

# comparing two independent means: hypothesis testing

Dr. Merlise Clyde

weight gain during pregnancy and mother's age

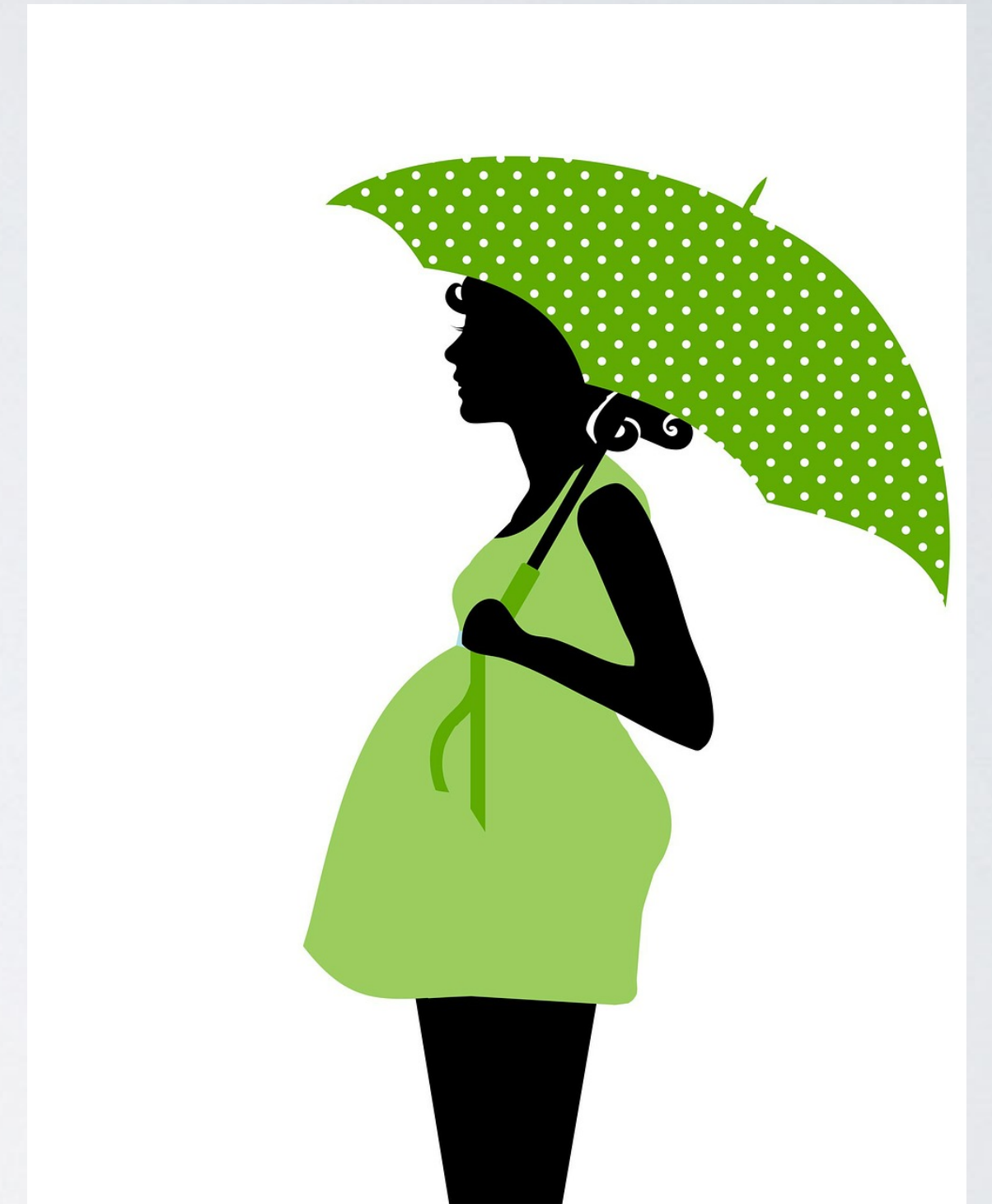
2004 survey of births on North Carolina



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is the average weight gain of older mothers different from the average weight gain of younger mothers?



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- ▶ need prior distributions for  $\alpha, \mu, \sigma^2$  under both hypotheses



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- ▶ Jeffreys-Zellner-Siow or 'JZS'

## R code

```
library(statsr)
data(nc)
bayes_inference(y=gained, x=mature, data=nc, type='ht',
                statistic='mean', alternative='twosided', null=0,
                prior='JZS', r=1, method='theo', show_summ=FALSE)

## Hypotheses:
## H1: mu_mature mom = mu_younger mom
## H2: mu_mature mom != mu_younger mom
##
## Priors:  $P(H1) = 0.5$   $P(H2) = 0.5$ 
##
## Results:
##  $BF[H1:H2] = 5.7162$ 
##  $P(H1|data) = 0.8511$ 
##  $P(H2|data) = 0.1489$ 
##
```



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next: putting it all together and summarizing results