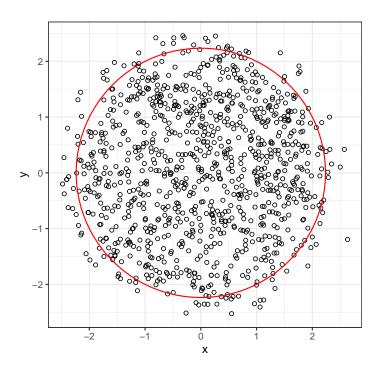
STA 602 Lab 7

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31 October, 2022

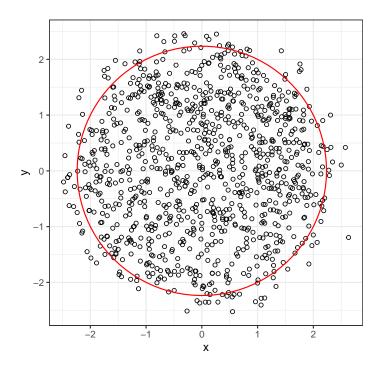
0. Visualize the data generating process



Exercise 1

```
# hyper-parameters
m <- 3
k <- 1
alpha \leftarrow 5/2
beta <-5/2
# generate random sample from pareto(m,k)
rpareto <- function(m, k, trunc = NULL){</pre>
  p \leftarrow m*(1 - runif(1))^(-1/k)
  if(!is.null(trunc)){
    while(p > trunc){
      p \leftarrow m*(1 - runif(1))^(-1/k)
    }
  }
  return(p)
}
uni_pareto_gibbs <- function(S, r, m, k, alpha, beta, burn_in = min(1000, S / 2), thin = 5){
  # Reparametrize X matrix to squared radius values
  Rsq <- r # squared distance
  n <- length(Rsq) # sample size</pre>
  R \leftarrow rep(1, S) # to save the S gibbs samples for R^2, also intialized it
  U \leftarrow matrix(0, nrow = S, ncol = n) # to save the S gibbs samples for u_1, \ldots, u_n
  U[1, ] \leftarrow runif(n, 0, R) # initialize the first sample for u_1, \ldots, u_n
  sigma \leftarrow rep(1, S) # to save the S gibbs samples for sigma (not sigma 2!)
```

```
U_{curr} \leftarrow U[1, ] # save the updated sample for u_1, \ldots, u_n
  R_{curr} \leftarrow R[1] # save the updated sample for R^2
  sigma_curr <- sigma[1] # save the updated sample for sigma (not sigma2!)</pre>
  for(s in 1:S){
    # Sample from full conditional of the inner radius
    R_curr <- rpareto(max(c(U_curr, m)), k + n)</pre>
    R[s] <- R curr
    # Sample from full conditional of U values
    U_curr <- truncnorm::rtruncnorm(n, a = 0, b = R_curr, mean = Rsq, sd = sigma_curr)
    U[s, ] <- U_curr</pre>
    # Sample from full conditional of sigma
    sigma_curr \leftarrow sqrt(1 / rgamma(1, n / 2 + alpha, 1/2 * sum((Rsq-U_curr)^2) + beta))
    sigma[s] <- sigma_curr</pre>
 return(list(R = R[seq(burn_in, S, by = thin)],
              U = U[seq(burn_in, S, by = thin),],
              sigma = sigma[seq(burn_in, S, by = thin)]))
}
gibbs_samps <- uni_pareto_gibbs(S = 100000, r, m, k, alpha, beta, burn_in=2000)
ggplot2::ggplot() +
  geom\_point(data = data.frame(x = sign(r)*sqrt(abs(r))*cos(theta), y = sign(r)*sqrt(abs(r))*sin(theta)
             aes(x = x, y = y), shape = 1) +
  geom_path(data = data.frame(R = gibbs_samps$R) %>%
                               plyr::ddply(~R, function(d){
                                 data.frame(x = sqrt(dR)*cos(seq(0, 2*pi, length.out = 100)),
                                             y = sqrt(dR)*sin(seq(0, 2*pi, length.out = 100)))
                               }),
                        aes(x = x, y = y), alpha = 0.005, colour = "blue") +
  geom_path(data = data.frame(R = true_Rsquared) %>%
                               plyr::ddply(~(R), function(d){
                                 data.frame(x = sqrt(dR)*cos(seq(0, 2*pi, length.out = 100)),
                                             y = sqrt(dR)*sin(seq(0, 2*pi, length.out = 100)))
                               }),
                        aes(x = x, y = y), alpha = 1, colour = "red") +
  coord fixed()
```

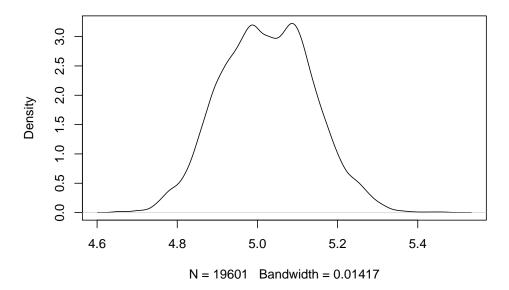


Exercise 2 & 3

Considering the true \mathbb{R}^2 is 5 and true σ^2 is 1.5625, both marginal posterior densities are pretty reasonable.

plot(density(gibbs_samps\$R))

density.default(x = gibbs_samps\$R)

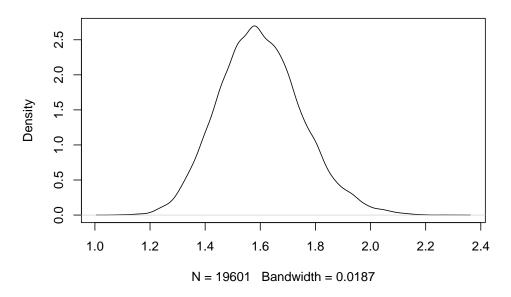


```
true_Rsquared
```

```
## [1] 5
```

```
plot(density((gibbs_samps$sigma)^2))
```

$density.default(x = (gibbs_sampssigma)^2)$



true_sigma^2

[1] 1.5625

Exercise 4

The true values of both parameters are at the center of the contour plot.

```
ggplot() +
  geom_bin2d(aes(x=gibbs_samps$R, y=(gibbs_samps$sigma)^2), bins = 70) +
  geom_point(aes(x=true_Rsquared, y=true_sigma^2), size = 5, color = "red") +
  annotate(
  "text", label = "True Values",
  x = true_Rsquared*1.01, y = true_sigma^2*0.95, size = 3, colour = "red"
) +
  scale_fill_continuous(type = "viridis")
```

