

Chapter 7

Question 2

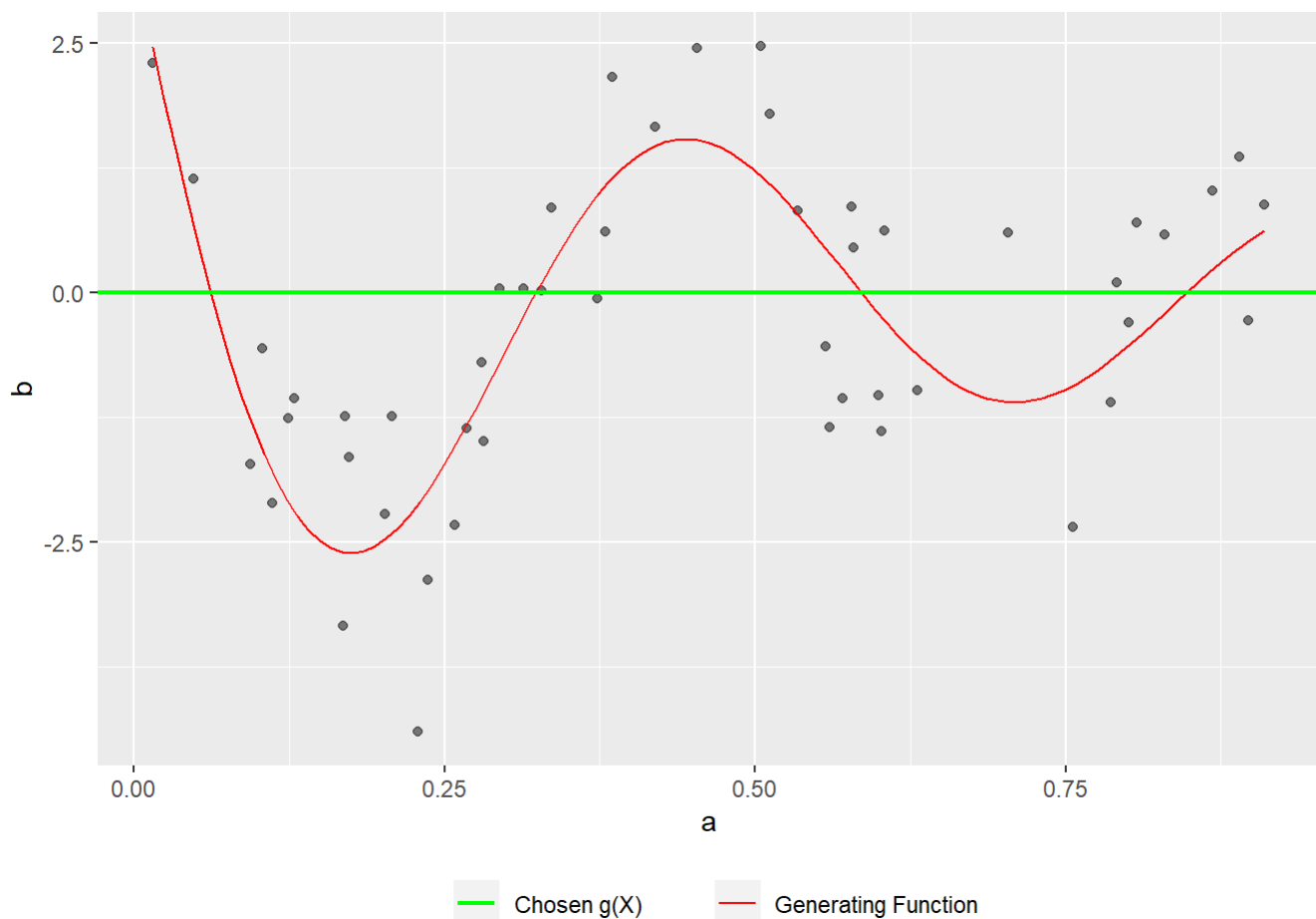
We need to generate some data before sketching \hat{g} under different conditions!

```
library(ggplot2)
set.seed(3)

a <- runif(50)
eps <- rnorm(50)
b <- sin(12*(a + 0.2)) / (a + 0.2) + eps
generating_fn <- function(a) {sin(12*(a + 0.2)) / (a + 0.2)}
df <- data.frame(a, b)
```

