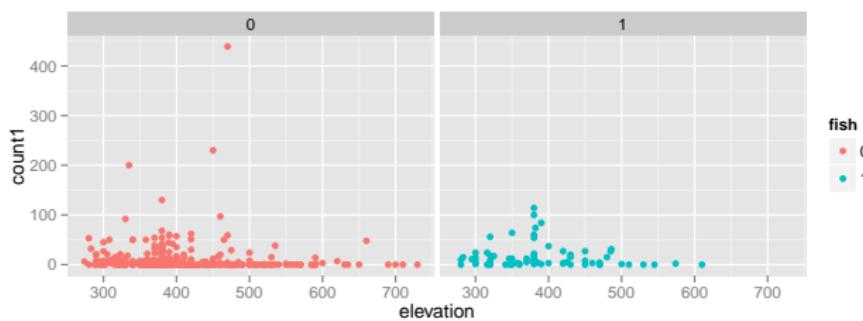
FACETS: SPLIT THE PLOTS BY GROUPS

```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point() +  
  facet_wrap(~fish)
```



FACETS: SPLIT THE PLOTS BY GROUPS

```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point() +  
  facet_wrap(~fish)
```



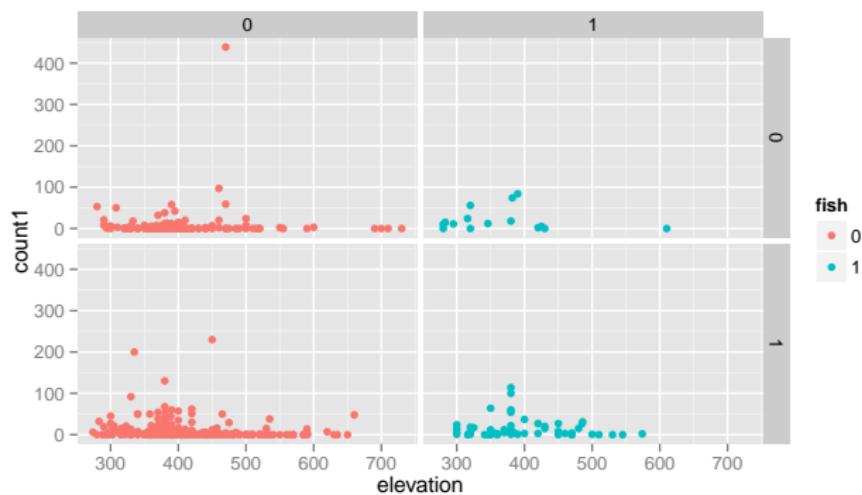
Tip:

The color here is redundant!

FACETS: SPLIT THE PLOTS BY GROUPS

- `facet_wrap()` wraps into a rectangular layout
- `facet_grid(rows ~cols)` wraps into rows and columns

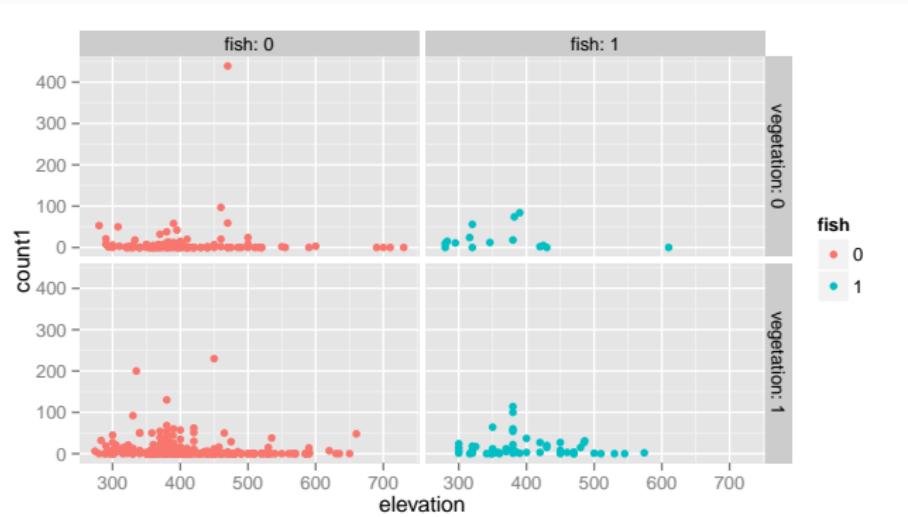
```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point() +  
  facet_grid(vegetation~fish)
```



FACETS: SPLIT THE PLOTS BY GROUPS

- add text to strip to identify variables with **labeller = label_both**

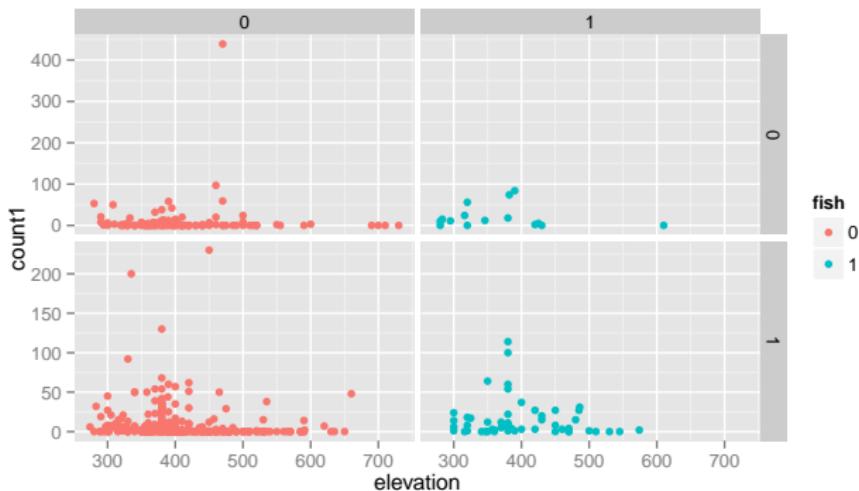
```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point() +  
  facet_grid(vegetation~fish, labeller = label_both)
```



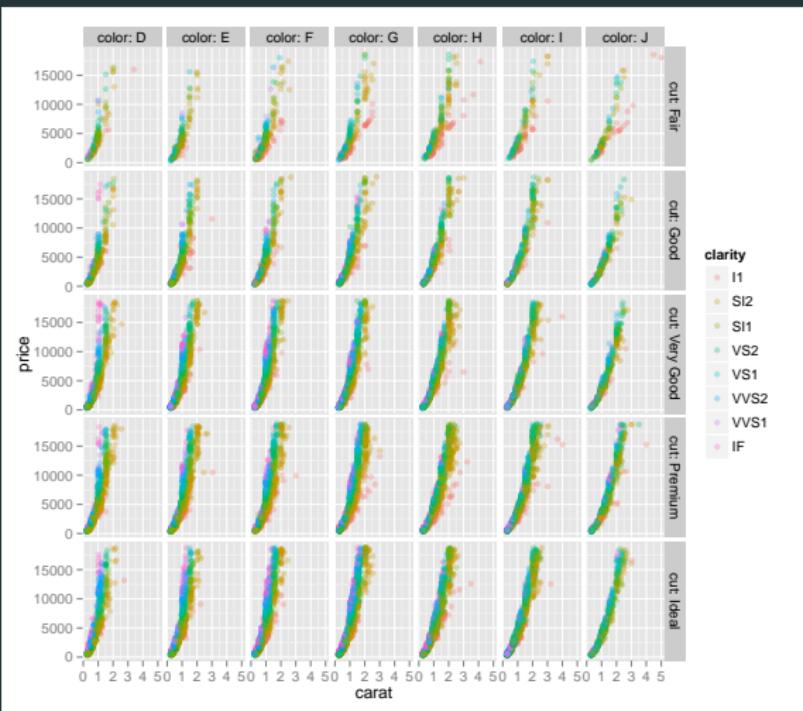
FACETS: SPLIT THE PLOTS BY GROUPS

- Both rows have the same scale.
- use `scales = "free"` or `"frex_x"`, `"free_y"` to adjust

```
ggplot(frogs, aes(x = elevation, y = count1, col = fish)) +  
  geom_point() +  
  facet_grid(vegetation~fish,  
             scales = 'free_y')
```



EXERCISE 4: CREATE THIS PLOT:

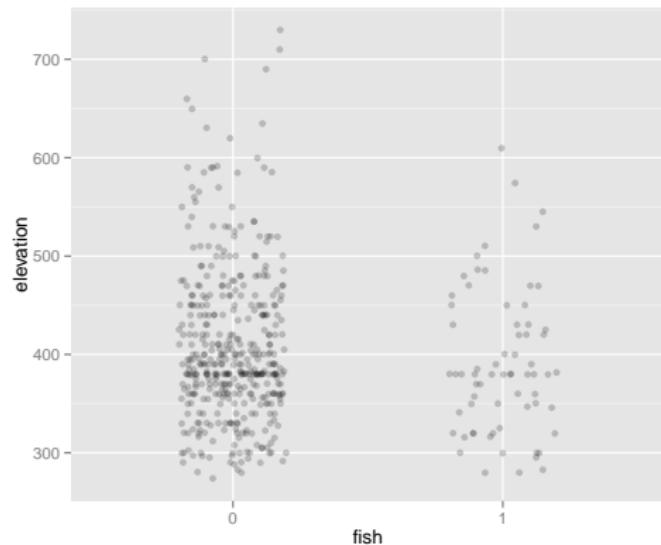


AGGREGATED DATA
(VISUALIZE ERRORS)

