

# Introduction to R Graphics

## Supplemental Lecture

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ggplot2

# Remember the tidyverse?

- A consistent framework for describing data visualizations.
- Helps think about and plan graphics outside of R... but implemented deeply in R's in ggplot2 package.
- May also be familiar if you've worked in Tableau – Wilkinson now works for Tableau

# Grammar of graphics components

data

aesthetic mapping

geometric object

statistical transformations

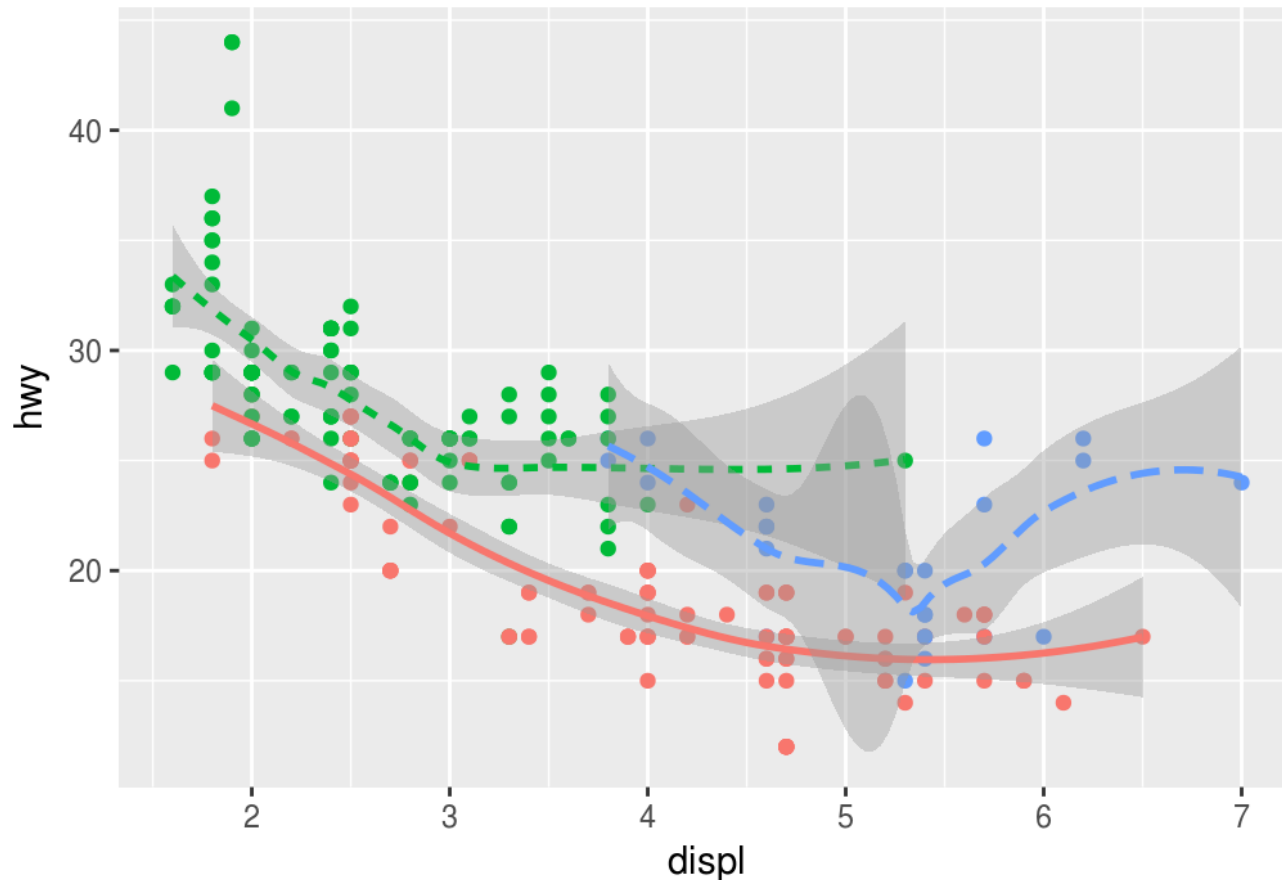
scales

coordinate system

position adjustments

faceting

# Anatomy of a ggplot



data  
aesthetic  
mapping  
geometric object  
statistical  
transformations  
scales  
coordinate  
system  
position  
adjustments  
faceting

