

# ps5\_code

AUTHOR

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```
library("patchwork")
library("tidyverse")
```

## 7.9

```
one_mh_iteration <- function(current, w, obs_value, tau, sigma){
  # Step 1: propose next location in chain
  proposal <- runif(1, min = current - w, current + w)

  # Step 2: decide whether or not to go there
  proposal_plausibility <- dnorm(proposal, 3, tau) * dnorm(obs_value, proposal, sigma)
  current_plausibility <- dnorm(current, 3, tau) * dnorm(obs_value, current, sigma)
  alpha <- min(1, proposal_plausibility / current_plausibility)
  next_stop <- sample(c(proposal, current),
                     size = 1, prob = c(alpha, 1 - alpha))

  # Return the results as a data frame
  return(data.frame(proposal, alpha, next_stop))
}
```

a

---

```
set.seed(320)
one_mh_iteration(w = 0.01, current = 3,
                 obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	2.990375	0.9457828	2.990375

There was about a 95% acceptance probability, and the proposed location was accepted.

b

---

```
set.seed(320)
one_mh_iteration(w = 0.5, current = 3,
                 obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	2.518728	0.04494218	3

There was about a 4% acceptance probability, and the proposed location was rejected.

## C

```
set.seed(320)
one_mh_iteration(w = 1, current = 3,
                 obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	2.037457	0.001061415	3

There was a virtually zero acceptance probability, and the proposed location was rejected.

## d

```
set.seed(320)
one_mh_iteration(w = 3, current = 3,
                 obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	0.1123701	5.303685e-13	3

There was a virtually zero acceptance probability, and the proposed location was rejected.

## 7.10

```
mh_tour <- function(N, current, w, obs_value, tau, sigma){
  # N: chain length
  # initialize vector
  mu <- rep(0, N)

  # simulate N Markov chain stops
  for(i in 1:N){
    # simulate one iteration
    this_iteration <- one_mh_iteration(current, w, obs_value, tau, sigma)