AGGREGATE DATA

- Often we want to show aggregates of our data (e.g. means and CIs)

```
ggplot(frogs, aes(x = fish, y = elevation)) +  
  geom_point(position = position_jitter(width = 0.2), alpha = 0.2)
```



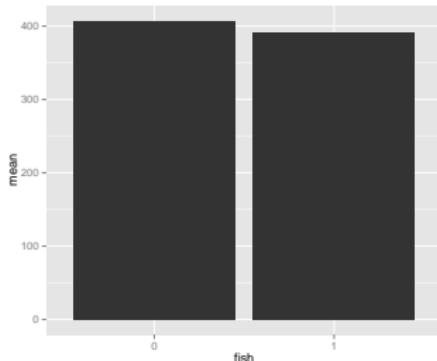
- Aggregate first, then plot!
- A lot of possibilities to aggregate (`plyr`, `dplyr`, `data.table`, `sqldf`, `ave()`, `doBy`, ...)
- I use a `plyr` solution here.

```
require(plyr)
mean_ci <- ddply(frogs, .(fish), summarise,
  mean = mean(elevation),
  err = qnorm(0.975) * sd(elevation) / sqrt(length(elevation)))
mean_ci

##    fish      mean      err
## 1     0 407.5990  7.509866
## 2     1 391.6119 17.560849
```

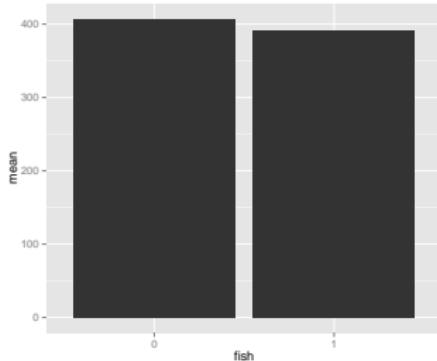
GEOM: BAR

```
ggplot(mean_ci, aes(x = fish, y = mean)) +  
  geom_bar(stat = 'identity')
```



- We need to set `stat = 'identity'` so that the heights of the bars represent the mean.
- Otherwise `geom_bar` tries to transform the data (height = number of cases per group)

```
ggplot(mean_ci, aes(x = fish, y = mean)) +  
  geom_bar(stat = 'identity')
```



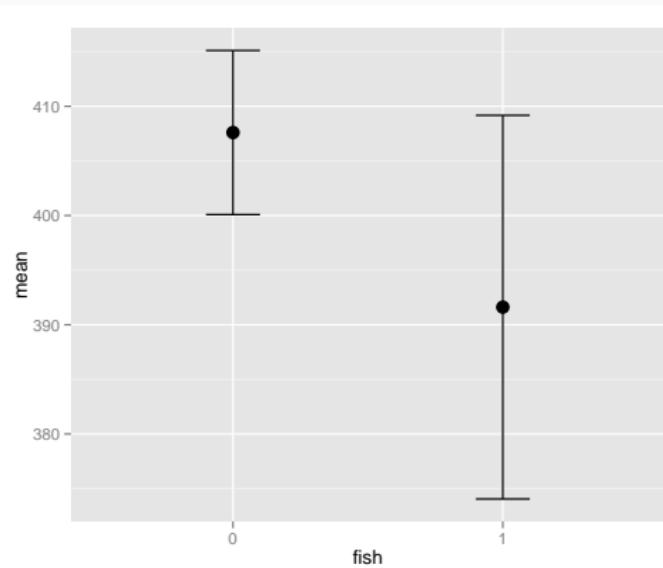
- We need to set `stat = 'identity'` so that the heights of the bars represent the mean.
- Otherwise `geom_bar` tries to transform the data (height = number of cases per group)

Tip:

Don't use bars to display means. Bad data-to-ink ratio (Tufte)!

GEOM: ERRORBARS

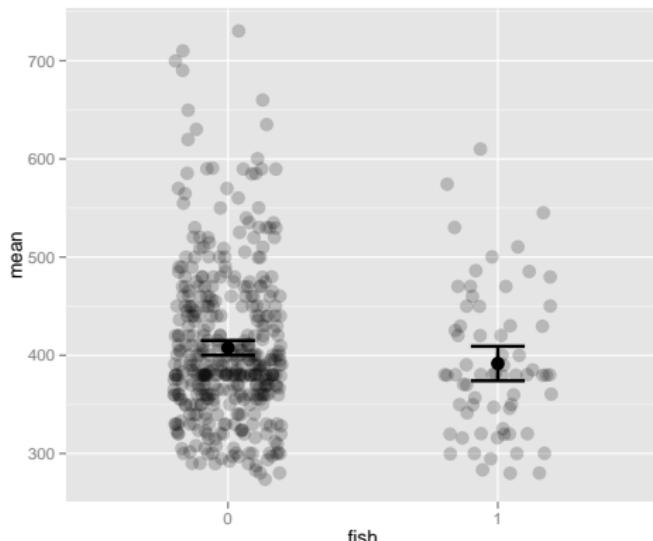
```
ggplot(data = mean_ci, aes(x = fish, y = mean)) +  
  geom_point(size = 4) +  
  geom_errorbar(aes(ymin = mean - err, ymax = mean + err), width = 0.2)
```



- points are enough to represent the means
- `width = 0.2` for smaller whisker (default = 1)

COMBINING DIFFERENT DATASETS INTO ONE PLOT

```
ggplot() +  
  geom_point(data = mean_ci, aes(x = fish, y = mean), size = 4) +  
  geom_errorbar(data = mean_ci, aes(x = fish, y = mean,  
                                     ymin = mean - err, ymax = mean + err),  
                width = 0.2, size = 1) +  
  geom_point(data = frogs, aes(x = fish, y = elevation),  
             position = position_jitter(width = 0.2),  
             alpha = 0.2, size = 4)
```



COMBINING DIFFERENT DATASETS INTO ONE PLOT

```
ggplot() +  
  geom_point(data = mean_ci, aes(x = fish, y = mean), size = 4) +  
  geom_errorbar(data = mean_ci, aes(x = fish, y = mean,  
                                     ymin = mean - err, ymax = mean + err),  
                 width = 0.2, size = 1) +  
  geom_point(data = frogs, aes(x = fish, y = elevation),  
             position = position_jitter(width = 0.2),  
             alpha = 0.2, size = 4)
```

- Because we have multiple datasets, we cannot pass them globally in `ggplot()`
- specify data and aesthetics for each layer / geom
- `geom_violin` + `geom_jitter` are a good combination

COMBINING DIFFERENT DATASETS INTO ONE PLOT

```
ggplot() +  
  geom_point(data = mean_ci, aes(x = fish, y = mean), size = 4) +  
  geom_errorbar(data = mean_ci, aes(x = fish, y = mean,  
                                     ymin = mean - err, ymax = mean + err),  
                width = 0.2, size = 1) +  
  geom_point(data = frogs, aes(x = fish, y = elevation),  
             position = position_jitter(width = 0.2),  
             alpha = 0.2, size = 4)
```

- Because we have multiple datasets, we cannot pass them globally in `ggplot()`
- specify data and aesthetics for each layer / geom
- `geom_violin` + `geom_jitter` are a good combination

Tip:

Plotting the raw data is a good idea (spot errors, make data available, etc...)