# ggplot components

*Or minimally,*

```
ggplot(data= <DATA>)+  
  <GEOM_FUNCTION>(mapping =  
  aes(<MAPPINGS>))
```

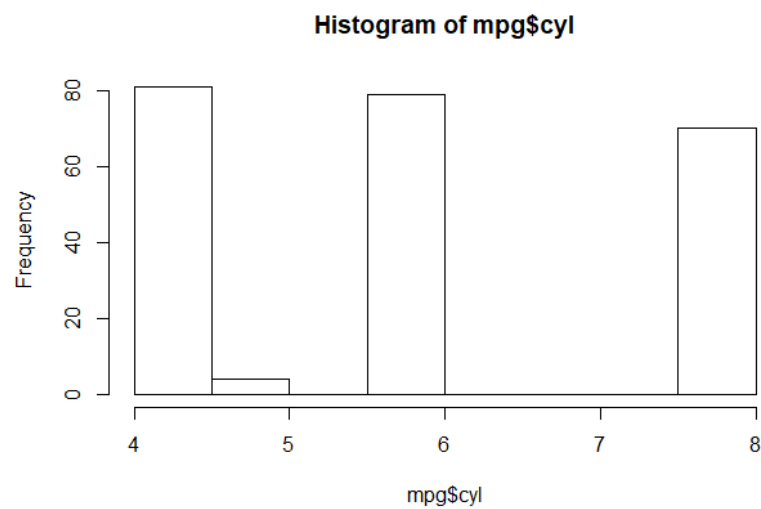
*e.g., using mpg dataset*

```
ggplot(mpg)+  
  geom_point(aes(displ, hwy, color=class))
```

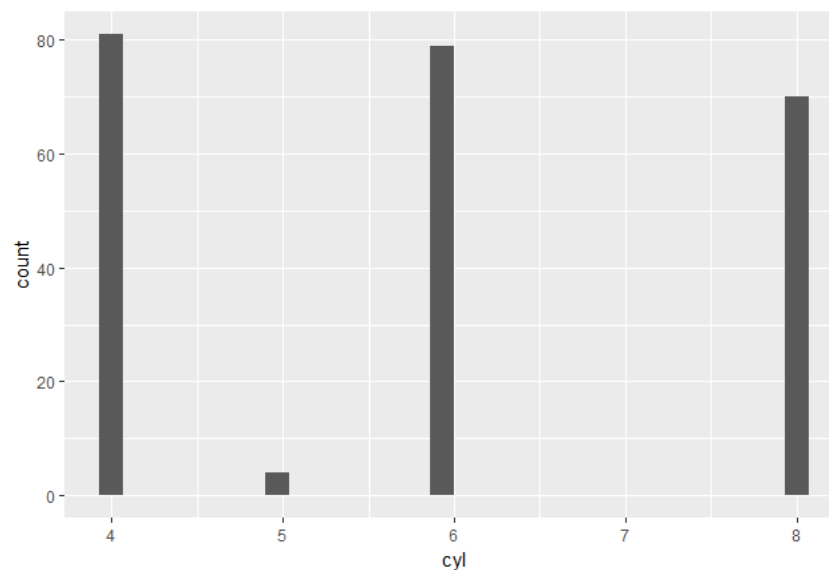
\* Here "+" means "and"

# Versus base / qplot shorthand

```
hist(mpg$cyl)
```



```
ggplot(mpg)+geom_histogram(aes(x=cyl))  
ggplot(mpg, aes(cyl))+geom_histogram()  
qplot(mpg$cyl, geom="histogram")
```



# data

ggplot likes “long”, well structured data.frames

ggplot “**stats**” can make quick transformations

dplyr will help with complicated transformations

tidyr will help go from wide to long



**Extensions** allow ggplot to understand other kinds of data (e.g. maps, network data)



# geoms

...and many more in related packages.

Like other key packages, a cheat sheet is built into R.

