

# Session 3 Live Coding

## Bayesian Modeling

```
# Load our libraries
library(tidyverse)
library(rstanarm)
library(bayesplot)

innovation.df <- read_csv("DSOM5509.csv") %>%
  mutate(myInnovation = ((INN1 + INN2 + INN3) / 3))

bayesian.model <- stan_lm(SGR ~ myInnovation, data = innovation.df,
  prior = R2(location = 0.2, what = "mean"),
  seed = 1, refresh = 0)
```

```
bayesian.model

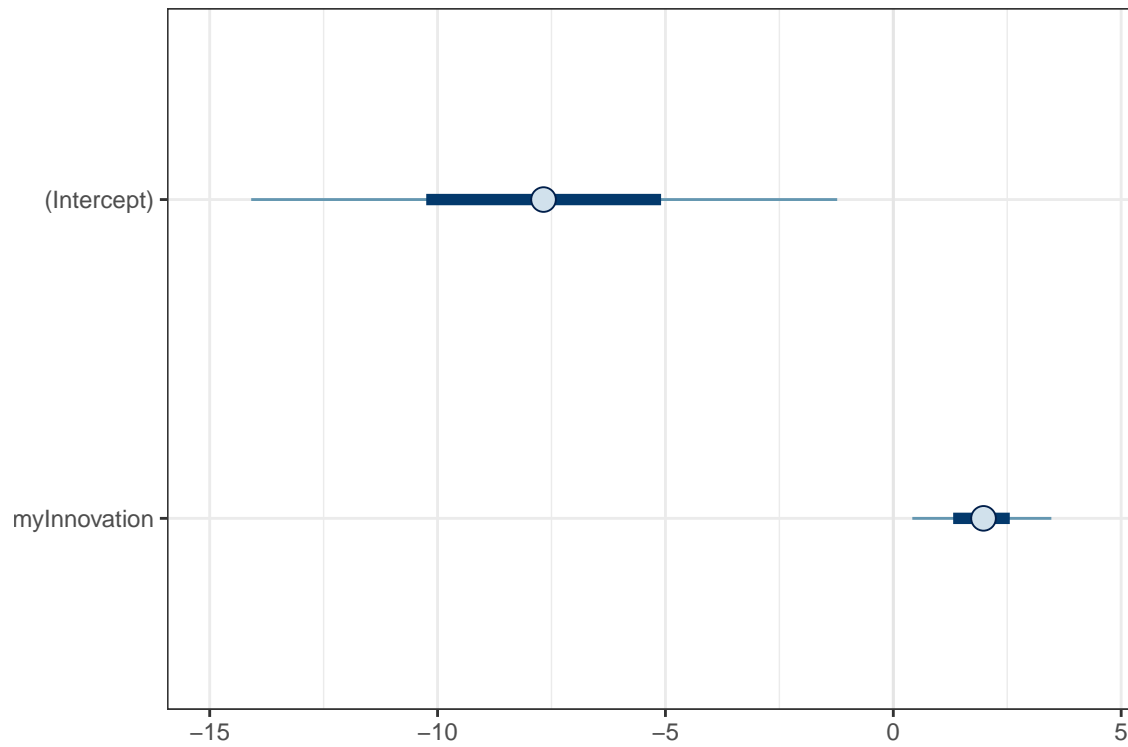
## stan_lm
## family:      gaussian [identity]
## formula:      SGR ~ myInnovation
## observations: 114
## predictors:   2
## -----
##              Median MAD_SD
## (Intercept)  -7.7      3.8
## myInnovation  2.0      0.9
##
## Auxiliary parameter(s):
##              Median MAD_SD
## R2              0.0      0.0
## log-fit_ratio   0.1      0.1
## sigma           14.7      1.0
##
## Sample avg. posterior predictive distribution of y:
##              Median MAD_SD
## mean_PPD 0.0      2.0
##
## -----
## * For help interpreting the printed output see ?print.stanreg
## * For info on the priors used see ?prior_summary.stanreg
```

## Create our posterior draws

```
posterior <- as.array(bayesian.model)
```

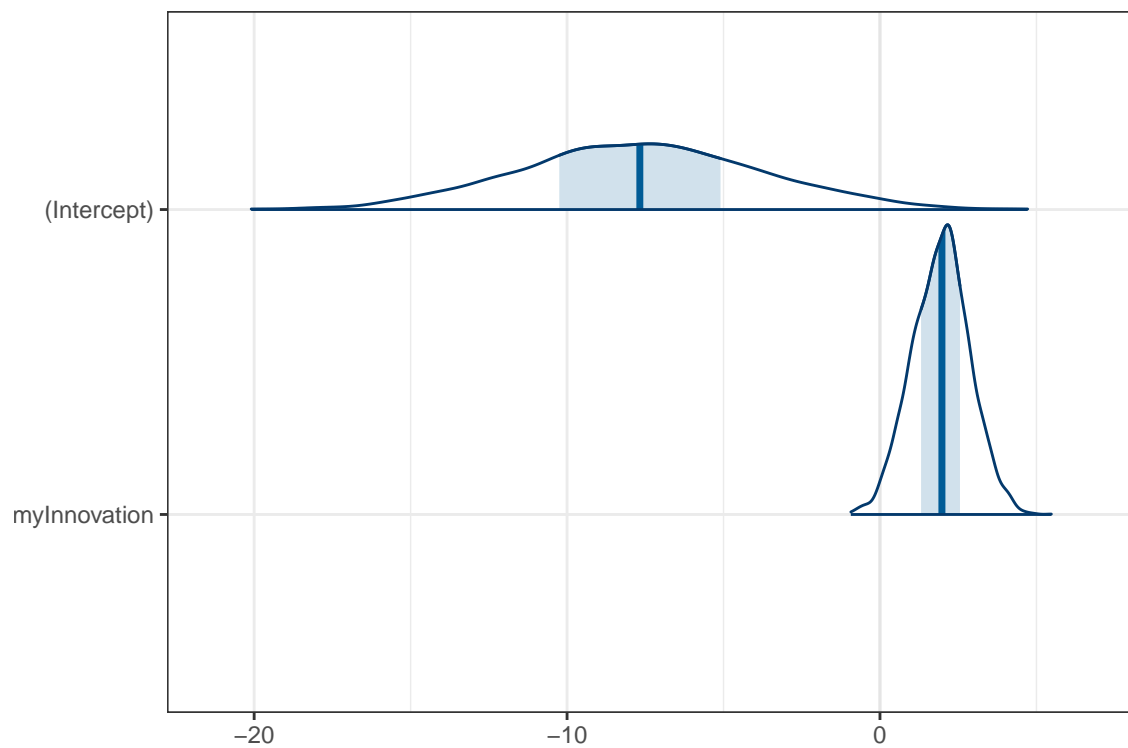
## Posterior uncertainty intervals

```
mcmc_intervals(posterior, pars = c("(Intercept)", "myInnovation")) +
  theme_bw()
```



### Posterior Area Density

```
mcmc_areas(posterior, pars = c("(Intercept)", "myInnovation")) +  
  theme_bw()
```



## Posterior Histogram

```
mcmc_hist(posterior, pars = c("(Intercept)", "myInnovation")) +  
  theme_bw()
```