MAPS

- `ggmap` can retrieve maps from different sources (Google Maps, OpenStreetMap, StamenMaps, and CloudMade Maps).
- Needs coordinates in lat/lon format
- easy, consistent and modular framework for spatial graphics
- Workflow: Prepare data > retrieve background map > add layers

1) PREPARE DATA: TRANSFORM COORDINATES TO LAT/LON

- Coordinates in frogs dataset are in CH1903 / LV03 format
- Transform using `sp` package to lat / lon

```
require(sp)
frogs_sp <- frogs
coordinates(frogs_sp) <- ~x+y
proj4string(frogs_sp) <- CRS("+init=epsg:21781")
# transform coordinates to WGS84 (lat/lon)
frogs_84 <- spTransform(frogs_sp, CRS("+init=epsg:4326"))
```

```
head(coordinates(frogs_sp))
```

```
##           x      y
## 1 649750 248850
## 2 647350 255750
## 3 650250 244600
## 4 649400 243850
## 5 646700 240750
## 6 646500 253650
```

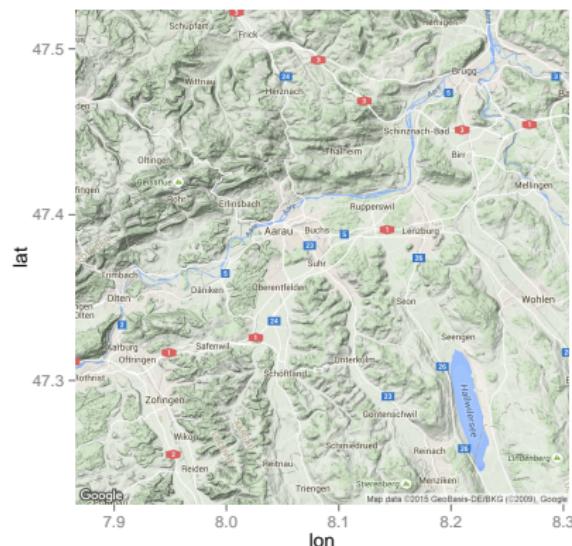
```
head(coordinates(frogs_84))
```

```
##           x      y
## 1 8.097525 47.38860
## 2 8.066471 47.45083
## 3 8.103670 47.35034
## 4 8.092338 47.34366
## 5 8.056288 47.31597
```

2) RETRIEVE BACKGROUND MAP

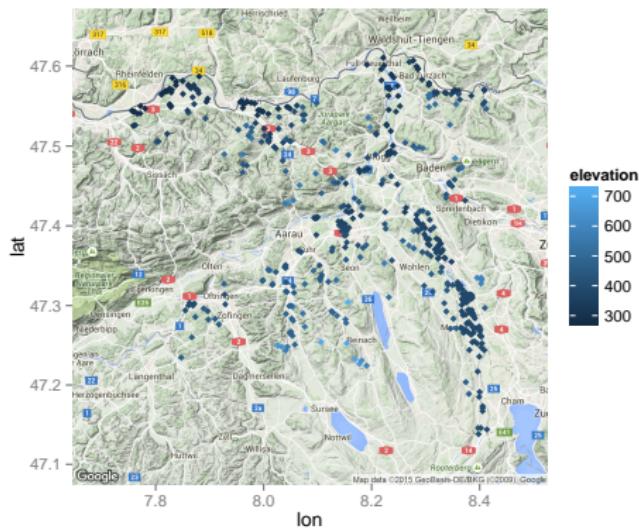
- `get_map()` needs a location of which to retrieve the map
 - I use here the bounding box of the coordinates, but address or a coordinate work also
 - other useful options are `zoom` and `maptype`

```
require(ggmap)
ggmap(get_map(frogs_84@bbox, source = 'google'), zoom = 10)
```



3) ADD LAYERS

```
ggmap(get_map(frogs_84@bbox, source = 'google', zoom = 10)) +  
  geom_point(data = data.frame(frogs_84),  
             aes(x = x, y = y, col = elevation),  
             size = 2,  
             shape = 18)
```



BREAK!

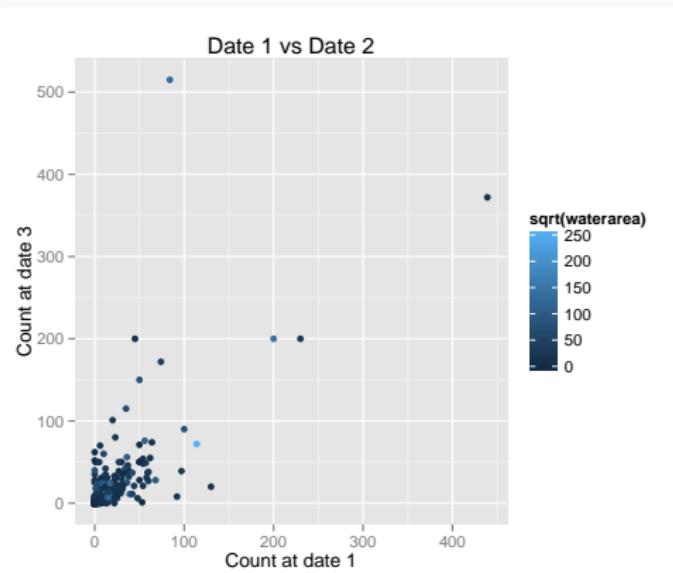
NOW WE HAD ENOUGH OF THE BASICS
NEXT PART: HOW TO CUSTOMIZE YOUR PLOTS.

CUSTOMIZATION AND PRETTIFICATION

WORKING WITH LABELS

- `labs()` can be used to set axis labels & title

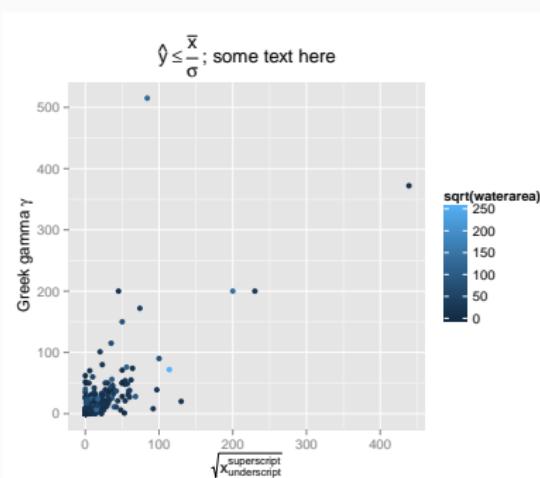
```
ggplot(frogs, aes(x = count1, y = count2, col = sqrt(waterarea))) +  
  geom_point() +  
  labs(x = 'Count at date 1', y = 'Count at date 3', title = 'Date 1 vs Date 2')
```



WORKING WITH LABELS: PLOTMATH

- For mathematical annotation use `plotmath ?plotmath`
- wrap into `expression()`
- separate text with ~

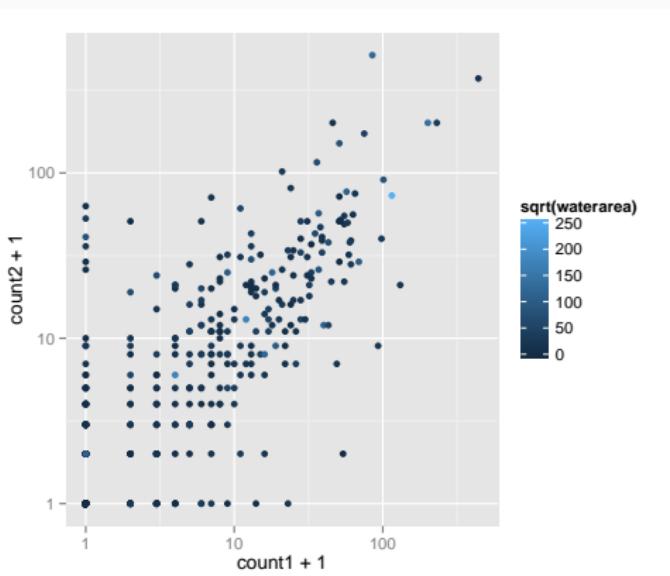
```
ggplot(frogs, aes(x = count1, y = count2, col = sqrt(waterarea))) +  
  geom_point() +  
  labs(x = expression(sqrt(x[underscript])^superscript)),  
       y = expression('Greek gamma'~gamma),  
       title = expression(hat(y) <= frac(bar(x), sigma)~'; some text here'))
```



WORKING WITH AXES: TRANSFORMATIONS / SCALES

- The distribution of counts is highly skewed
- try a double log scale

```
ggplot(frogs, aes(x = count1 + 1, y = count2 + 1, col = sqrt(waterarea))) +  
  geom_point() +  
  scale_x_log10() +  
  scale_y_log10()
```

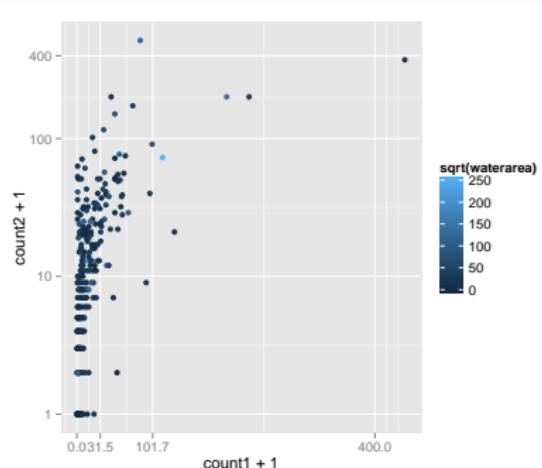


```
ggplot(frogs, aes(x = count1 + 1, y = count2 + 1, col = sqrt(waterarea))) +  
  geom_point() +  
  scale_x_log10() +  
  scale_y_log10()
```

- General form: `scale_aesthetic_type`
- type of scale (e.g. `log10`, `grey`, `continuous` etc.)
- data -> `scale` -> `aesthetic`

SCALES: AXES TICKS

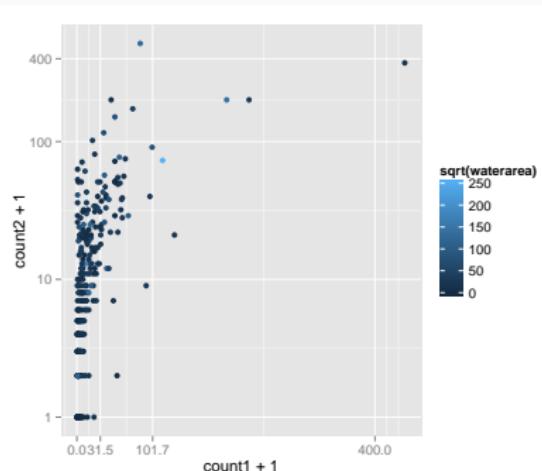
```
ggplot(frogs, aes(x = count1 + 1, y = count2 + 1, col = sqrt(waterarea))) +  
  geom_point() +  
  scale_x_continuous(breaks = c(0, 31.5, 101.7, 400)) +  
  scale_y_log10(breaks = c(1, 10, 100, 400))
```



- use the **breaks=** argument to specify where ticks should be drawn.