2a) $\lambda = \infty, m=0$

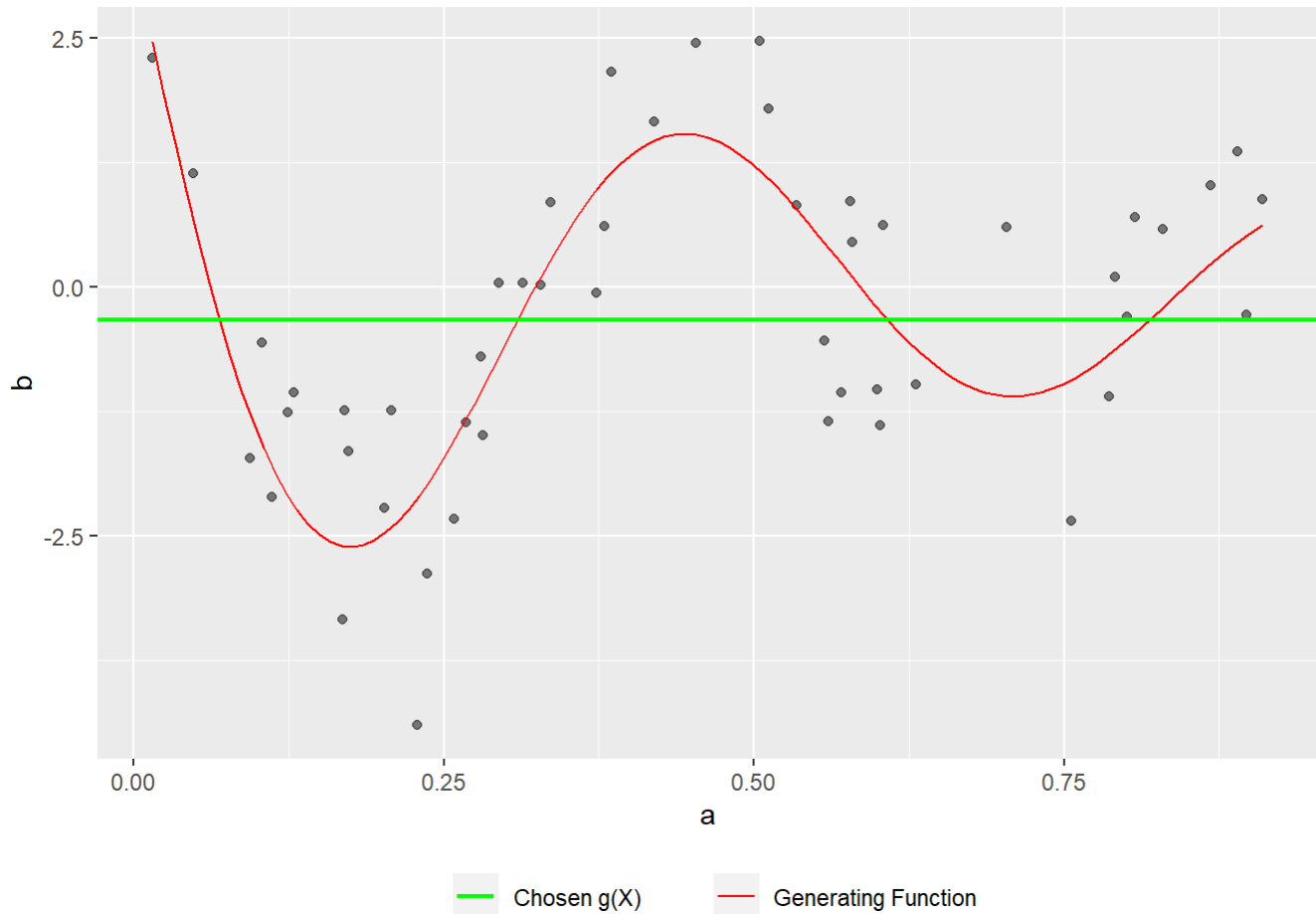
```
ggplot(df, aes(x = a, y = b)) +
  geom_point(alpha = 0.5) +
  stat_function(fun = generating_fn, aes(col = "Generating Function")) +
  geom_hline(aes(yintercept = 0, linetype = "Chosen g(X)", col = "green", size = 0.8) +
  scale_color_manual(values = "red") +
  theme(legend.position = "bottom", legend.title = element_blank())
```



