

# Homework 2

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## Exercise 1

**i  $\iff$  iii**

we want to show that i  $\iff$  iii.

We follow the hint.

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(\eta)(x - \eta)^2$$

We then see that  $f''(\eta) \geq 0$  means that the last term in the right-hand side is non-negative. Thus,  $f(x) \geq f(x_0) + f'(x_0)(x - x_0)$ . For the reverse implication, reverse the above steps.

We thus see that i  $\iff$  iii.

**ii  $\implies$  iii**

We follow the hint.

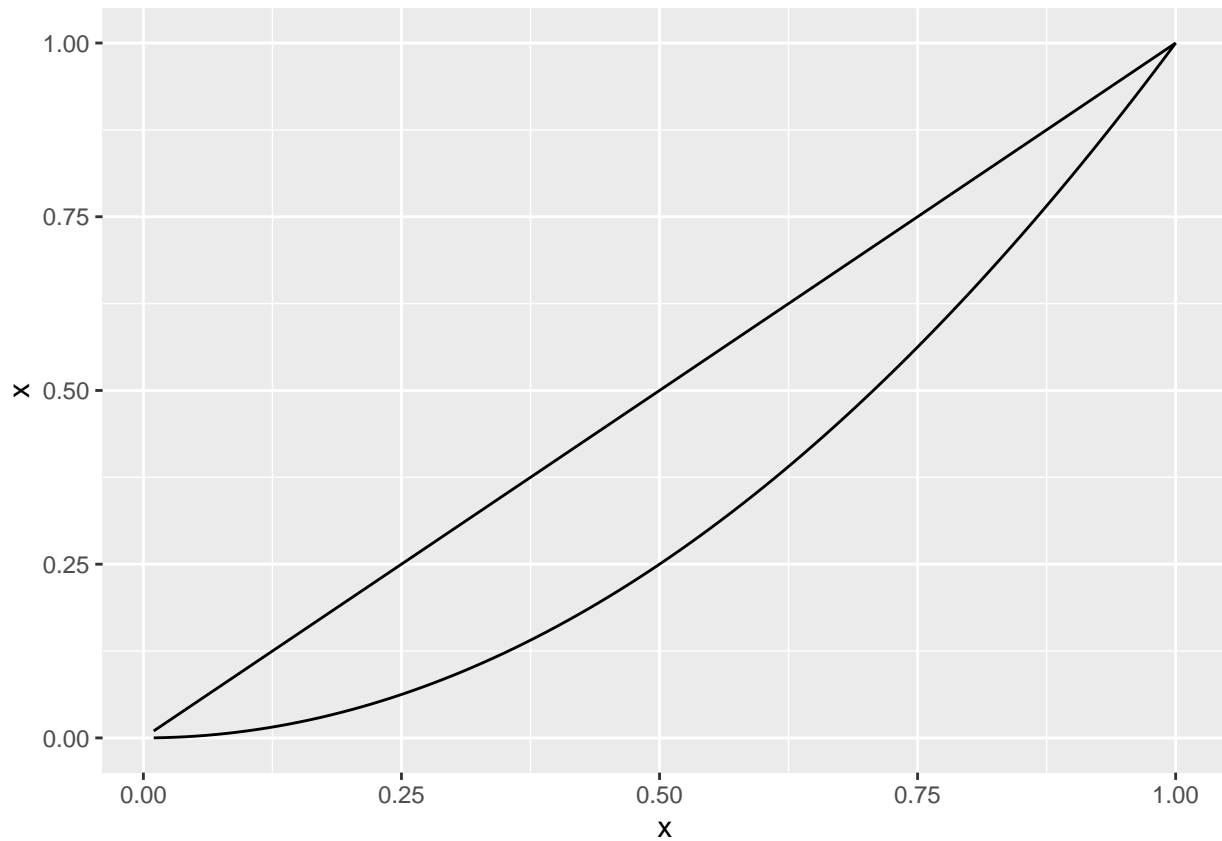
$$\begin{aligned} f(x + h(x - x_0)) &= f((1 + h)x - hx_0) \\ &= f((1 - (-h))x + (-h)x_0) \\ &\leq -hf(x_0) + (1 + h)f(x) \\ &= f(x) + h[f(x) - f(x_0)] \\ &\implies f(x + h(x - x_0)) \leq h[f(x) - f(x_0)] \\ &\implies \frac{f(x + h(x - x_0))}{h} \leq f(x) - f(x_0) \\ &\implies f'(x_0)(x - x_0) \leq f(x) - f(x_0) \end{aligned}$$