SCALES: AXES TICKS

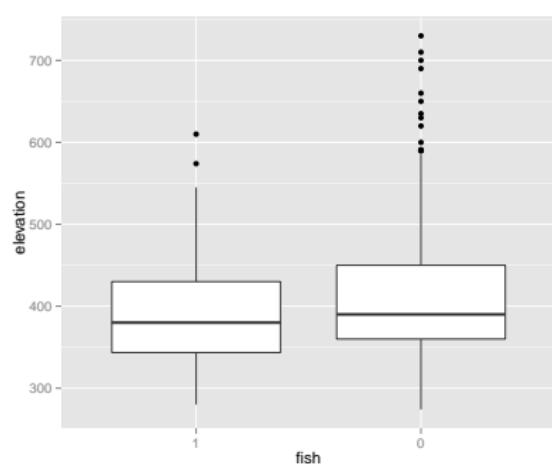
```
ggplot(frogs, aes(x = count1 + 1, y = count2 + 1, col = sqrt(waterarea))) +  
  geom_point() +  
  scale_x_continuous(breaks = c(0, 31.5, 101.7, 400)) +  
  scale_y_log10(breaks = c(1, 10, 100, 400))
```



- use the **breaks=** argument to specify where ticks should be drawn.
- Simplify the axes ticks to min / max only for presentations.

SCALES: CHANGE ORDER DISCRETE AXES

```
ggplot(frogs, aes(x = fish, y = elevation)) +  
  geom_boxplot() +  
  scale_x_discrete(limits = c("1", "0"))
```

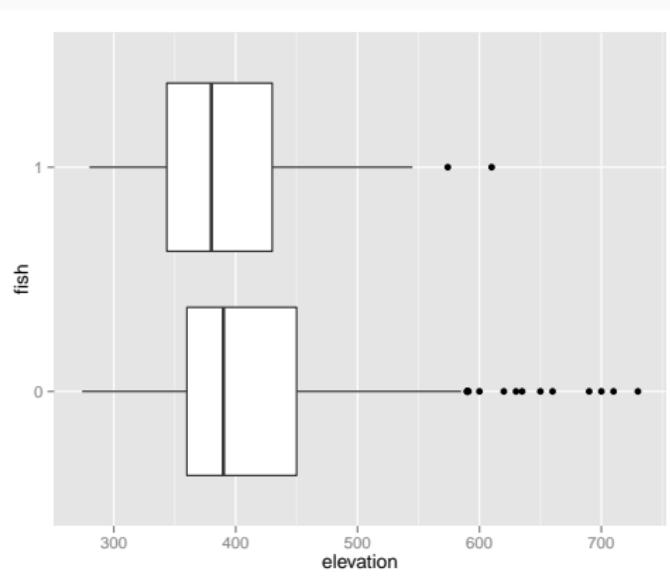


```
# or to generally reverse  
# scale_x_discrete(limits = rev(levels(frogs$fish)))
```

- use the **limits=** argument to specify the order

AXES: SWAP X AND Y

```
ggplot(frogs, aes(x = fish, y = elevation)) +  
  geom_boxplot() +  
  coord_flip()
```

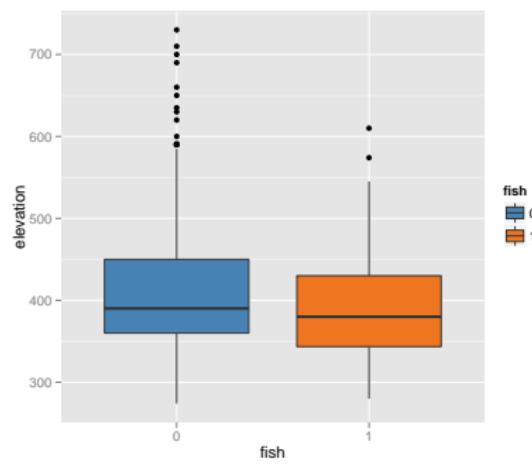


- Sometimes useful to compare the order & magnitude (see slide 12)

SCALES: COLOR (MANUAL)

- Manually select colors from built-in colors
- see `?colors` and `demo("colors")` for possible values

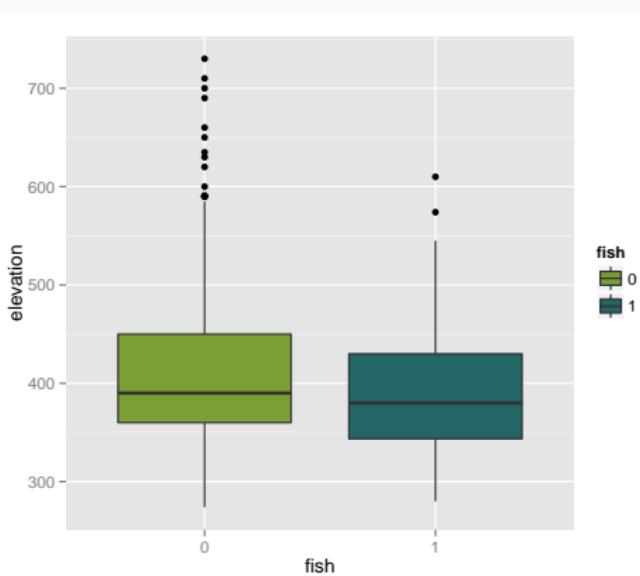
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_fill_manual(values = c('steelblue', 'chocolate2'))
```



SCALES: COLOR (MANUAL)

- Manually select colors from Hex code
- a lot of generators on the web (e.g. paletton.com)

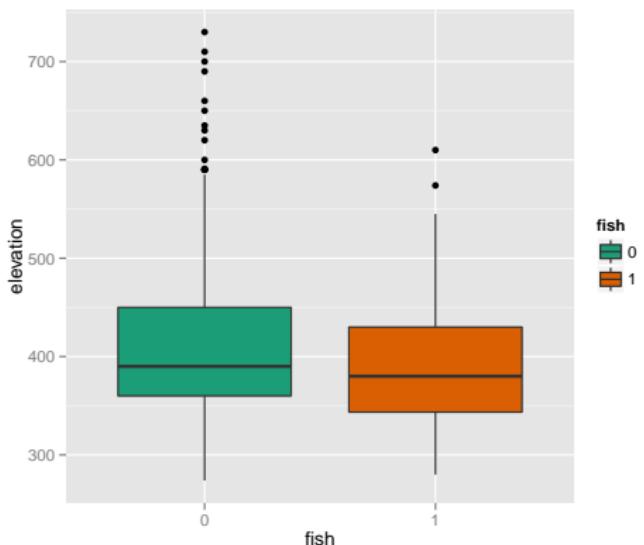
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_fill_manual(values = c('#7A9F35', '#226666'))
```



SCALES: COLOR (PALETTES)

- A palette is a specific order of colors
- Select from a predefined palette
- e.g. <http://colorbrewer2.org/>

```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_fill_brewer(palette = 'Dark2')
```

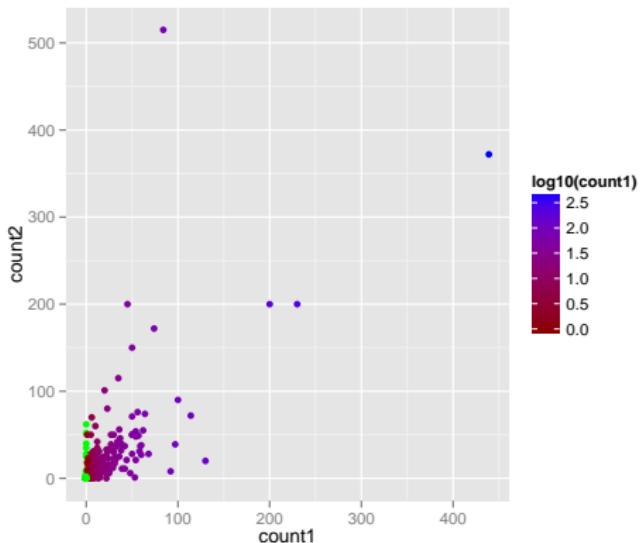


- The palettes can be classified into 3 categories of data:
 - sequential** change of a unique color. Use for qualitative data from low to high.
 - diverging** change from one color to another. Used to show deviations from the mean.
 - qualitative** maximize difference between colors. Used for categorical data.
- Colorblind safe palettes (8% of men are colorblind)