    # record next location
    mu[i] <- this_iteration$next_stop

    # update current location
    current <- this_iteration$next_stop
  }

  # return the chain locations
  return(data.frame(iteration = c(1:N), mu))
}
```

```
bayesrules::summarize_normal_normal(
```

```
# from prior
mean = 3, sd = 1,

# from observations
y_bar = 6.25, sigma = 0.75, n = 1
) |>
mutate_if(is.numeric, round, digits = 4)
```

	model	mean	mode	var	sd
1	prior	3.00	3.00	1.00	1.0
2	posterior	5.08	5.08	0.36	0.6

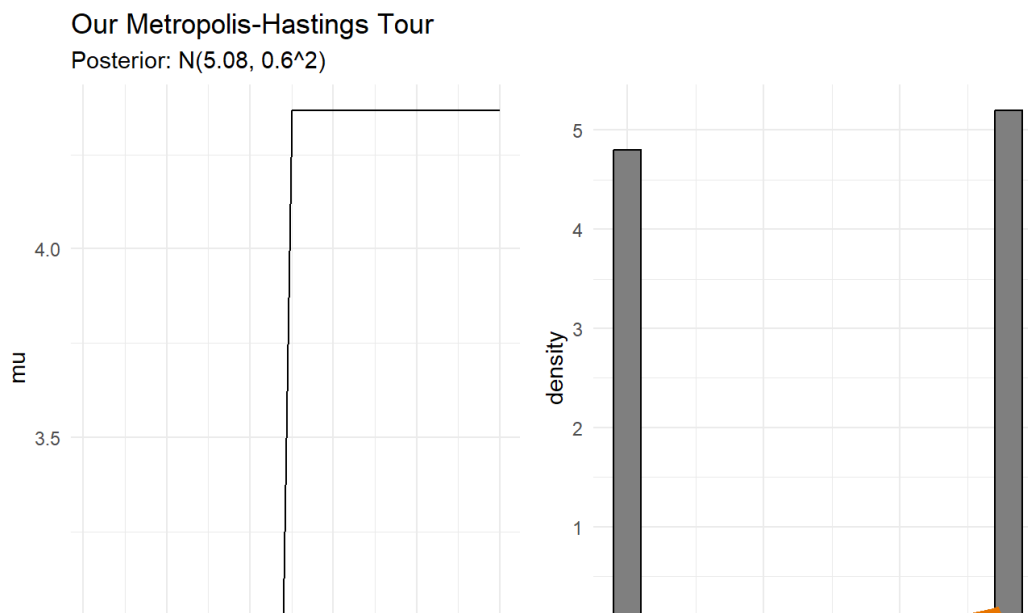
a

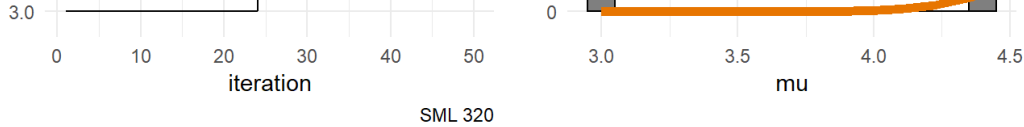
```
set.seed(320)
our_mh_tour <- mh_tour(N = 50, current = 3, w = 50,
                      obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
       subtitle = "Posterior: N(5.08, 0.6^2)",
       caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
               binwidth = 0.1,
               color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
               color = "#E77500",
               linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2
```





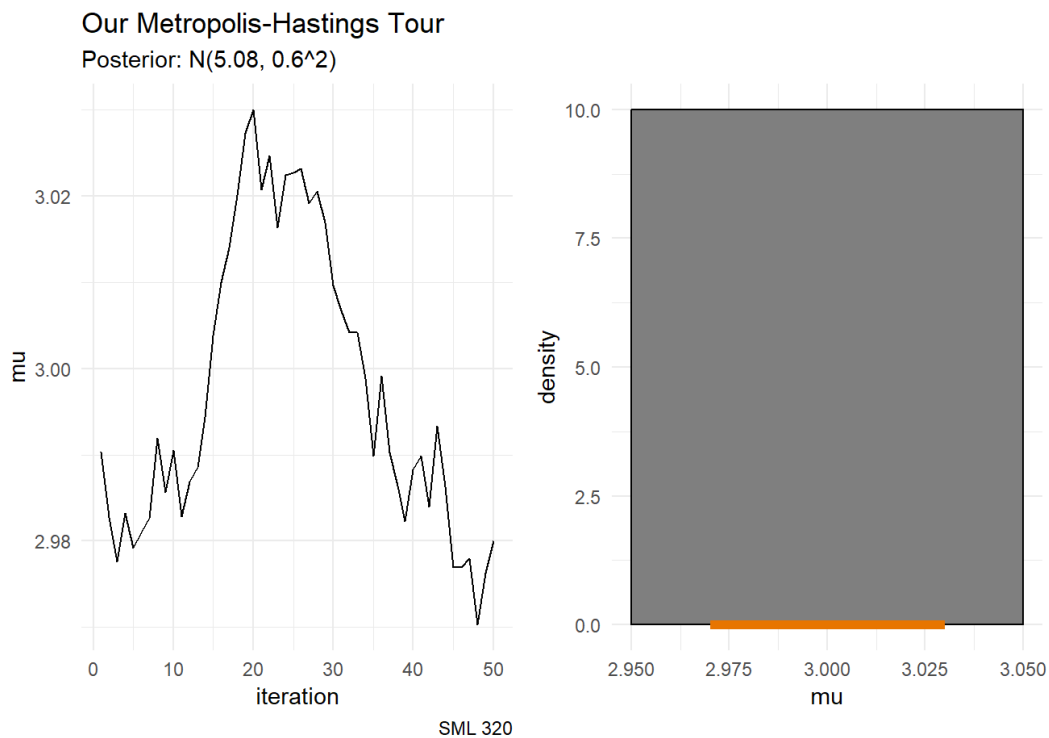
b

```
set.seed(320)
our_mh_tour <- mh_tour(N = 50, current = 3, w = 0.01,
  obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
    subtitle = "Posterior: N(5.08, 0.6^2)",
    caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
    binwidth = 0.1,
    color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
    color = "#E77500",
    linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2
```



c