**iii  $\implies$  ii**

### Plot for ii

We let  $f(x) = x^2$  with domain  $[0, 1]$ . We vary  $\lambda$  from zero to one below.

```
library(magrittr)
library(ggplot2)
tibble::tibble(x = 1:100 / 100, xsq = x ^ 2) %>%
  ggplot() +
  geom_line(aes(y = x, x = x)) +
  geom_line(aes(y = xsq, x = x))
```



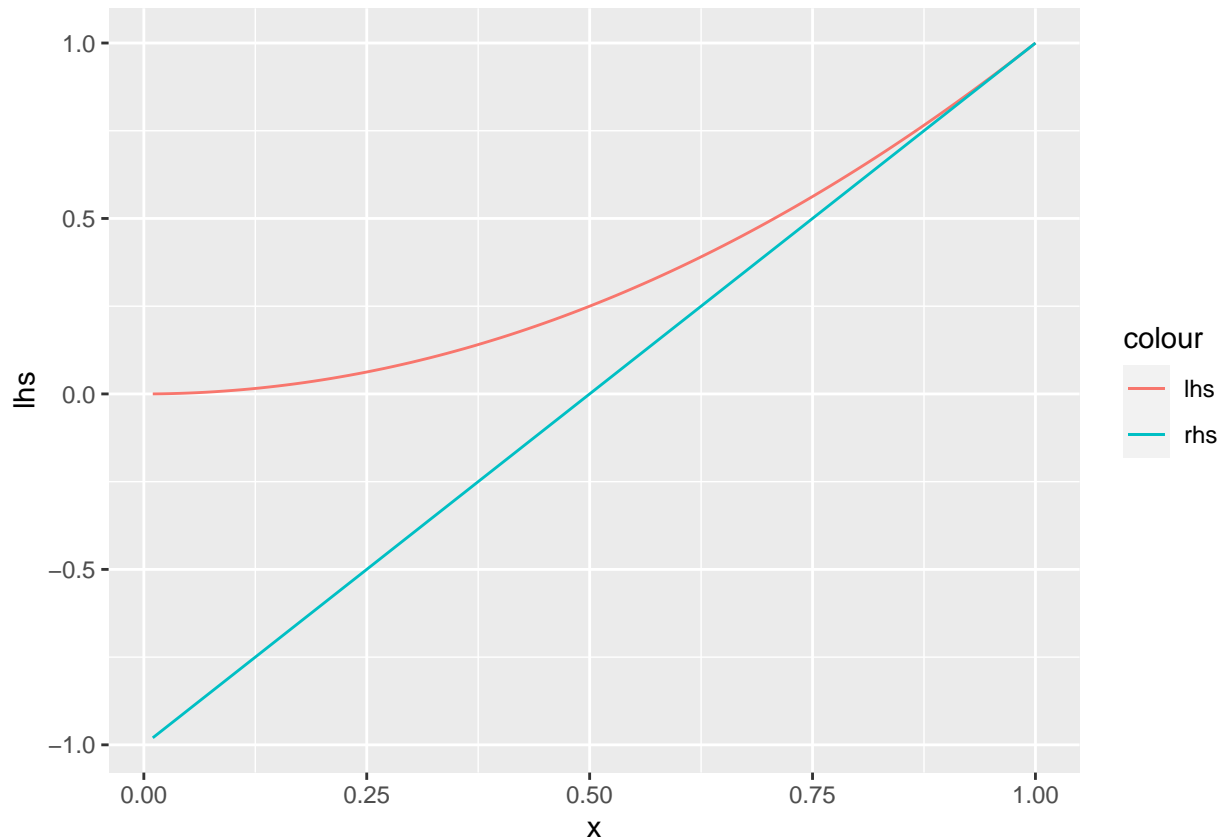
For  $x = 0, y = 1$ , we see that

$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$$

### Plot for iii

we set  $x_0 = 1$  and vary  $x$ .

```
tibble::tibble(x = 1:100 / 100, lhs = x ^ 2, rhs = 1 ^ 2 + 2 * (x - 1) ) %>%
  ggplot() + geom_line(aes(y = lhs, x = x, colour = "lhs")) + geom_line(aes(y = rhs, x = x, colour = "rhs"))
```



We plot along the horizontal axis the values of  $x$  and along the vertical axis the values of  $f(x_0) + f'(x_0)(x - x_0)$  and those of  $f(x)$ . “lhs” denotes  $f(x)$ , while “rhs” denotes  $f(x_0) + f'(x_0)(x - x_0)$

## Exercise 2

### Discrete random variables

A discrete random variable  $X$  taking countably many values  $x_i$  has expectation  $\mathbb{E}X = \sum_{i=1}^{\infty} p_i x_i$ .

$$f(\mathbb{E}X) = f\left(\sum_i p_i x_i\right) \leq \sum_i p_i f(x_i) = \mathbb{E}(f(X))$$

where we use induction on the natural numbers to argue that the inequality ii from question 1 extends to countably many summands.