As λ increases, the penalty term becomes more and more important in the equation. As $\lambda \rightarrow \infty$, this forces $g(x) \rightarrow 0$. We therefore get $\hat{g}(x) = 0$

2b) $\lambda=\infty, m=1$

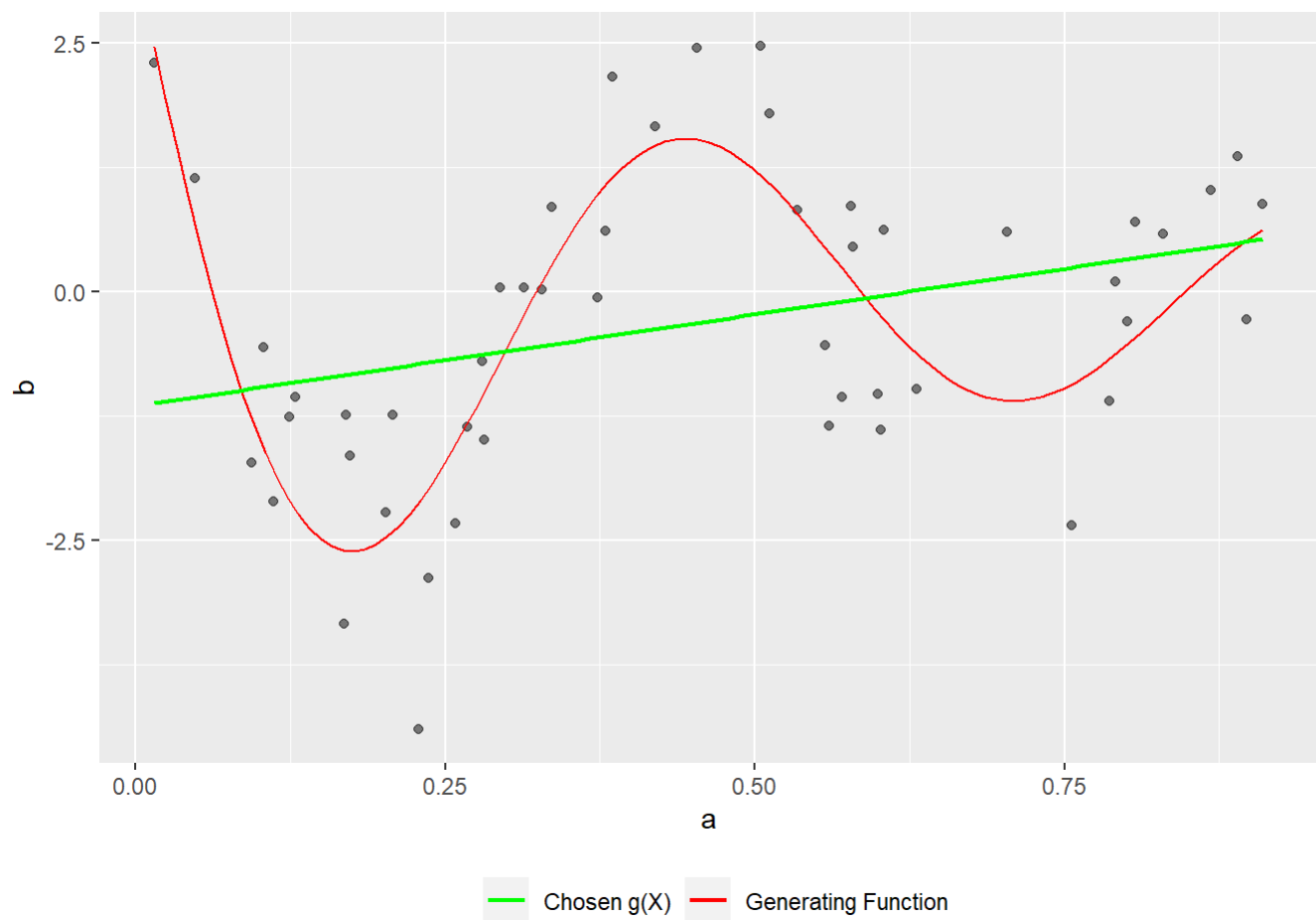
```
ggplot(df, aes(x = a, y = b)) +
  geom_point(alpha = 0.5) +
  stat_function(fun = generating_fn, aes(col = "Generating Function")) +
  geom_hline(aes(yintercept = mean(b)), linetype = "Chosen g(X)", col = "green", size = 0.8) +
  scale_color_manual(values = "red") +
  theme(legend.position = "bottom", legend.title = element_blank())
```