**Geoms** - Use a geom function to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

## Graphical Primitives

```
a <- ggplot(economics, aes(date, unemploy))
b <- ggplot(seals, aes(x = long, y = lat))

a + geom_blank()
(Useful for expanding limits)

b + geom_curve(aes(yend = lat + 1,
xend = long + 1, curvature = z)) - x, xend, y, yend,
alpha, angle, color, curvature, linetype, size

a + geom_path(lineend = "butt",
linejoin = "round", linemitre = 1)
x, y, alpha, color, group, linetype, size

a + geom_polygon(aes(group = group))
x, y, alpha, color, fill, group, linetype, size

b + geom_rect(aes(xmin = long, ymin = lat,
xmax = long + 1, ymax = lat + 1)) - xmax, xmin,
ymax, ymin, alpha, color, fill, linetype, size

a + geom_ribbon(aes(ymin = unemploy - 900,
ymax = unemploy + 900)) - x, ymax, ymin
alpha, color, fill, group, linetype, size
```

## Line Segments

common aesthetics: x, y, alpha, color, linetype, size

```
b + geom_abline(aes(intercept = 0, slope = 1))
b + geom_hline(aes(yintercept = lat))
b + geom_vline(aes(xintercept = long))
b + geom_segment(aes(yend = lat + 1, xend = long + 1))
b + geom_spoke(aes(angle = 1:1155, radius = 1))
```

## One Variable

### Continuous

```
c <- ggplot(mpg, aes(hwy)); c2 <- ggplot(mpg)

c + geom_area(stat = "bin")
x, y, alpha, color, fill, linetype, size

c + geom_density(kernel = "gaussian")
x, y, alpha, color, fill, group, linetype, size, weight

c + geom_dotplot()
x, y, alpha, color, fill

c + geom_freqpoly()
x, y, alpha, color, group, linetype, size

c + geom_histogram(binwidth = 5)
x, y, alpha, color, fill, linetype, size, weight

c2 + geom_qq(aes(sample = hwy))
x, y, alpha, color, fill, linetype, size, weight
```

### Discrete

```
d <- ggplot(mpg, aes(fl))

d + geom_bar()
x, alpha, color, fill, linetype, size, weight
```

## Two Variables

### Continuous X, Continuous Y

```
e <- ggplot(mpg, aes(cty, hwy))

e + geom_label(aes(label = cty), nudge_x = 1,
nudge_y = 1, check_overlap = TRUE)
x, y, label, alpha, angle, color, family, fontface,
hjust, lineheight, size, vjust

e + geom_jitter(height = 2, width = 2)
x, y, alpha, color, fill, shape, size

e + geom_point()
x, y, alpha, color, fill, shape, size, stroke

e + geom_quantile()
x, y, alpha, color, group, linetype, size, weight

e + geom_rug(sides = "bl")
x, y, alpha, color, linetype, size

e + geom_smooth(method = lm)
x, y, alpha, color, fill, group, linetype, size, weight

e + geom_text(aes(label = cty), nudge_x = 1,
nudge_y = 1, check_overlap = TRUE)
x, y, label, alpha, angle, color, family, fontface,
hjust, lineheight, size, vjust
```

### Discrete X, Continuous Y

```
f <- ggplot(mpg, aes(class, hwy))

f + geom_col()
x, y, alpha, color, fill, group, linetype, size

f + geom_boxplot()
x, y, lower, middle, upper, ymax, ymin, alpha,
color, fill, group, linetype, shape, size, weight

f + geom_dotplot(binaxis = "y",
stackdir = "center")
x, y, alpha, color, fill, group

f + geom_violin(scale = "area")
x, y, alpha, color, fill, group, linetype, size,
weight
```

### Discrete X, Discrete Y

```
g <- ggplot(diamonds, aes(cut, color))

g + geom_count()
x, y, alpha, color, fill, shape, size, stroke
```

### Continuous Bivariate Distribution

```
h <- ggplot(diamonds, aes(carat, price))

h + geom_bin2d(binwidth = c(0.25, 500))
x, y, alpha, color, fill, linetype, size, weight

h + geom_density2d()
x, y, alpha, colour, group, linetype, size

h + geom_hex()
x, y, alpha, colour, fill, size
```

## Continuous Function

```
i <- ggplot(economics, aes(date, unemploy))

i + geom_area()
x, y, alpha, color, fill, linetype, size

i + geom_line()
x, y, alpha, color, group, linetype, size

i + geom_step(direction = "hv")
x, y, alpha, color, group, linetype, size
```

## Visualizing error

```
df <- data.frame(grp = c("A", "B"), fit = 4:5, se = 1:2)
j <- ggplot(df, aes(grp, fit, ymin = fit - se, ymax = fit + se))
```

```
j + geom_crossbar(fatten = 2)
x, y, ymax, ymin, alpha, color, fill, group,
linetype, size

j + geom_errorbar()
x, ymax, ymin, alpha, color, group, linetype,
size, width (also geom_errorbarh())

j + geom_linerange()
x, ymin, ymax, alpha, color, group, linetype, size

j + geom_pointrange()
x, y, ymin, ymax, alpha, color, fill, group,
linetype, shape, size
```