```

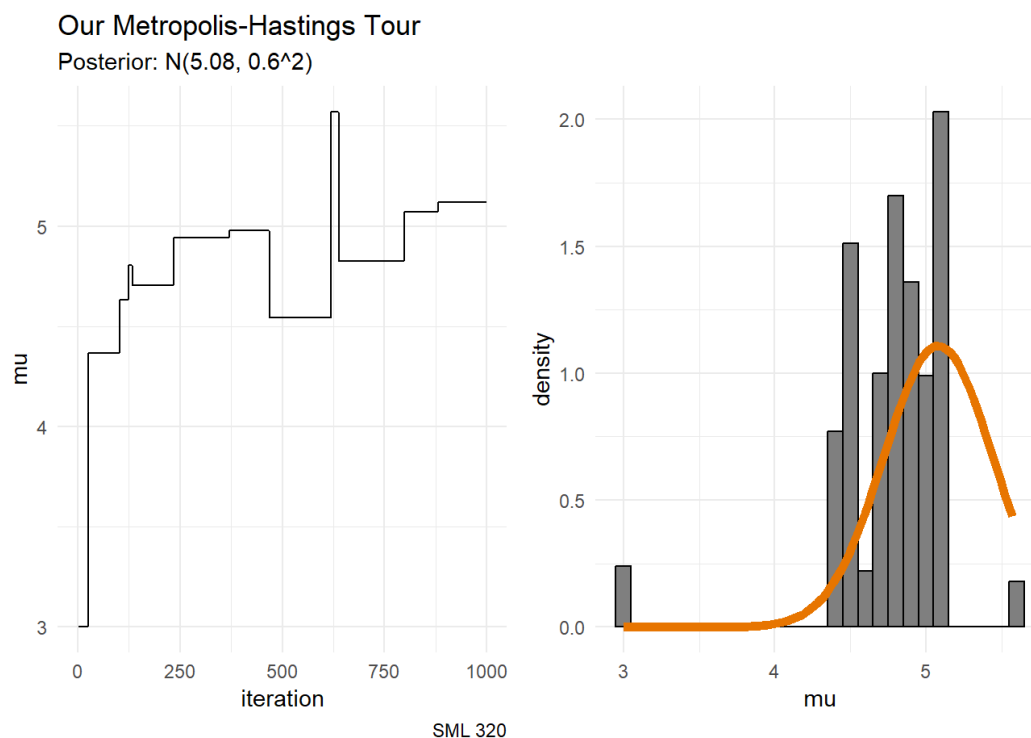
set.seed(320)
our_mh_tour <- mh_tour(N = 1000, current = 3, w = 50,
  obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
    subtitle = "Posterior: N(5.08, 0.6^2)",
    caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
    binwidth = 0.1,
    color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
    color = "#E77500",
    linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2

```



d

```

set.seed(320)
our_mh_tour <- mh_tour(N = 1000, current = 3, w = 1000,
  obs_value = 6.25, tau = 1, sigma = 0.75)

```

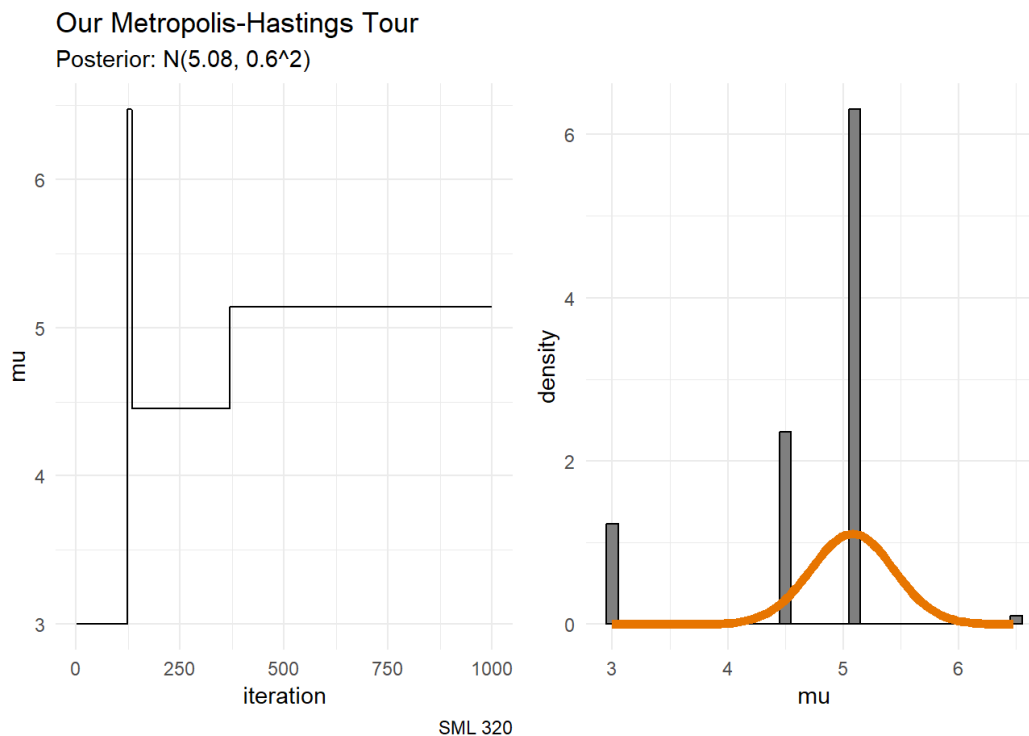
```

