

Exercise 3

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Proof, without using the above theorem, that for any $a \in \mathbb{R}$, $f(x) = e^{ax}$ is a convex function.

We will use the definition of convexity that has been seen in class. A function $f : C \rightarrow \mathbb{R}$ defined on a convex set C is called convex if for all $x_1, x_2 \in C$ and $0 \leq \lambda \leq 1$, it satisfies that:

$$f(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda f(x_1) + (1 - \lambda)f(x_2) \quad (1)$$

The given function $f(x) = e^{ax}$ is defined in the convex set $C = \mathbb{R}$. Translated to our function, the expression (1) is equivalent to:

$$e^{a(\lambda x_1 + (1 - \lambda)x_2)} \leq \lambda e^{ax_1} + (1 - \lambda)e^{ax_2} \quad (2)$$

We have to show that expression (2) is true for all $a \in \mathbb{R}$ and all $x_1, x_2 \in \mathbb{R}$. We may assume that $x_1 > x_2$ without any loss of generalization. We will suppose that $a \geq 0$ from now on. The inequality in (2) may be transformed as it follows:

$$\begin{aligned} e^{a(\lambda x_1 + (1 - \lambda)x_2)} &\leq \lambda e^{ax_1} + (1 - \lambda)e^{ax_2} \\ e^{a\lambda x_1} e^{a\lambda x_2} e^{-a\lambda x_2} &\leq \lambda e^{ax_1} + (1 - \lambda)e^{ax_2} \end{aligned}$$

Now, we may divide both sides by e^{ax_2} , which always is a positive value:

$$e^{a\lambda x_1} e^{-a\lambda x_2} \leq \lambda e^{a(x_1 - x_2)} + (1 - \lambda)$$

At this point, it is convenient to define $t = (x_1 - x_2) > 0$.

$$e^{a\lambda t} \leq \lambda e^{at} + (1 - \lambda) \quad (3)$$

Suppose that $a = 0$. Then:

$$1 \leq \lambda + (1 - \lambda)$$

This last expression is trivially true. From now on, we will suppose that $a > 0$. We need to consider the Taylor Series expansions of both $e^{a\lambda x}$ and e^{ax} :

$$e^{a\lambda x} = 1 + \lambda ax + \frac{\lambda^2 a^2 x^2}{2} + \frac{\lambda^3 a^3 x^3}{6} + \dots \quad (4)$$

$$e^{ax} = 1 + ax + \frac{a^2 x^2}{2} + \frac{a^3 x^3}{6} + \dots \quad (5)$$

We will now express the right side of (3) using (5).

$$\begin{aligned} \lambda e^{ax} &= \lambda + \lambda ax + \lambda \frac{a^2 x^2}{2} + \lambda \frac{a^3 x^3}{6} + \dots \\ (1 - \lambda) + \lambda e^{ax} &= 1 + \lambda ax + \lambda \frac{a^2 x^2}{2} + \lambda \frac{a^3 x^3}{6} + \dots \end{aligned} \quad (6)$$

Evaluating $x = t$, we see that the left side of (3) is equivalent to (4) and the right side of (3) is equivalent to (6). Since $0 \leq \lambda \leq 1$, it is clear that (4) \leq (6) because:

$$\frac{\lambda^2 a^2 x^2}{2} + \frac{\lambda^3 a^3 x^3}{6} + \dots \leq \lambda \frac{a^2 x^2}{2} + \lambda \frac{a^3 x^3}{6} + \dots$$

The case when $a < 0$ is analogous, but instead of dividing both sides by e^{ax_2} , we divide them by $e^{-|a|x_1}$. The sign of t in this case is such that $t < 0$, and the sign of a and t compensate each other.

Therefore, (2) is true and in consequence, $f(x) = e^{ax}$ has been proven to be convex for any $a \in \mathbb{R}$.