## Maps

```
data <- data.frame(murder = USArrests$Murder,
state = tolower(rownames(USArrests)))
map <- map_data("state")
k <- ggplot(data, aes(fill = murder))

k + geom_map(aes(map_id = state), map = map) +
expand_limits(x = map$long, y = map$lat)
map_id, alpha, color, fill, linetype, size
```

## Three Variables

```
seals$z <- with(seals, sqrt(delta_long^2 + delta_lat^2))
l <- ggplot(seals, aes(long, lat))

l + geom_contour(aes(z = z))
x, y, z, alpha, colour, group, linetype, size,
weight

l + geom_raster(aes(fill = z), hjust = 0.5,
vjust = 0.5, interpolate = FALSE)
x, y, alpha, fill

l + geom_tile(aes(fill = z))
x, y, alpha, color, fill, linetype, size, width
```

# Aesthetic Mappings

## Data

Numbers & Factors (characters coerced)

- meduc
- mage
- cigdur
- wksgest
- preterm\_f
- pnc5\_f
- county\_name
- raceeth\_f
- ...and more

## Aesthetics

- x
- y
- color (name, rgb)
- fill
- size
- linetype (int or name)
- alpha
- height
- width
- shape
- angle
- ....and more

## Geometry

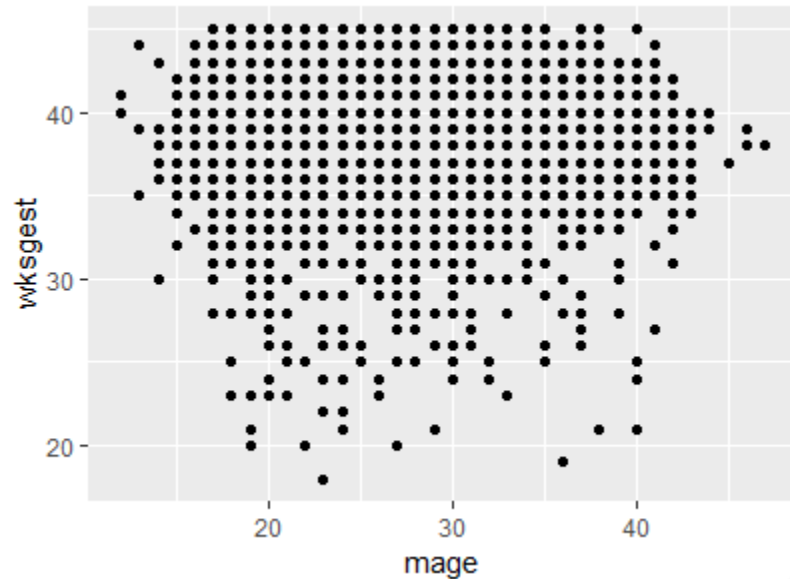
Minimum for `geom_point()`,  
with rest are defaulted

<https://cran.r-project.org/web/packages/ggplot2/vignettes/ggplot2-specs.html>

# Basic geometries and mappings

1. Let's create a **scatterplot** of `wksgest` and `mage` using `ggplot` and `geom_point`.

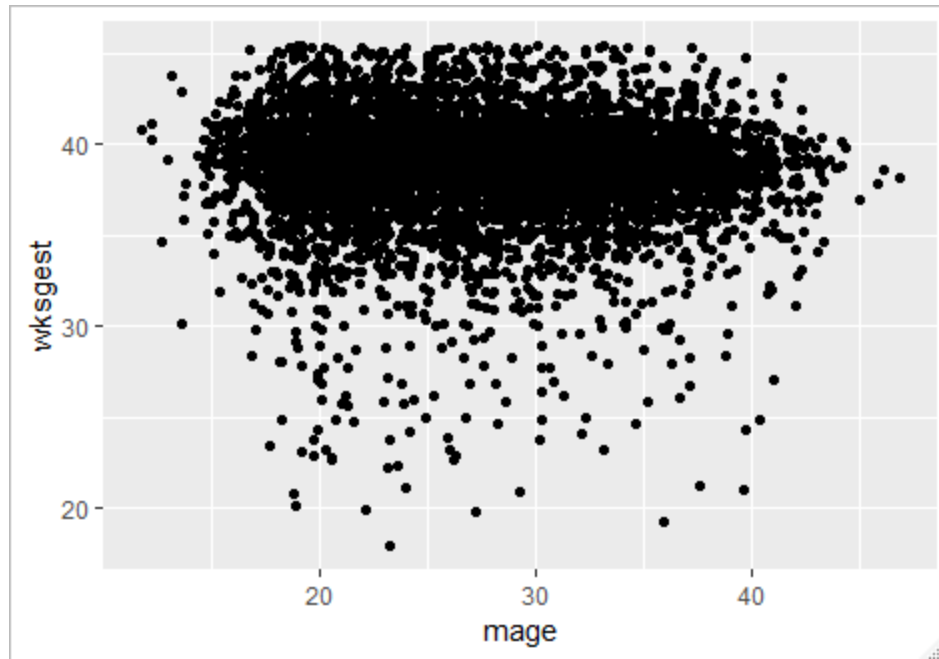
```
ggplot(births_10k, aes(mage, ksgest))+geom_point()
```