SCALES: COLOR (GRADIENTS)

- for continuous variable use a gradient (no palette)
- Change the gradient for a continuous variable

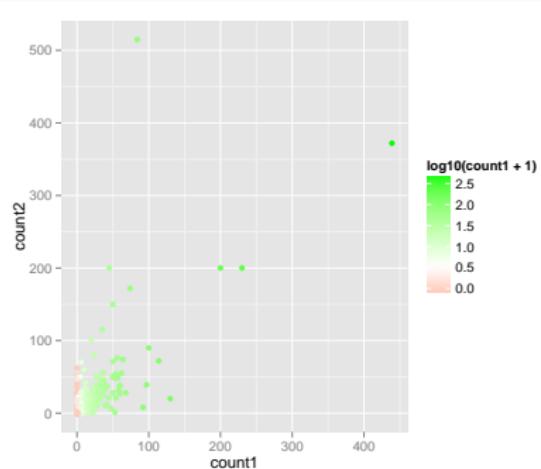
```
ggplot(frogs, aes(x = count1, y = count2, col = log10(count1))) +  
  geom_point() +  
  scale_color_gradient(low = 'darkred', high = 'blue', na.value = 'green')
```



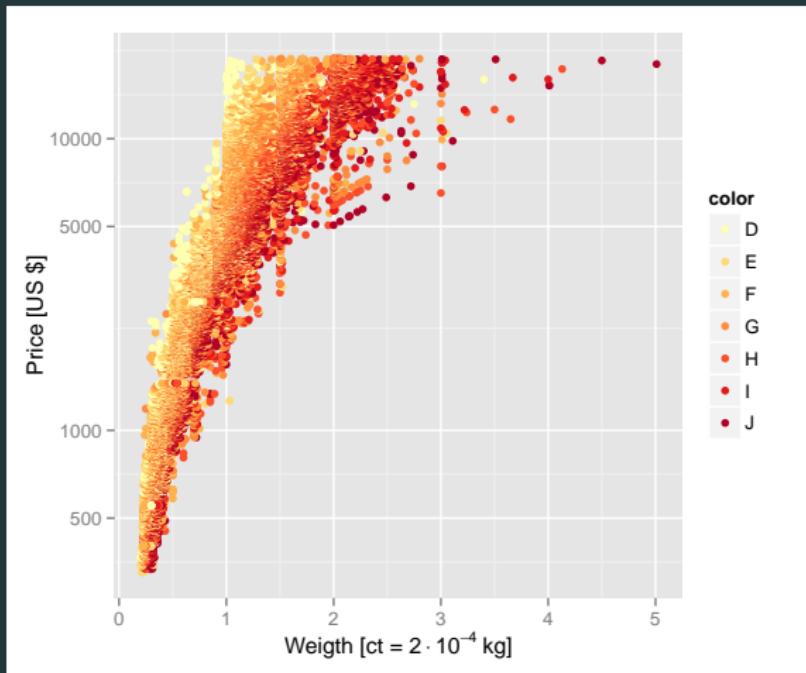
SCALES: COLOR (GRADIENTS)

- `scale_*_gradient2` for a diverging gradient
- midpoint defaults to 0, changed to mean here

```
midp <- mean(log10(frogs$count1+1))
ggplot(frogs, aes(x = count1, y = count2, col = log10(count1+1))) +
  geom_point() +
  scale_color_gradient2(low = 'red', mid = 'white', high = 'green',
                        midpoint = midp, space = 'Lab')
```



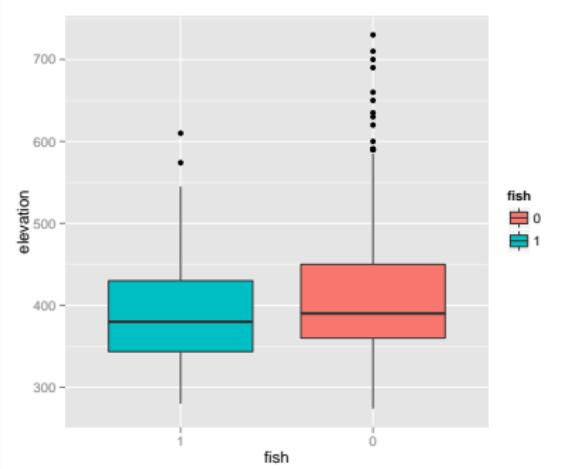
EXERCISE 5: CREATE THIS PLOT:



CHANGE ORDER ON X-AXIS

- used `scale_x_discrete(limits = c("1", "0"))` to change order on x-axis
- how to change order in legend?

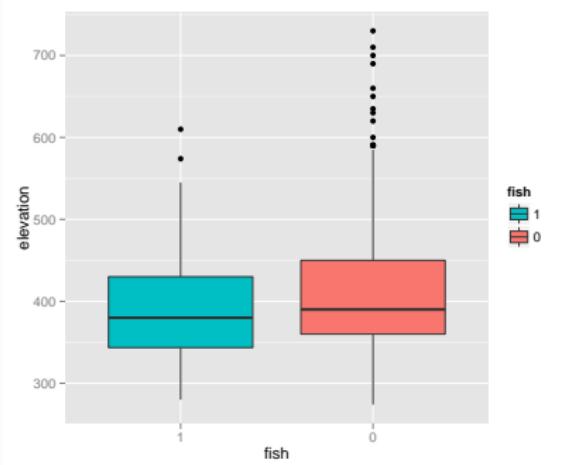
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_x_discrete(limits = c("1", "0"))
```



WORKING WITH LEGENDS: CHANGE ORDER IN LEGEND

- use scale for `fill` aesthetic.
- use `breaks = <levels>`.

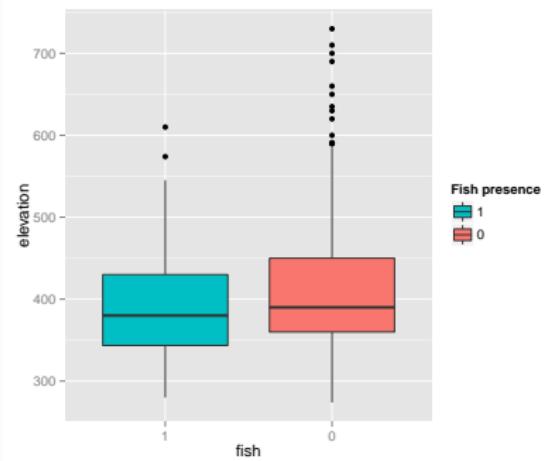
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_x_discrete(limits = c('1', '0')) +  
  scale_fill_discrete(breaks = c('1', '0'))
```