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
       subtitle = "Posterior: N(5.08, 0.6^2)",
       caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
               binwidth = 0.1,
               color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
               color = "#E77500",
               linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2

```



e

Comparing the results from parts (a) and (b), we observe that having a half-width  $w$  that is too large can create a situation that is nearly constant because of the very low acceptance probabilities.

f

Comparing the results from parts (a) and (b), we observe that having a half-width  $w$  that is too large can create situation that are nearly constant for several iterations

because of the very low acceptance probabilities.

## 7.11

```
one_mh_iteration_normal <- function(current, s, obs_value, tau, sigma){  
  # Step 1: propose next location in chain  
  proposal <- current + rnorm(1, mean = 0, sd = s)  
  
  # Step 2: decide whether or not to go there  
  proposal_plausibility <- dnorm(proposal, 3, tau) * dnorm(obs_value, proposal, tau, sigma)  
  current_plausibility <- dnorm(current, 3, tau) * dnorm(obs_value, current, tau, sigma)  
  alpha <- min(1, proposal_plausibility / current_plausibility)  
  next_stop <- sample(c(proposal, current),  
                     size = 1, prob = c(alpha, 1 - alpha))  
  
  # Return the results as a data frame  
  return(data.frame(proposal, alpha, next_stop))  
}
```

a

---

```
set.seed(320)  
one_mh_iteration_normal(s = 0.01, current = 3,  
                        obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	2.979192	0.8861912	2.979192

The acceptance probability was about 87%, and the proposed location was accepted.

b

---

```
set.seed(320)  
one_mh_iteration_normal(s = 0.5, current = 3,  
                        obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	1.959624	0.0005451953	3

There was a virtually zero acceptance probability, and the proposed location was rejected.

c

---

```
set.seed(320)  
one_mh_iteration_normal(s = 1, current = 3,  
                        obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	0.9192488	1.470104e-08	3

There was a virtually zero acceptance probability, and the proposed location was rejected.

d

---

```
set.seed(320)
one_mh_iteration_normal(s = 3, current = 3,
                        obs_value = 6.25, tau = 1, sigma = 0.75)
```

	proposal	alpha	next_stop
1	-3.242254	6.807206e-40	3

There was a virtually zero acceptance probability, and the proposed location was rejected.

## 7.12

```
mh_tour_normal <- function(N, current, s, obs_value, tau, sigma){
  # N: chain length
  # initialize vector
  mu <- rep(0, N)

  # simulate N Markov chain stops
  for(i in 1:N){
    # simulate one iteration
    this_iteration <- one_mh_iteration_normal(current, s,
                                              obs_value, tau, sigma)

    # record next location
    mu[i] <- this_iteration$next_stop

    # update current location
    current <- this_iteration$next_stop
  }

  # return the chain locations
  return(data.frame(iteration = c(1:N), mu))
}
```

a

---

```
set.seed(320)
our_mh_tour <- mh_tour_normal(N = 20, current = 3, s = 0.01,
                              obs_value = 6.25, tau = 1, sigma = 0.75)
```



