## STAT 447C: Bayesian Statistics

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## Exercise 4

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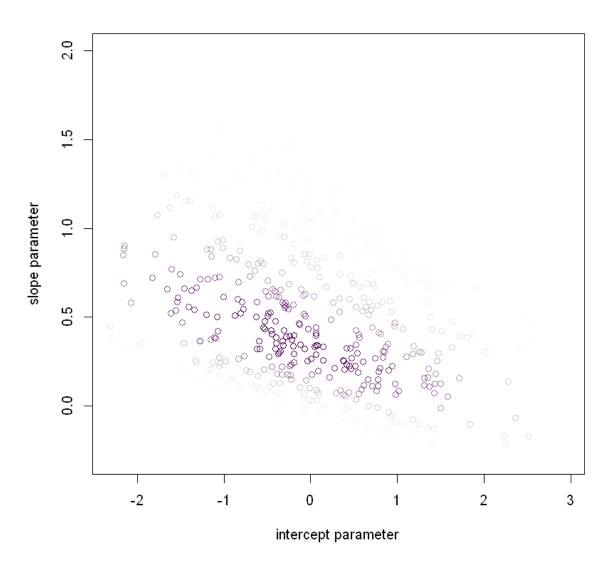
## 1 logistic rocket improvement

```
1 suppressPackageStartupMessages(library(extraDistr))
2 suppressPackageStartupMessages(library(distr))
source("./simple.R")
source("./simple_utils.R")
5 set.seed(2024)
7 success_indicators = c(1, 0, 1, 1, 0, 1, 1, 1, 1, 1)
9 # (1)
10 logistic_regression = function() {
n = length(success_indicators)
intercept = simulate(Norm(0, 1))

slope = simulate(Norm(0, 1))

thetas = plogis(intercept + slope * (1:n))
15
    for (i in (1:n)) {
     observe(success_indicators[i], Bern(thetas[i]))
16
17
    next_theta = intercept + slope * (n + 1)
18
pred = ifelse(next_theta >= 1 / 2, 1, 0)
return (c(intercept, slope, pred))
21 }
```

```
21 # (2)
2 posterior = posterior_particles(logistic_regression, 1000)
3 weighted_scatter_plot(posterior, plot_options = list(xlab="intercept parameter", ylab=" slope parameter"))
```



```
3_1 # (3)
post_obj = posterior(logistic_regression, 1000)
intercept = post_obj[1]
4 slope = post_obj[2]
5 plogis(slope * (length(success_indicators) + 1) + intercept)
6 # 0.990931632727317
41 # (4)
2 logistic_regression_2 = function() {
    n = length(success_indicators)
    intercept = simulate(Norm(0, 1))
    theta = plogis(intercept)
    for (i in (1:n)) {
       observe(success_indicators[i], Bern(theta))
    pred = ifelse(theta >= 1 / 2, 1, 0)
10
    return (c(intercept, pred))
11 }
plogis((posterior(logistic_regression_2, 1000))[1])
# 0.733947433558989
```

## 2 choosing a model

```
unified_model = function() {
  post_logis = logistic_regression()
  post_simple = logistic_regression_2()
  p_logis = plogis(post_logis[1] + post_logis[2] * length(success_indicators))
  p_simple = plogis(post_simple[1])
  return (ifelse(p_logis > p_simple, 1, 0))
}

posterior(unified_model, 1000)
# 0.955874166557784
```