# Basic geometries and mappings

D'oh! **Overplotting!** Use the `geom_jitter()` geometry instead.

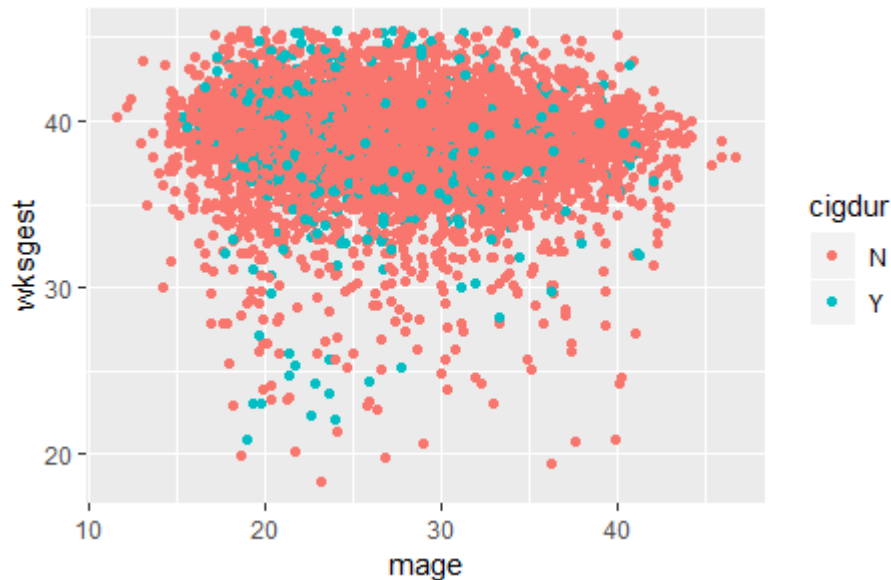
```
ggplot(births_10k, aes(mage, ksgest))+geom_jitter()
```



# Basic geometries and mappings

Let's try **colors**. Map `cigdur` to color. That's it!

```
ggplot(births_10k, aes(mage, wksgest, color=cigdur))+  
  geom_jitter()
```

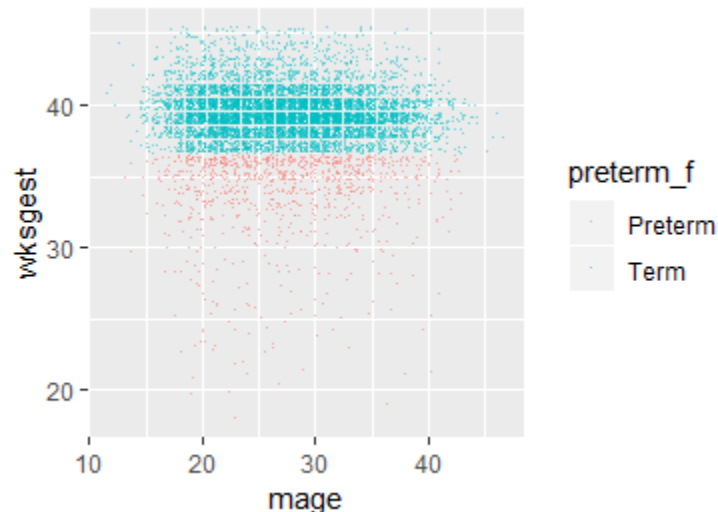


# Basic geometries and mappings

Fancier: change color to `preterm_f`, change the point character to a "." and alpha to 0.5. Note global aes!

```
ggplot(births_10k, aes(mage, wksgest, color=preterm_f))+  
  geom_jitter(pch=".", alpha=0.1)
```