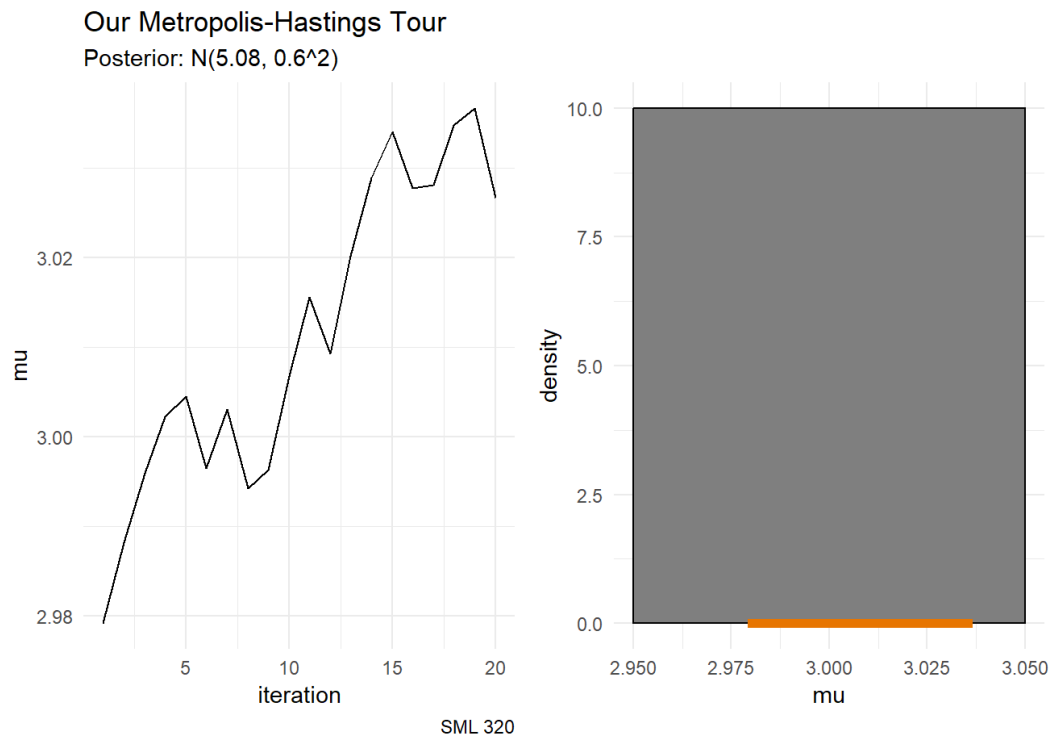
```

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
        subtitle = "Posterior: N(5.08, 0.6^2)",
        caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
                binwidth = 0.1,
                color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
                color = "#E77500",
                linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2

```



b

```

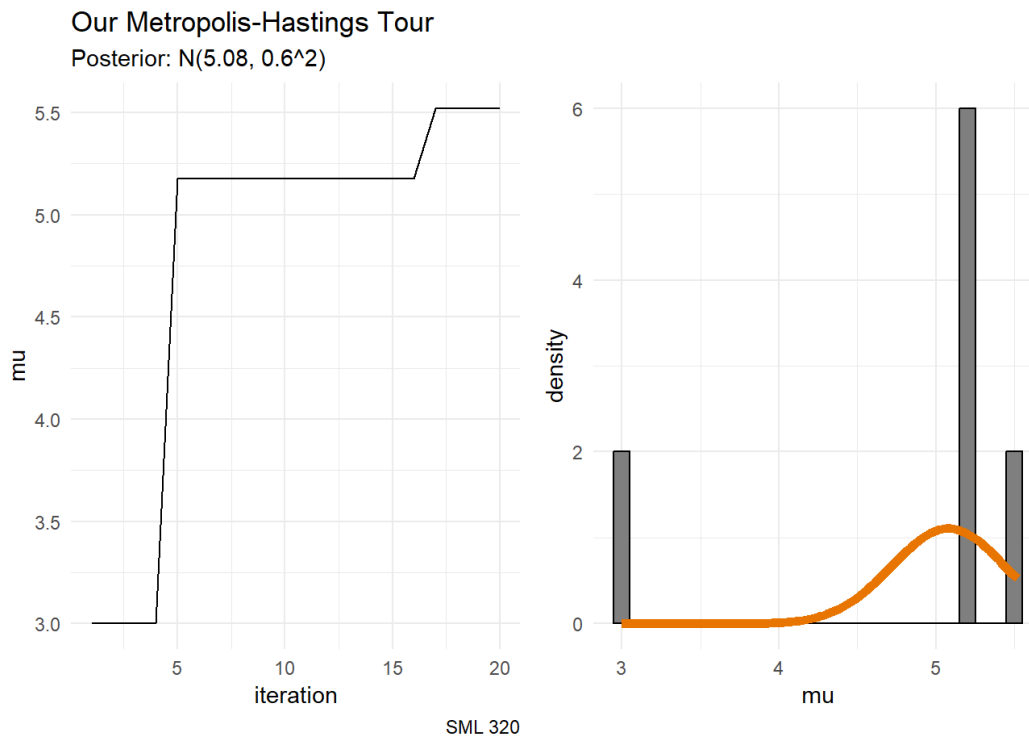
set.seed(320)
our_mh_tour <- mh_tour_normal(N = 20, current = 3, s = 10,
                              obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
        subtitle = "Posterior: N(5.08, 0.6^2)",
        caption = "SML 320") +
  theme_minimal()