WORKING WITH LEGENDS: CHANGE LEGEND TITLE

- use name = <legendtitle>

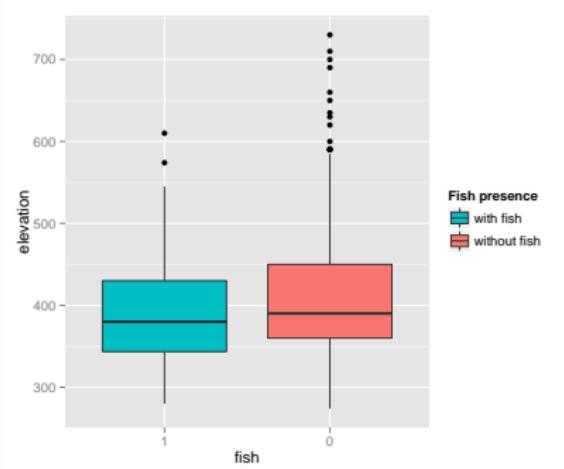
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_x_discrete(limits = c('1', '0')) +  
  scale_fill_discrete(name = 'Fish presence', breaks = c('1', '0'))
```



WORKING WITH LEGENDS: CHANGE LEGEND LABELS

- use name = <legendtitle>

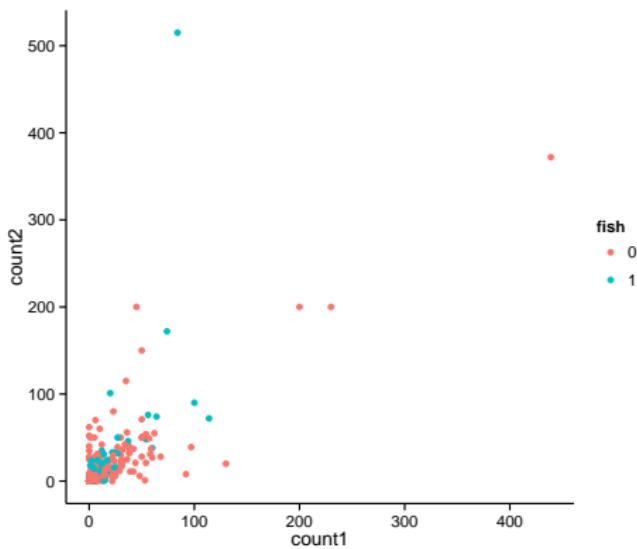
```
ggplot(frogs, aes(x = fish, y = elevation, fill = fish)) +  
  geom_boxplot() +  
  scale_x_discrete(limits = c('1', '0')) +  
  scale_fill_discrete(name = 'Fish presence', breaks = c('1', '0'), la-  
  bels = c('with fish', 'without fish'))
```



CHANGING THE APPEARANCE OF THE PLOTS: THEMES

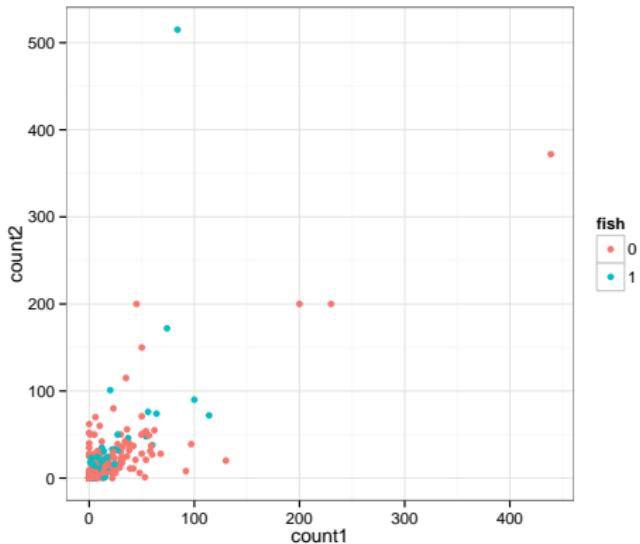
THEMES: PREDEFINED THEMES

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme_classic()
```



THEMES: PREDEFINED THEMES

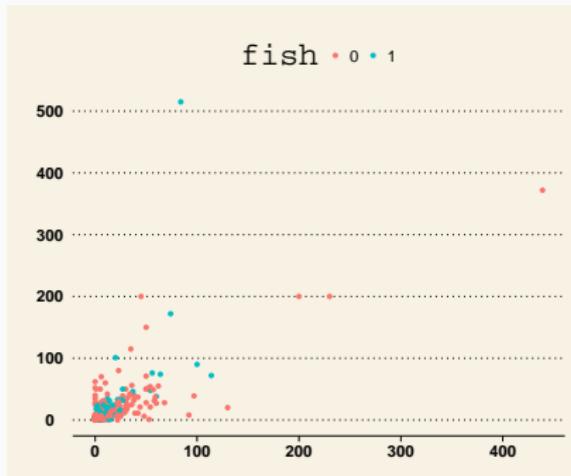
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme_bw()
```



THEMES: PREDEFINED THEMES

- ggplot2 comes with the themes: `bw`, `classic`, `gray`, `light` & `minimal`
- see `ggthemes` package for more
<https://github.com/jrnold/ggthemes>

```
require(ggthemes)
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +
  geom_point() +
  theme_wsj()
```

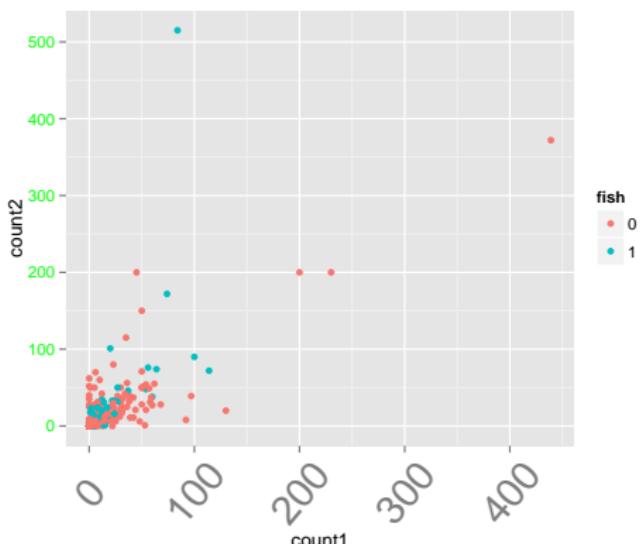


- Allows very fine control over your plots
- If done once, can be reused
- Every theme consists of several elements
- see `?theme` for a complete list
- we go through the most important elements one by one
- Each element consists of
 - `element_text`
 - `element_line`
 - `element_rect`
 - `element_blank`

THEMES: AXIS TEXT

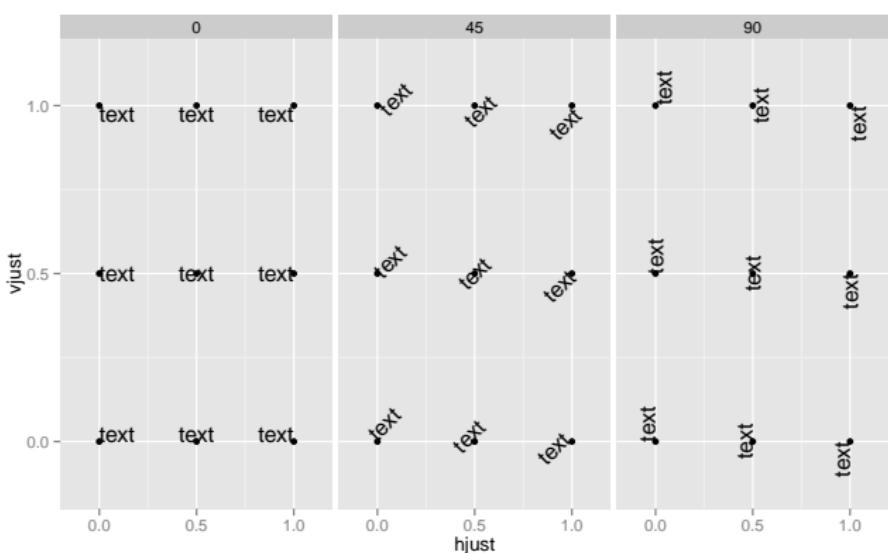
- each element can be controlled with `element_*`
- `text`, `line`, `rect`, `blank` (=omit element)

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(axis.text.x = element_text(angle = 50, size = 25, vjust = 0.5),  
        axis.text.y = element_text(colour = 'green'))
```



hjust, vjust, angle?!

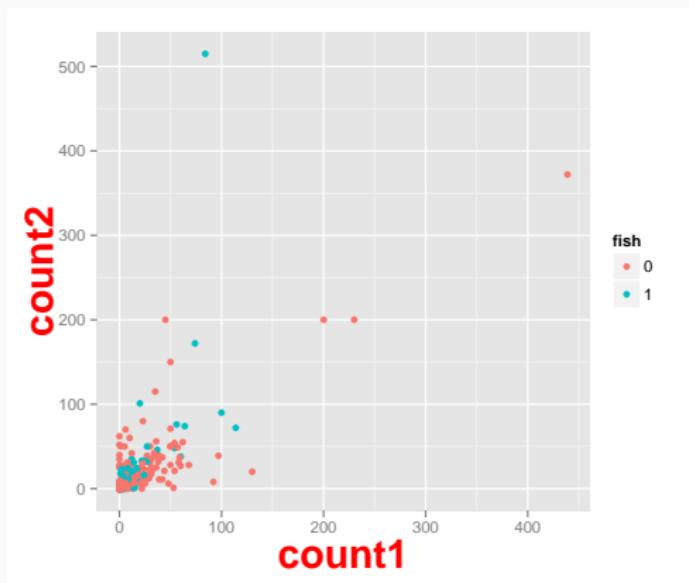
- controls the vertical and horizontal justification
- 0 = left-justified; 1 = right-justified