# ^ Typical chained spacing in last two examples, like `dplyr`



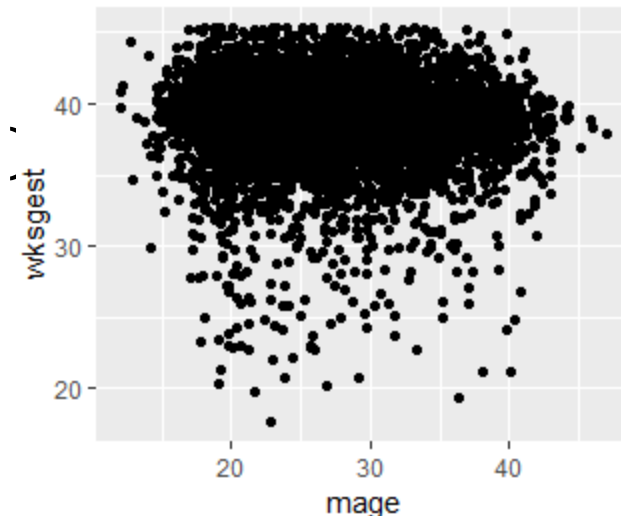
# Aesthetic inheritance

Subsequent geometric layers will inherit the aesthetic mappings from the original `ggplot()` call unless they're overwritten.

Meaning these are equivalent:

```
ggplot(births_10k, aes(mage, wksgest))+  
  geom_jitter()
```

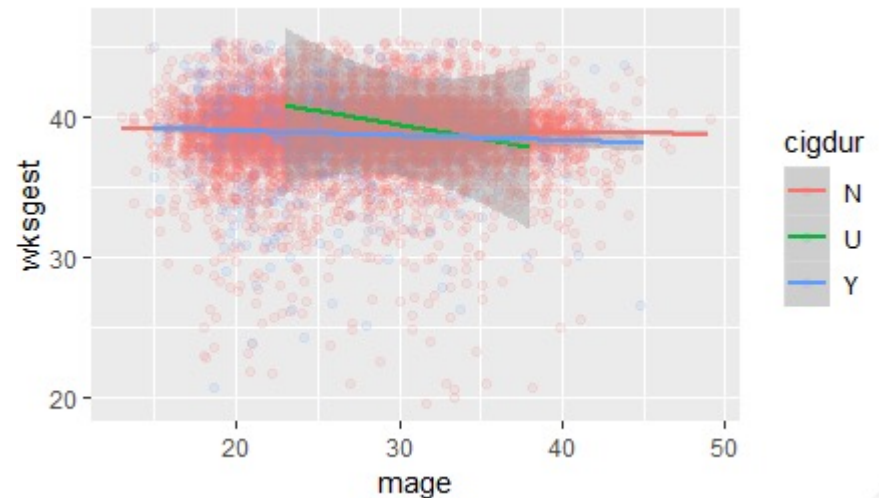
```
ggplot(births_10k)+  
  geom_jitter(aes(mage, wksgest))  
#^ equivalent
```



# Aesthetic inheritance 2

...but there's good reason to be intentional about this – like when we want multiple geometries to use the same mappings

```
ggplot(births_10k, aes(mage, wksgest, color=cigdur))+  
  geom_jitter(alpha=0.1) +  
  geom_smooth(method="lm")
```