```

```
p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
    binwidth = 0.1,
    color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
    color = "#E77500",
    linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2
```



C

```
set.seed(320)
our_mh_tour <- mh_tour_normal(N = 1000, current = 3, s = 0.01,
  obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
    subtitle = "Posterior:  $N(5.08, 0.6^2)$ ",
    caption = "SML 320") +
  theme_minimal()

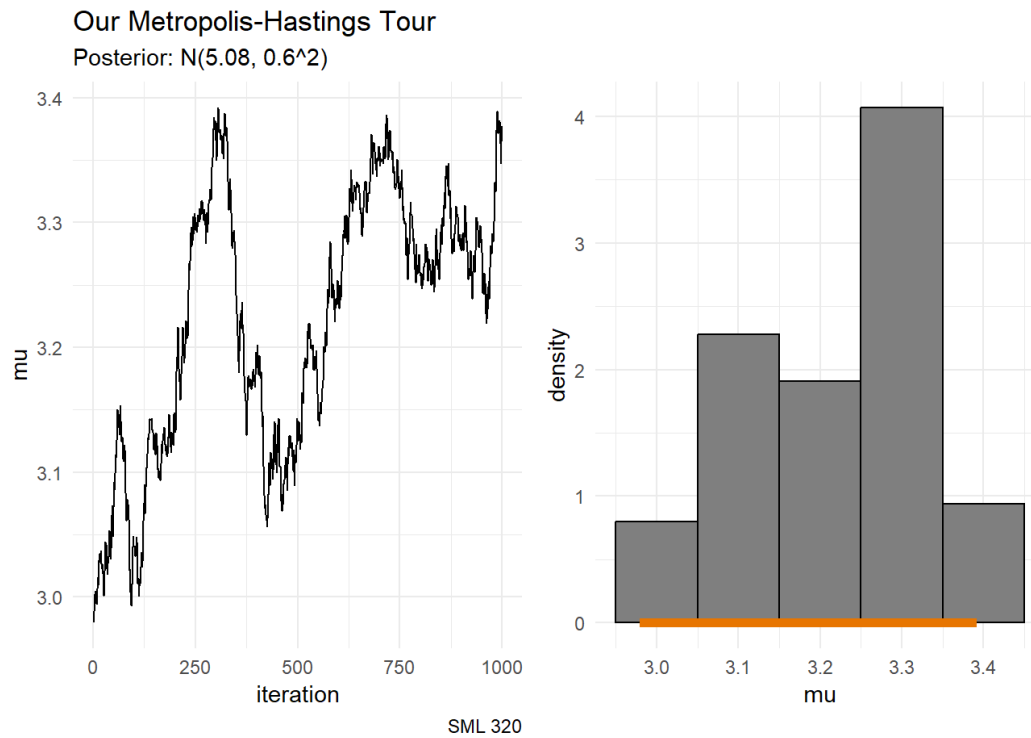
p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
    binwidth = 0.1,
    color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
    color = "#E77500",
```

```

        linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2