As $\lambda \rightarrow \infty$, this forces $g'(x) \rightarrow 0$. This means we would get $\hat{g}(x) = c$.

2c) $\lambda=\infty, m=2$

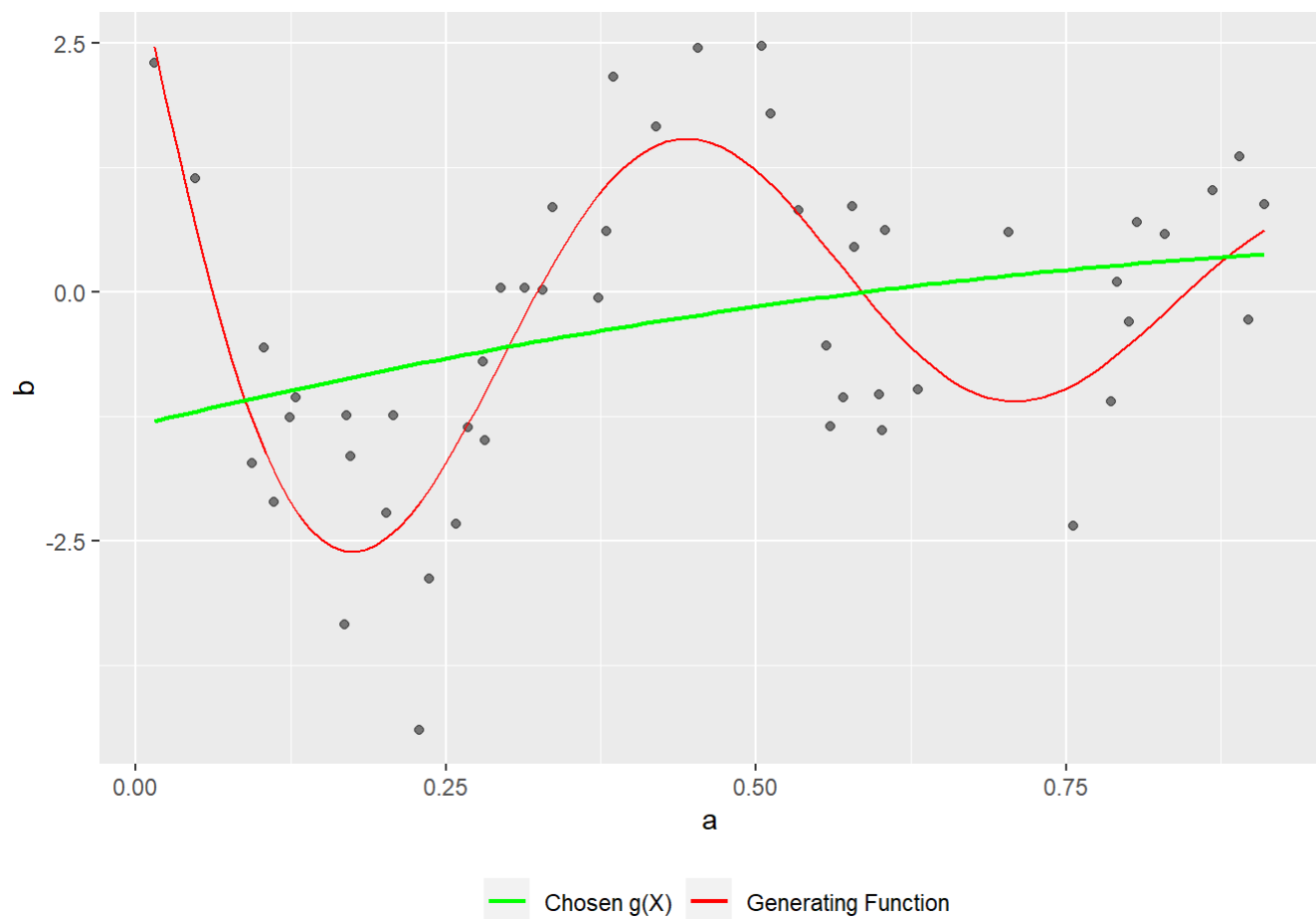
```
ggplot(df, aes(x = a, y = b)) +
  geom_point(alpha = 0.5) +
  stat_function(fun = generating_fn, aes(col = "Generating Function")) +
  geom_smooth(method = "lm", formula = "y ~ x", se = F, size = 0.8, aes(col = "Chosen g(X)")) +
  scale_color_manual(values = c("green", "red")) +
  theme(legend.position = "bottom", legend.title = element_blank())
```