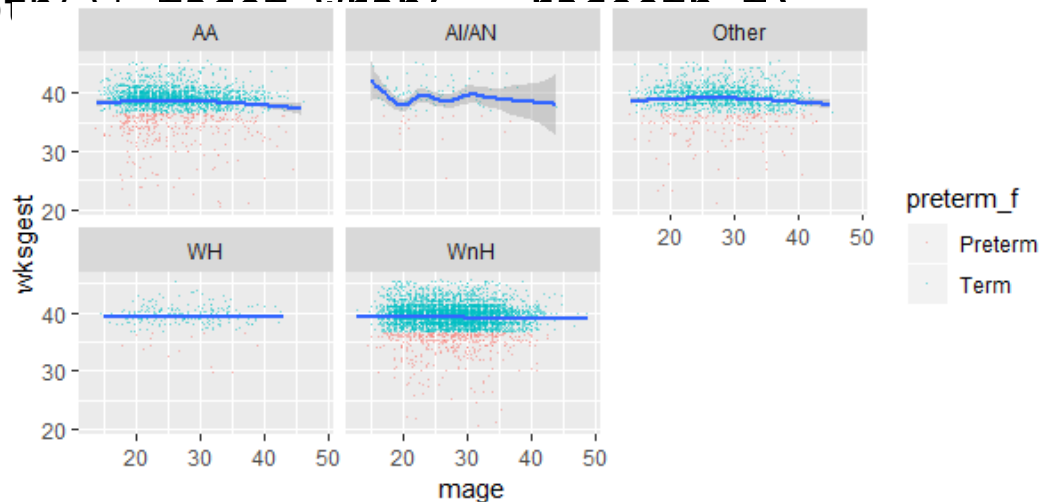




# Facets (wrap/grid):small multiples

- Facets take an R formula object (e.g.  $y \sim x$ ) and split your graph into small multiples based on that.
- Can also “free the scales” so they aren’t shared across plots with `scales = “free_x”` and/or `“free_y”`

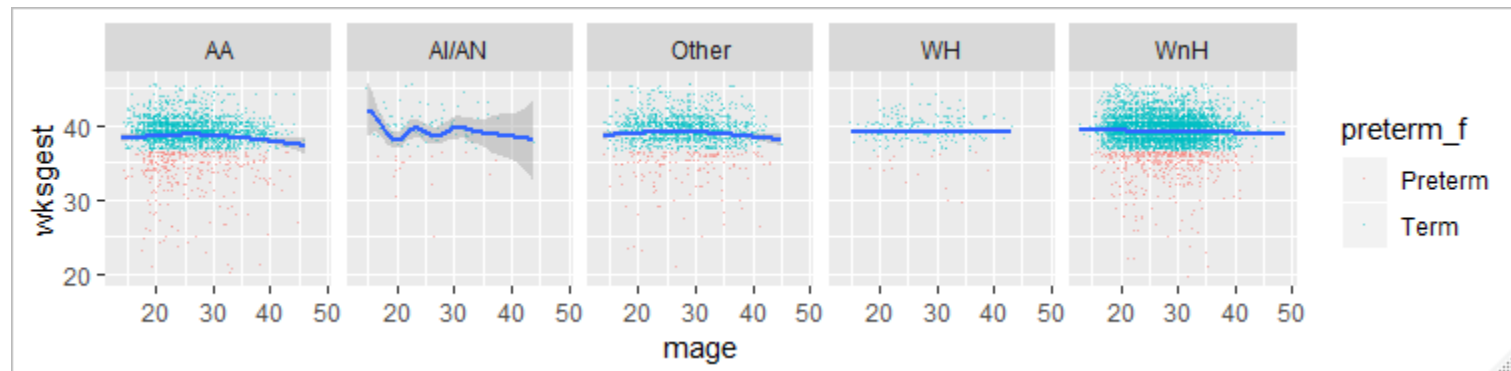
```
ggplot(births_10k, aes(mage, wksgest)) +  
  geom_jitter(aes(color=preterm_f), pch=".", alpha=0.5) +  
  geom_smooth()
```



# Facets (grid/wrap):small multiples

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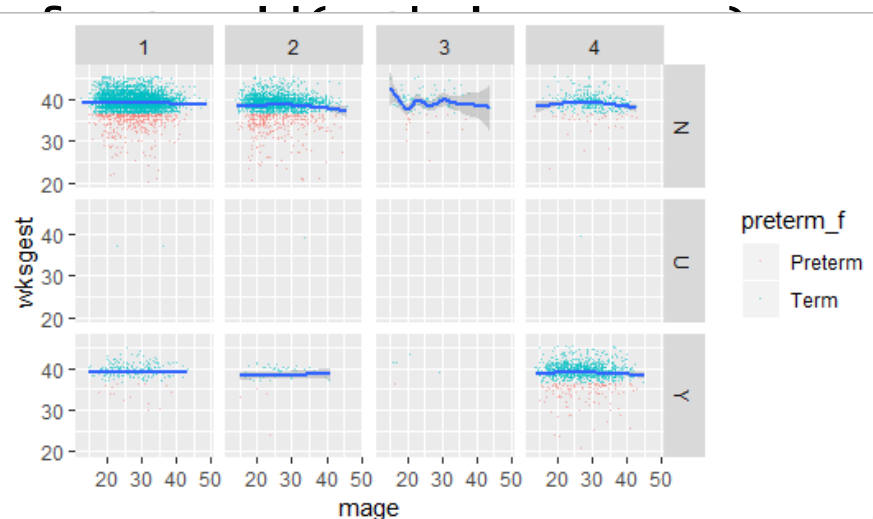
```
ggplot(births_10k, aes(mage, wksgest)) +  
  geom_jitter(aes(color=preterm_f), pch=".", alpha=0.5) +  
  geom_smooth() + facet_grid(~ raceeth_f)
```