```



d

```

set.seed(320)
our_mh_tour <- mh_tour_normal(N = 1000, current = 3, s = 10,
  obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
    subtitle = "Posterior:  $N(5.08, 0.6^2)$ ",
    caption = "SML 320") +
  theme_minimal()

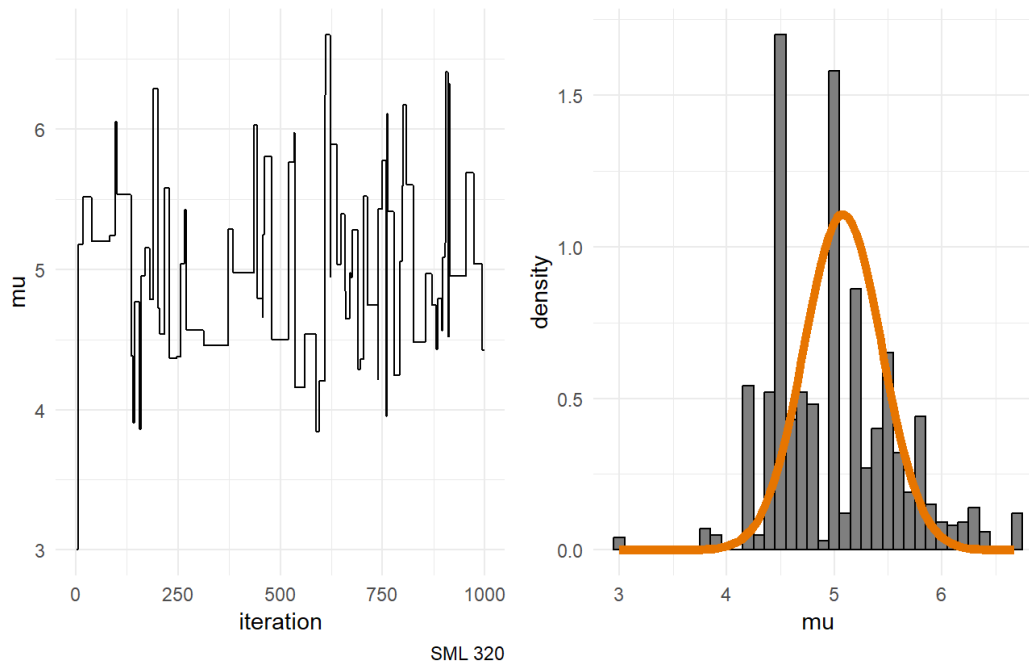
p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
    binwidth = 0.1,
    color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
    color = "#E77500",
    linewidth = 2) +
  theme_minimal()

# patchwork
p1 + p2

```

## Our Metropolis-Hastings Tour

Posterior:  $N(5.08, 0.6^2)$



e

Comparing the results from parts (a) and (b), we observe that having a standard deviation  $s$  that is too large can create a situation that is nearly constant because of the very low acceptance probabilities.

f

Perhaps we should try a standard deviation that is closer to those in the prior and likelihood models. Here, with  $s = 0.8$ ,

```
set.seed(320)
our_mh_tour <- mh_tour_normal(N = 1000, current = 3, s = 0.8,
                             obs_value = 6.25, tau = 1, sigma = 0.75)

p1 <- ggplot(our_mh_tour, aes(x = iteration, y = mu)) +
  geom_line() +
  labs(title = "Our Metropolis-Hastings Tour",
        subtitle = "Posterior:  $N(5.08, 0.6^2)$ ",
        caption = "SML 320") +
  theme_minimal()

p2 <- ggplot(our_mh_tour, aes(x = mu)) +
  geom_histogram(aes(y = after_stat(density)),
                 binwidth = 0.1,
                 color = "black", fill = "gray50") +
  stat_function(fun = dnorm, args = list(5.08, 0.36),
               color = "#E77500",
               linewidth = 2) +
  theme_minimal()
```

```
# patchwork  
p1 + p2
```

### Our Metropolis-Hastings Tour

Posterior:  $N(5.08, 0.6^2)$