As $\lambda \rightarrow \infty$, this forces $g''(x) \rightarrow 0$. This means we would get $\hat{g}(x) = ax + b$

2d) $\lambda = \infty, m = 3$

```
ggplot(df, aes(x = a, y = b)) +
  geom_point(alpha = 0.5) +
  stat_function(fun = generating_fn, aes(col = "Generating Function")) +
  geom_smooth(method = "lm", formula = "y ~ x + I(x^2)", se = F, size = 0.8, aes(col = "Chosen g
(X)")) +
  scale_color_manual(values = c("green", "red")) +
  theme(legend.position = "bottom", legend.title = element_blank())
```

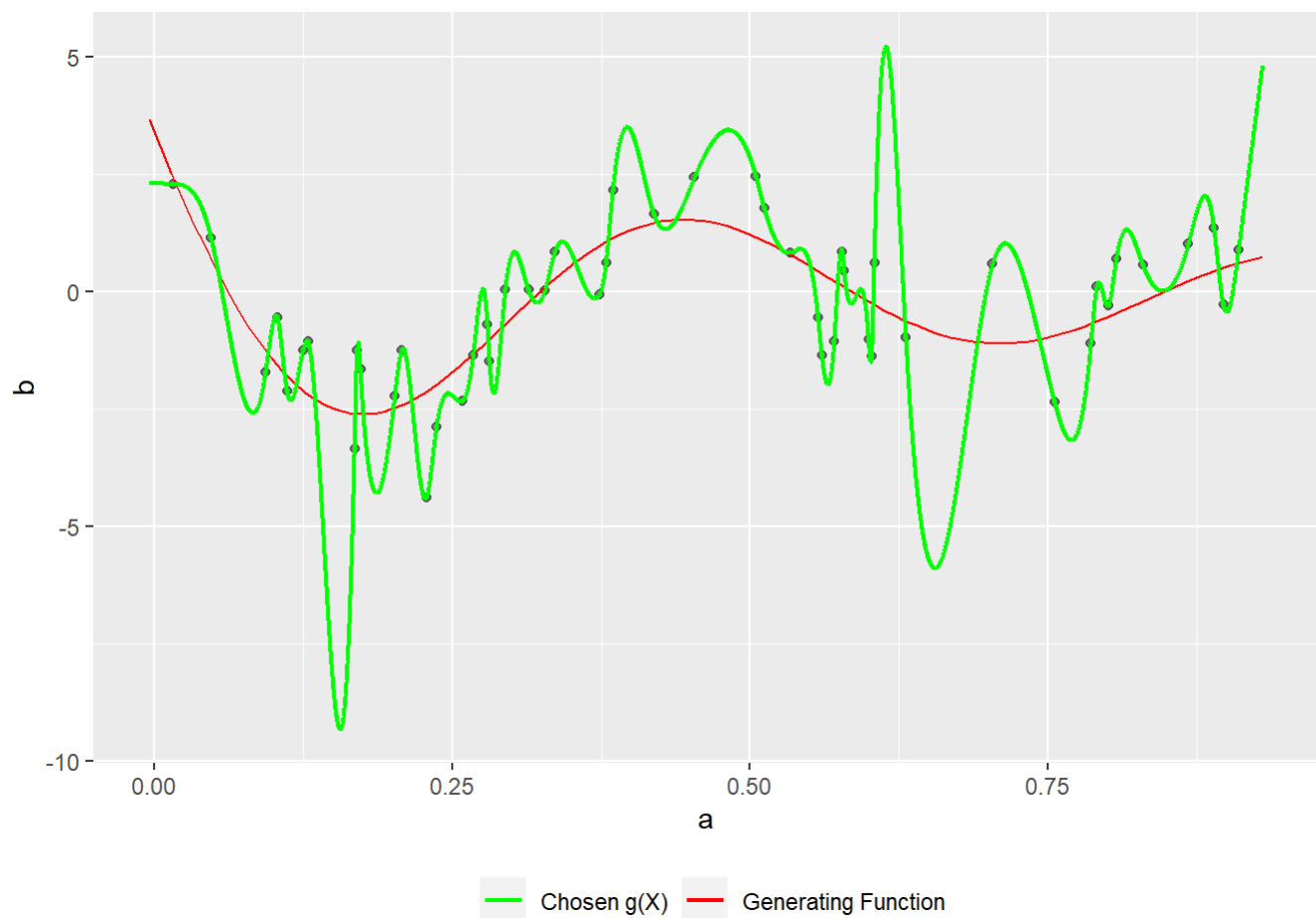


As $\lambda \rightarrow \infty$, this forces $g(3)(x) \rightarrow 0$. This means we would get $\hat{g}(x) = ax^2 + bx + c$.

2e) $\lambda=0, m=3$

```
interp_spline <- smooth.spline(x = df$a, y = df$b, all.knots = T, lambda = 0.0000000000001)
fitted <- predict(interp_spline, x = seq(min(a) - 0.02, max(a) + 0.02, by = 0.0001))
fitted <- data.frame(x = fitted$x, fitted_y = fitted$y)

ggplot(df, aes(x = a, y = b)) +
  geom_point(alpha = 0.5) +
  stat_function(fun = generating_fn, aes(col = "Generating Function")) +
  geom_line(data = fitted,
            aes(x = x, y = fitted_y, col = "Chosen g(X)"), size = 0.8) +
  scale_color_manual(values = c("green", "red")) +
  theme(legend.position = "bottom", legend.title = element_blank())
```