THEMES: AXIS TITLE

- if we omit the .x / .y the element is applied to both axes

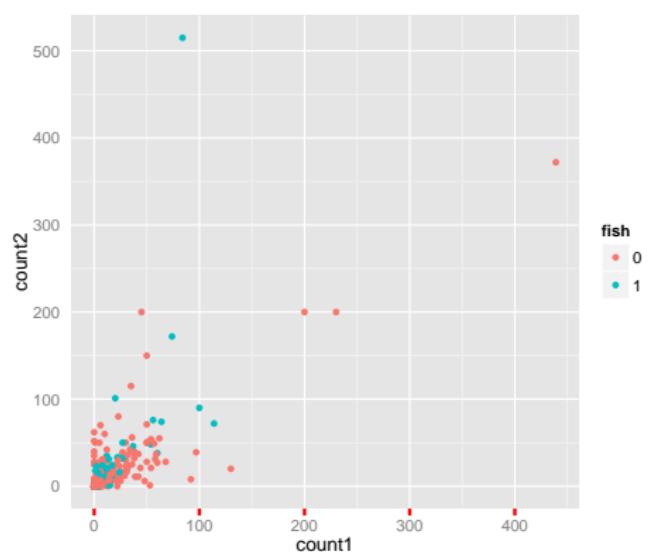
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(axis.title = element_text(face = 'bold', colour = 'red', size = 25))
```



THEMES: AXIS TICKS

- `element_line()` for line elements.

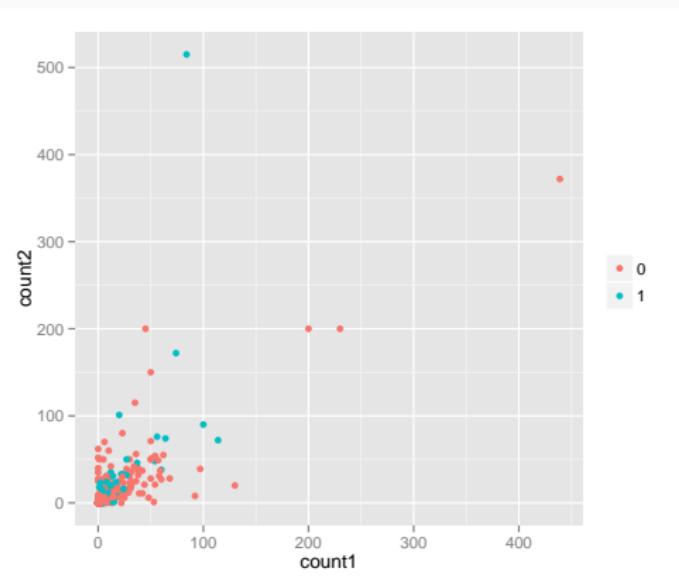
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(axis.ticks.x = element_line(colour = 'red', size = 1),  
        axis.ticks.y = element_blank())
```



THEMES: LEGEND TITLE

- `element_blank()` removes the element from the theme.

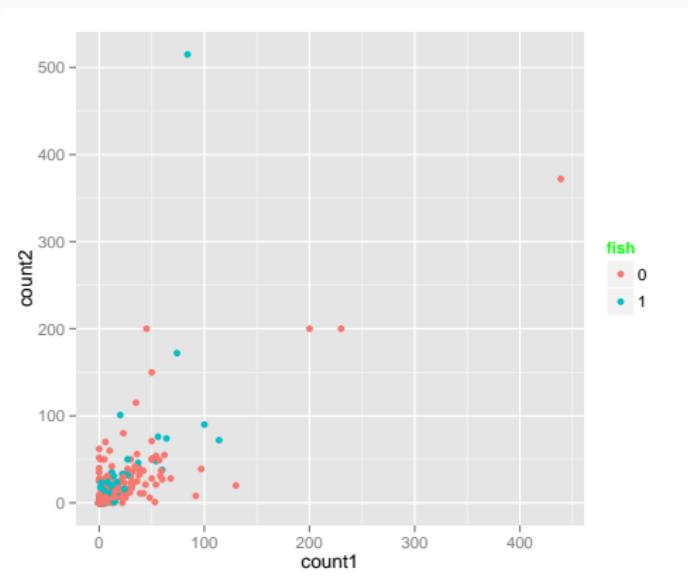
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(legend.title=element_blank())
```



THEMES: LEGEND TITLE

- `element_blank()` removes the element from the theme.

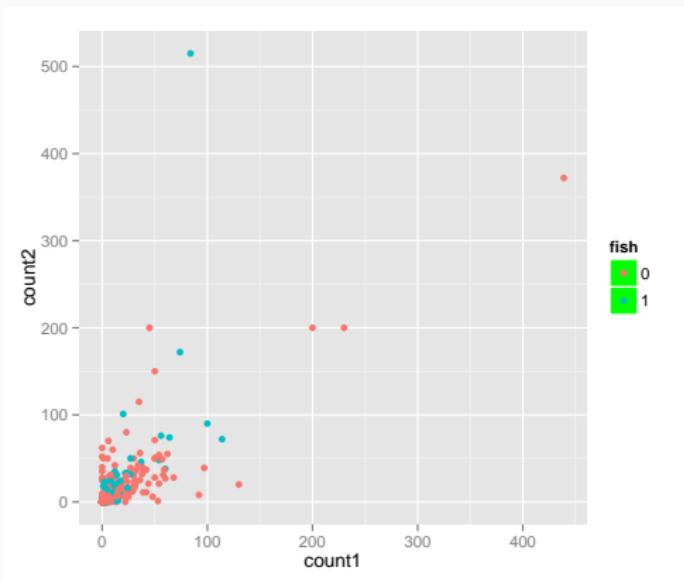
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(legend.title=element_text(colour = 'green'))
```



THEMES: LEGEND BACKGROUND

- `element_rect()` for rectangular elements.

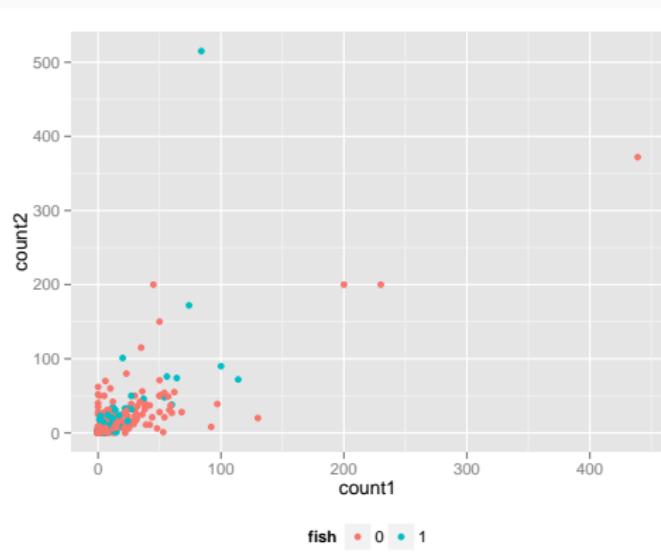
```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(legend.key = element_rect(fill = 'green'))
```



THEMES: POSITION OF THE LEGEND

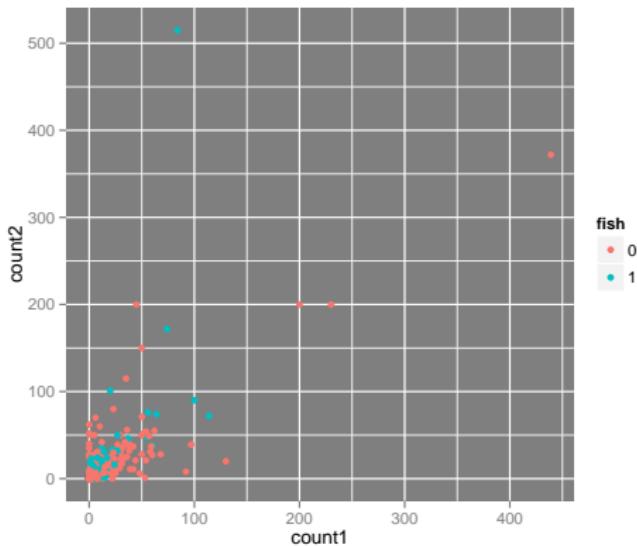
- position can be "none", "left", "right", "bottom", "top"
- also check legend.direction ("horizontal" or "vertical")

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(legend.position = 'bottom')
```



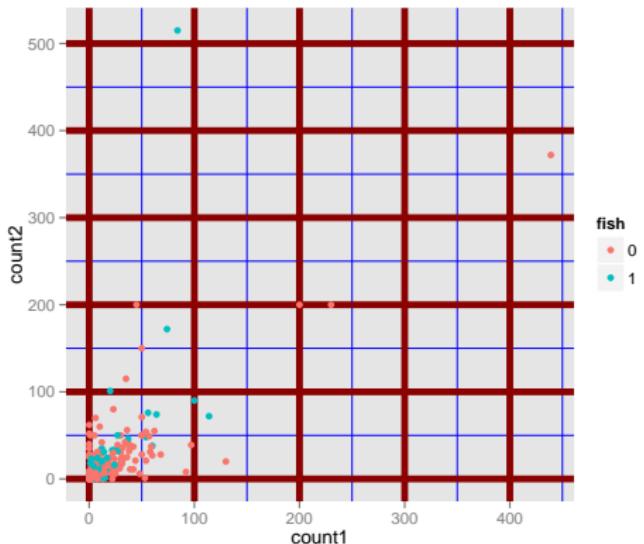
THEMES: PANEL BACKGROUND

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(panel.background = element_rect(fill = 'grey50'))
```