### General random variables

We follow the hint.

$$\begin{aligned} f(X) &\geq f(\mathbb{E}X) + f'(\mathbb{E}X)([X - \mathbb{E}X]) \\ \implies \mathbb{E}(f(X)) &\geq f(\mathbb{E}X) + f'(\mathbb{E}X)(\mathbb{E}[X - \mathbb{E}X]) \\ \implies \mathbb{E}(f(X)) &\geq f(\mathbb{E}X) \end{aligned}$$

Note that the last inequality follows from line 2 because  $\mathbb{E}[X - \mathbb{E}X] = \mathbb{E}X - \mathbb{E}X = 0$ .

## Exercise 3

### Q3, Part i

We recall the definition of  $f$  continuous:  $f$  continuous if for any  $\epsilon > 0$  there is a  $\delta > 0$  such that  $d_X(x, y) < \delta \implies d_Y(f(x), f(y)) < \epsilon$ . To see that  $f$  is continuous, let  $\epsilon > 0$ . Then, observe that  $d_X(x, y) < \frac{\epsilon}{L} \implies d_Y(f(x), f(y)) < \epsilon$ , so  $\frac{\epsilon}{L}$  is the desired  $\delta$ .

### Q3, Part ii

I had to depart from the hint here. I started with the statement of the mean value theorem.

We note that  $f$  is continuous, since it is differentiable, so the mean value theorem applies. For some  $c \in (x, y)$  we have

$$\begin{aligned} f'(c) = \frac{f(x) - f(y)}{\|x - y\|} &\implies f(x) - f(y) = f'(c)[\|x - y\|] \\ &\leq L\|x - y\| \end{aligned}$$

### Q3, Part iii

$f$  is not differentiable because  $\lim_{x \downarrow 0} f'(x) = 1$  while  $\lim_{x \uparrow 0} f'(x) = -1$ .

To show that  $f$  is 1-lipchitz on  $\mathbb{R}$ , consider the arbitrary interval  $(a, b)$ . If  $a \geq 0$  then  $f(b) - f(a) = b - a$ . If  $b \leq 0$ , then  $f(b) - f(a) = -(b - a)$ . If  $a < 0$  and  $b > 0$  then  $f(b) - f(a) = |b - a| \leq |b - a|$ .

### Q3, Part iv

We note that the derivative is unbounded on the interval. This implies that there's a neighborhood near zero where  $\frac{f(x) - f(y)}{x - y} > L$  for any finite  $L$ .

### Q3, Part v

No. On  $\mathbb{R}$ , the slope,  $f'(x) = 2x$  goes to infinity as  $x \rightarrow \infty$ .

To be precise, suppose that there is a constant  $C$  such that  $f$  is  $C$ -lipschitz.  $f$   $C$ -lipschitz on  $\mathbb{R}$  means that for any two points  $x$  and  $y$ ,  $|f(x) - f(y)| \leq C|x - y|$ .

Choose  $x = C$  and  $y = C + 1$ . Then  $f(x) = C^2$  and  $f(y) = C^2 + 2C + 1$ , so  $|f(x) - f(y)| = 2C + 1$ , which is a contradiction. Thus, it must be that  $f$  is not  $L$ -lipschitz on  $\mathbb{R}$ .

## Exercise 4

### Q4, i

First, recall the definition of  $\epsilon$ -net. A set  $A_\epsilon \subset B$  is an  $\epsilon$ -net for  $B$  if, for any point  $x \in B$ ,  $d(x, A_\epsilon) \leq \epsilon$ .

We want to show that the algorithm in question creates an  $\epsilon$ -net.

Suppose that we enumerate the elements of  $T = \{x_1, \dots, x_n\}$ .

I claim that the set  $T$ , generated by the algorithm, is an  $\epsilon$ -net for  $B$ .

Suppose  $T$  is not an  $\epsilon$ -net for  $B$ . Then there is a point  $x_0 \in B$  such that  $d(x_0, B) \geq \epsilon$ . This implies that  $d(x_0, x_j) \geq \epsilon$ , but that, in turn, implies that  $x_0 \in B$ , a contradiction. Thus,  $T$  is an  $\epsilon$ -net for  $B$ .