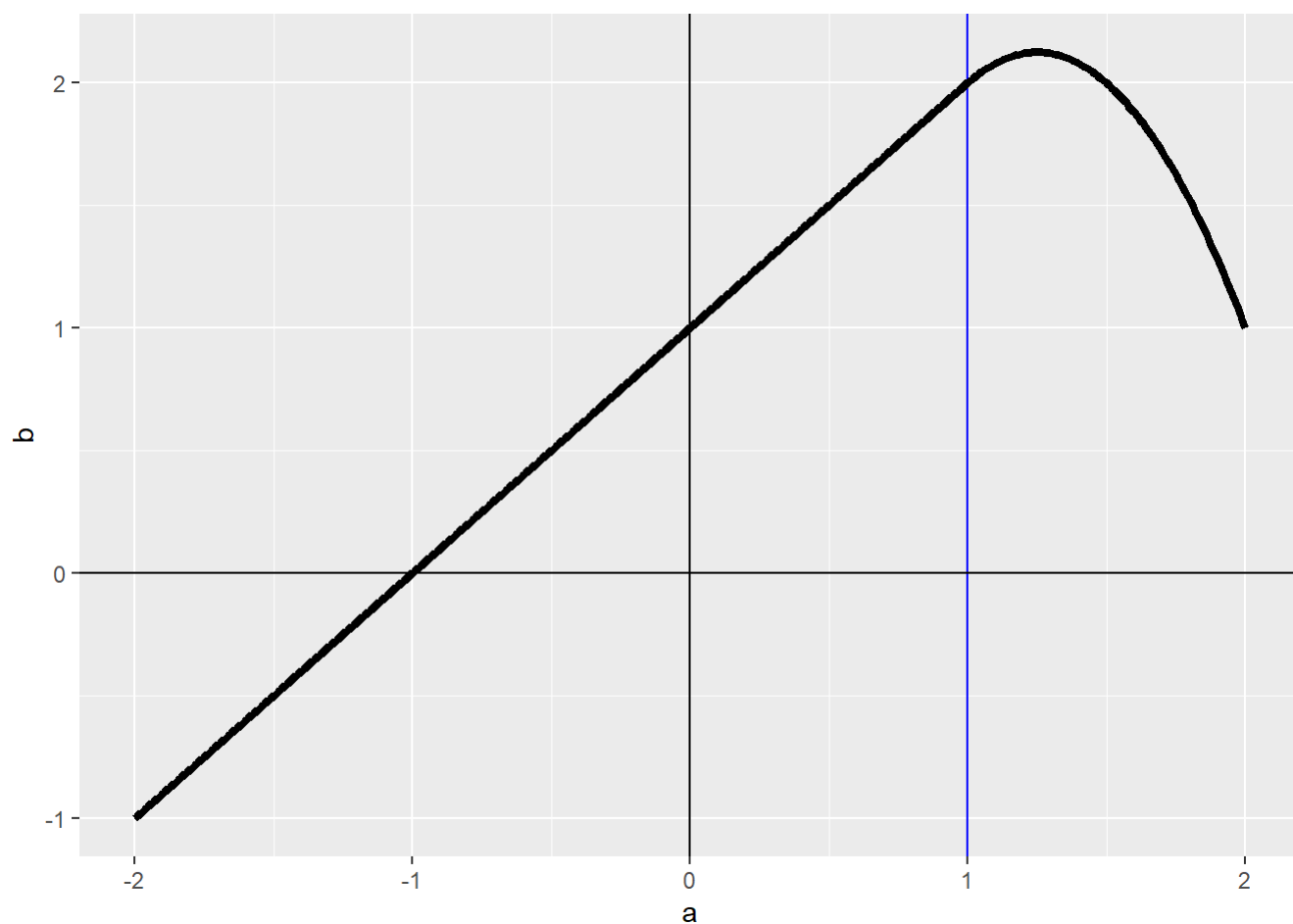


However, since $\lambda=0$, the penalty term no longer plays any role in the selection of $\hat{g}(x)$. For this reason, we can achieve $RSS = 0$

Question 3

```
a = seq(-2, 2, 0.01)
b = 1 + a + -2 * (a - 1)^2 * (a >= 1)
df <- data.frame(a, b)

ggplot(df, aes(x = a, y = b)) +
  geom_vline(xintercept = 0) +
  geom_vline(xintercept = 1, col = "blue") +
  geom_hline(yintercept = 0) +
  geom_line(size = 1.5)
```