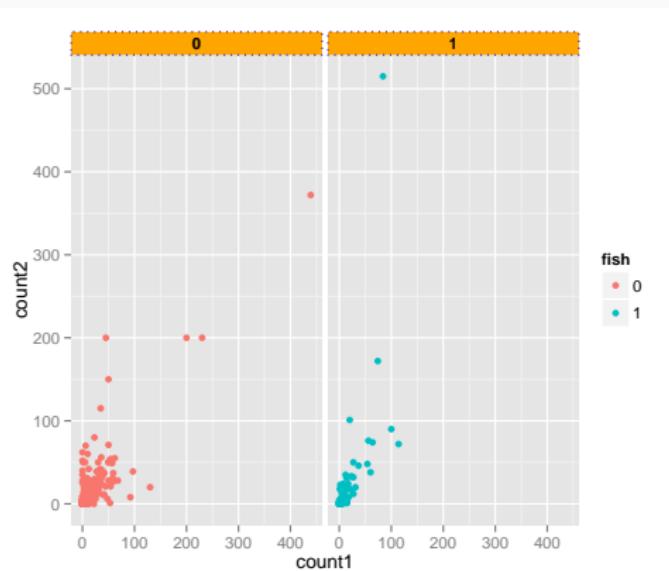
THEMES: PANEL GRID

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  theme(panel.grid.major = element_line(colour = 'darkred', size = 2),  
        panel.grid.minor = element_line(color = 'blue'))
```



THEMES: STRIPS (FACETS)

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  facet_wrap(~fish) +  
  theme(strip.background = element_rect(colour = 'blue', fill = 'orange', line-  
type = 'dotted'),  
        strip.text = element_text(face = 'bold'))
```

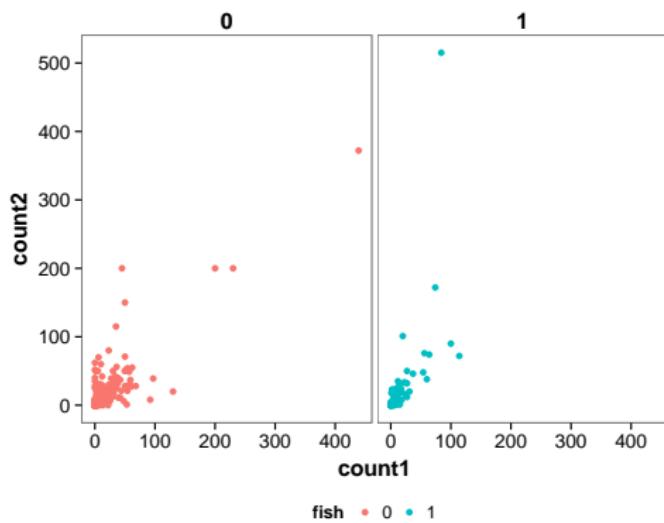


- my (often used) theme is based on `theme_bw()`
- bold and bigger fonts, cleaner strips
- save your theme as object (or better in a separate .R file) and add it to the plot to apply it

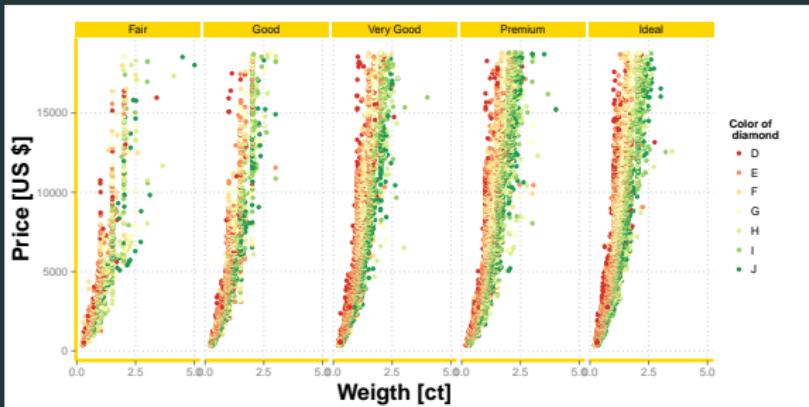
```
edis_theme <- theme_bw(base_size = 12, base_family = "Helvetica") +  
  theme(panel.grid.major = element_blank(),  
        panel.grid.minor = element_blank(),  
        text = element_text(size = 14),  
        axis.text = element_text(size = 12),  
        axis.title.x = element_text(size = 14, face = "bold", vjust = 0),  
        axis.title.y = element_text(size = 14, face = "bold", vjust = 1),  
        legend.position = "bottom",  
        legend.key = element_blank(),  
        strip.background = element_blank(),  
        strip.text = element_text(size = 14, face = 'bold'))
```

THEMES: MY THEME

```
ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point() +  
  facet_wrap(~fish) +  
  edis_theme
```



EXERCISE 6: CREATE THIS PLOT:



EXPORTING YOUR PLOTS

EXPORT YOUR PLOT TO A FILE

- save your plot as object
- use `ggsave()` to save the object

```
p <- ggplot(frogs, aes(x = count1, y = count2, col = fish)) +  
  geom_point()
```

```
# export to different formats  
ggsave('myplot.pdf', p)  
ggsave('myplot.png', p)  
ggsave('myplot.svg', p)
```

USE THE VECTOR FORMAT!

- use the vector format for graphics without loss
- either .pdf (no post-manipulation) or .svg (easy post-manipulation)
- For MS Office use high-resolution png:

```
ggsave('myplot.png', p, width = 7, height = 6, dpi = 300)
```

USE THE VECTOR FORMAT!

- My workflow:
 1. save as .svg
 2. (optional) manipulate in inkscape
 3. convert to needed format using inkscape (bash script)

```
ggsave('myplot.svg', p, width = 7, height = 6)
```

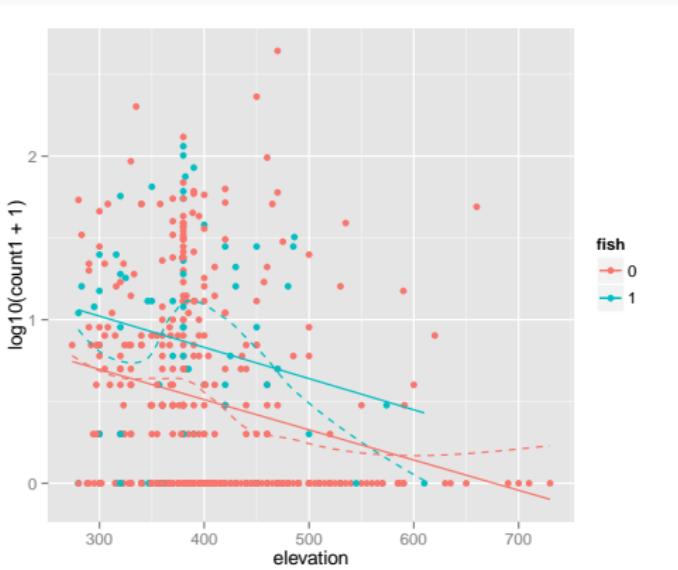
Inkscape Demo.

MISC STUFF

GEOM: SMOOTH

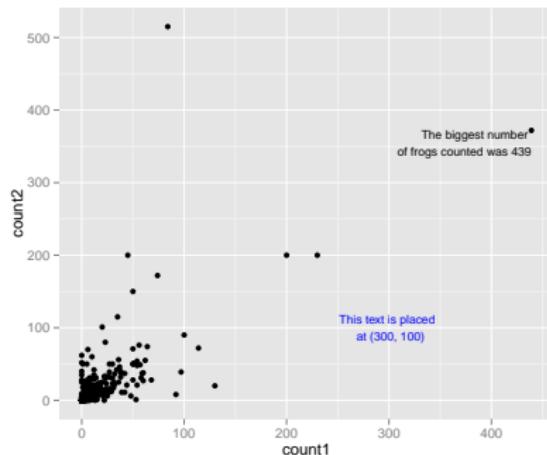
- `geom_smooth` is great to quickly visualize relationships
- I do **not** recommend the built-in model fitting functions
- better separate modelling from plotting

```
ggplot(frogs, aes(y = log10(count1 + 1), x = elevation, col = fish)) +  
  geom_point() +  
  geom_smooth(se = FALSE, linetype = 'dashed') +  
  geom_smooth(se = FALSE, method = 'lm')
```



ANNOTATE THE PLOTS

```
ggplot(frogs, aes(x = count1, y = count2)) +  
  geom_point() +  
  annotate('text',  
    x = max(frogs$count1),  
    y = frogs$count2[which.max(frogs$count1)],  
    label = paste0('The biggest number \n of frogs counted was ',  
                  max(frogs$count1)),  
    vjust = 1, hjust = 1, size = 3) +  
  annotate('text', x = 300, y = 100,  
    label = 'This text is placed \n at (300, 100)',  
    size = 3, col = 'blue')
```





THANKS FOR YOUR ATTENTION!

& A SUCCESSFUL CONFERENCE