#### Q4, ii

We want to show that

$$\cup_{x_i \in T} B_{\frac{\epsilon}{2}}(x_i) \subset B_{1+\frac{\epsilon}{2}}(0)$$

This is true because the elements of  $T$  are in  $B = B_1(0)$ , and thus, the  $\frac{\epsilon}{2}$  balls centered at the elements of  $T$  are contained in the  $1 + \frac{\epsilon}{2}$  ball centered at zero.

More precisely, choose a  $y \in \cup_{x_i \in T} B_{\frac{\epsilon}{2}}(x_i)$ . For some  $x_k$ ,  $y \in B_{\frac{\epsilon}{2}}(x_k)$ . Note that the distance from 0 to  $y$  can be no more than  $1 + \frac{\epsilon}{2}$ , so  $y \in B_{1+\frac{\epsilon}{2}}(0)$ .

Now, we want to clarify why this statement implies that  $N(B, \epsilon) \leq |T|$  and  $|T| \leq \left(\frac{3}{\epsilon}\right)^d$ .

We begin by verifying the first inequality. This follows directly from the definition of the algorithm.  $T$  is an  $\epsilon$ -net, so, by definition, the covering number is no bigger than the cardinality of  $T$ .

Consider the case  $d = 1$ . Here,  $B$  is the interval  $(-1, 1)$  on the real line.  $\frac{3}{\epsilon}$  is the number of  $\epsilon$ -balls needed when we permit overlap of balls, but prohibit any ball center from being in more than one ball.

#### Q4, iii

We first consider the case  $d = 1$ , where we see that we need at  $\frac{1}{\epsilon}$  balls to ensure that every point in the 1-ball is within  $\epsilon$  of an  $\epsilon$ -ball. This is because the diameter of the  $\epsilon$ -ball is  $2\epsilon$ , and the interval is length 2.

$$\frac{2}{2\epsilon} = \frac{1}{\epsilon}$$

We note that decreasing the number of  $\epsilon$ -balls would not result in a  $\epsilon$ -net.

Now, consider the case  $d > 1$ . We need to show that the  $N \geq \left(\frac{1}{\epsilon}\right)^d$

We follow the approach given in Bartlett's notes (<https://www.stat.berkeley.edu/~bartlett/courses/2013spring-stat210b/notes/12notes.pdf>).

Namely, we observe that, for an  $\epsilon$ -net of size  $N$ , we have

$$B \subset \cup_{i=1}^N (x_i + \epsilon B)$$

Consider, then, the volume of the ball.

$$\text{Volume}(B) \leq N \text{Volume}(\epsilon B) = N \epsilon^d \text{Volume}(B)$$

from which we see that

$$N \geq \frac{1}{\epsilon^d}$$

#### Q4, iv

To answer the question - I'm not sure because I can't recall if all norms on  $\mathbb{R}^d$  arise as  $\|x\|_p = (x_1^p + \dots + x_d^p)^{\frac{1}{p}}$ . It seems that for any natural number  $p \in \mathbb{N}$ ,  $\|x\|_p$  yields the results above. If it's true that all norms on  $\mathbb{R}^d$  look like a p-norm for some p, then the answer to the question is "no" because all norms give the same result.

### Exercise 5

#### Q5, i

We try to follow the hint. We have  $\frac{1}{\epsilon}$  "starting points". Each "starting point" has two choices for slope, either positive or negative at every increment of  $\epsilon$ , yielding  $2^{\frac{1}{\epsilon}}$  "paths" for each starting point.

Adding the number of “paths”, we get an upper bound of  $\frac{1}{\epsilon} 2^{\frac{1}{\epsilon}}$ .

**Q5, ii**

**Q5, iii**

I drew on Bartlett’s explanation (12notes.pdf, see above url) when attempting the solution below.

It’s slightly trickier, but we can approach this by again using a grid. Instead of a square grid with equal increment sizes, we now have a grid with  $\epsilon$  increments on the vertical axis and  $\frac{\epsilon}{L}$  increments on the horizontal axis. We again have  $\frac{1}{\epsilon}$  starting points, but, now, we have  $2^{\frac{L}{\epsilon}}$  paths.

## Exercise 6

**Q6, i**

**Q6, ii**

We first use Dudley’s result.

$$\mathbb{E} \left[ \sup_{\theta \in \Theta} |Y_n^{(f)}| \right] \leq 12K \int_0^K \sqrt{\frac{\log N}{n}} d\epsilon \leq 12K \int_0^K \sqrt{\frac{d \log \frac{3}{\epsilon}}{n}}$$

**Q6, iii**

$d$  and  $n$  balance each other. Linear increases in  $d$  can be offset by linear increases in  $n$ .