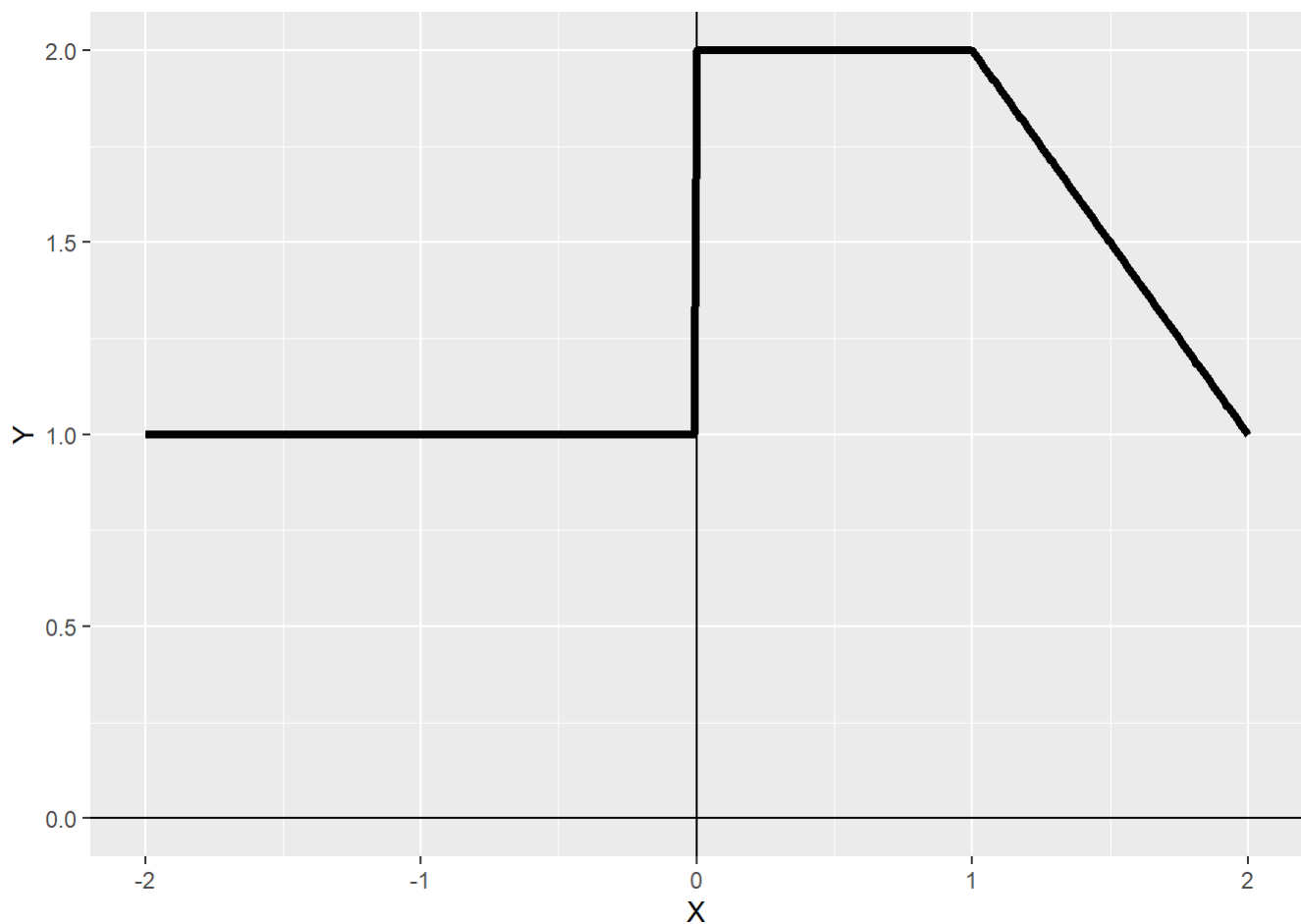


The curve is linear between -2 and 1 with $y=1+x$ Quadratic between 1 , and 2 with $y=1+x-2(x-1)^2$

Question 4

```
X = seq(-2, 2, 0.01)
Y = 1 + (X >= 0 & X <= 2) - (X - 1)*(X >= 1 & X <= 2) + 3*(X - 3)*(X >= 3 & X <= 4) + 3*(X > 4 &
X <= 5)
df <- data.frame(X, Y)

ggplot(df, aes(x = X, y = Y)) +
  geom_vline(xintercept = 0) +
  geom_hline(yintercept = 0) +
  geom_line(size = 1.5)
```



The curve is constant between -2 and 0 with $y=1$ Constant between 0 and 1 with $y=2$ Linear between 1 and 2 with $y=3-x$.

Question 5

- As $\lambda \rightarrow \infty$, will g_1 or g_2 have the smaller training RSS? Answer:- The smoothing spline g_2 will most likely have the smaller training RSS since it is a higher order polynomial owing to the penalty term's order (it will be more flexible).
- As $\lambda \rightarrow \infty$, will g_1 or g_2 have the smaller test RSS? Answer:- The test RSS will depend on the distribution of test data. If we have to provide the behavior of test RSS based on the nature of curve, g_2 will have more test RSS as it is more flexible and hence may overfit the data.
- For $\lambda = 0$, will g_1 or g_2 have the smaller training and test RSS? Answer:- If $\lambda=0$, we have $g_1=g_2$, so they will have the same training and test RSS.