# Facets (grid/wrap):small multiples

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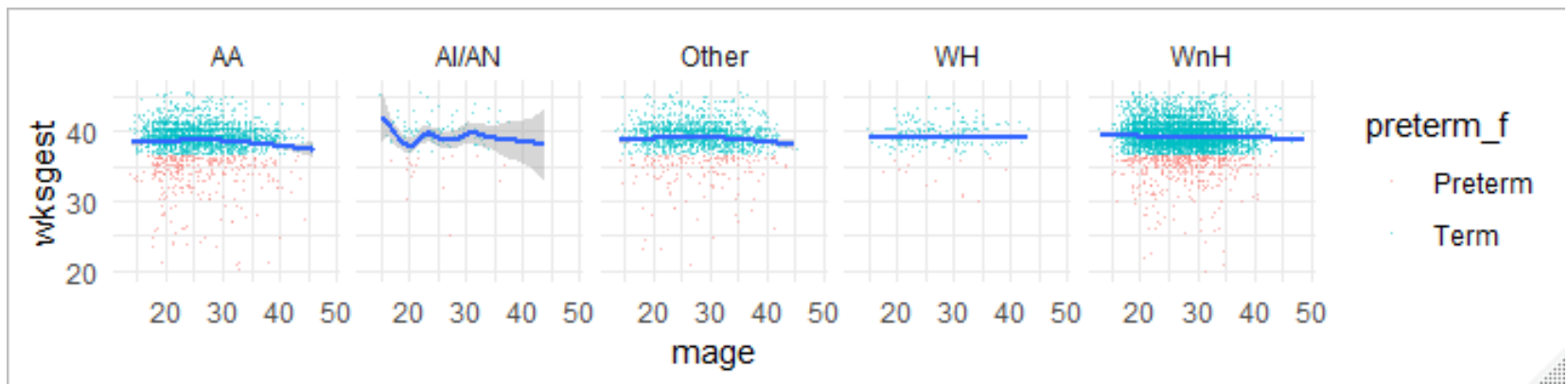
```
ggplot(births_10k, aes(mage, wksgest)) +  
  geom_jitter(aes(color=preterm_f), pch=".", alpha=0.5) +  
  geom_smooth()
```



# Themes

- Change the theme with `theme_NAME()`, e.g. `theme_minimal()`.
- Can define your own themes, or tweak existing ones.
- See <https://cran.r-project.org/web/packages/ggthemes/vignettes/ggthemes.html> for more themes. More on ggplot extensions later!

```
ggplot(births_10k, aes(mage, wksgest))+  
  geom_jitter(aes(color=preterm_f), pch=".", alpha=0.5)+  
  geom_smooth()+ facet_grid( ~ raceeth_f)+  
  theme_minimal()
```



# Layer=data+stats+geoms

- Hadley's (package author) underlying theory from grammar of graphics: layer=data+stats+geoms. Often stat or geom imply the other, so each has a default parameter of the other.
  - <http://vita.had.co.nz/papers/layered-grammar.pdf>
- Excellent Stack Overflow Review
  - <https://stackoverflow.com/questions/38775661/what-is-the-difference-between-geoms-and-stats-in-ggplot2>
- Default stats for geoms: <http://sape.inf.usi.ch/quick-reference/ggplot2/geom>

# Stat variables

- geoms, behind the scenes, often calculate a new dataframe to actually plot on screen.
- `stat_<thing>` calculates a new dataframe explicitly
- That dataframe has some “secret” (documented) variable names, accessible by special inline variables

`..count..`      `..ncount..`      `..density..`      `..ndensity..`

`..count..`      `..prop..`      Etc.

- What if I want to refer to those new variables elsewhere in the ggplot call? (rare use – `dplyr` instead!)