

Some exercises

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1. Show that the set $A = \{(x, y) \in \mathbb{R}^2 : x > y\}$ is open in \mathbb{R}^2 .

Solution: Let $f(x, y) = x - y$. f is a difference of continuous functions, then, it is continuous. The codomain of f is open (given that $x > y \Rightarrow f(x, y) = x - y > 0$) Finally, continuous functions take open sets to open sets, and therefore the domain must be open. Then A must be open.

2. Show that $B \subset U \subset \mathbb{R}^2$ is closed in $U \Leftrightarrow \forall \{x_n\}_{n \in \mathbb{N}} \rightarrow \bar{x}, \{x_n\}_{n \in \mathbb{N}} \subset B, \bar{x} \in B$

Solution: (\Rightarrow)

Let $\{x_n\}_{n \in \mathbb{N}} \in B, \{x_n\}_{n \in \mathbb{N}} \rightarrow \bar{x}$.

Let $\bar{x} \in B^c$, given that B is closed, then by definition B^c is open. It follows that:

$$\exists \epsilon > 0 \text{ tal que } B_\epsilon(\bar{x}) \subset B^c$$

Using the definition of converging sequence, then $\exists n \in \mathbb{N}$ such that: $\|x_n - \bar{x}\| < \epsilon$

Then, if we count the $n > N$ we obtain a convergent sequence with part of it in B^c , but by hypothesis, the sequence is contained in B , contradiction.

(\Leftarrow)

The sequence $\{x_n\}_{n \in \mathbb{N}} \subset B$ converges to $\bar{x} \in B$. Assume that B is not closed, then B^c is not open, and therefore $\exists \bar{x} \in B^c$ such that $\forall \epsilon > 0, B_\epsilon(\bar{x})$ has some elements of B .

If you pick positive integers such that $\|x_n - \bar{x}\| < \frac{1}{n}$ with $x_n \in B$ then we have a convergent sequence with limit \bar{x} and $\{x_n\}_{n \in \mathbb{N}} \in B$ with $\bar{x} \in B^c$, contradiction.

3. Given $U \subset \mathbb{R}^n$, if the function $g : U \rightarrow \mathbb{R}$ is continuous in U , then $\{x \in U : g(x) \geq 0\}$ is closed in U . Show that you cannot change the *then* with an *if and only if* (i.e. \Leftrightarrow).

Solution: It is enough to show that $\{x \in U : g(x) \geq 0\}$ closed in U is not enough to show that $g : U \rightarrow \mathbb{R}$ is continuous in U . Even further, it is enough to show some g that satisfies $\{x \in U : g(x) \geq 0\}$ closed in U . Take $U = [-3, 3] \subset \mathbb{R}^n$ such that $g(x) = -1$ when $x \in [-3, 1)$ and $g(x) = 1$ when $x \in [1, 3]$. $\{x \in U : g(x) \geq 0\}$ is closed in U , $U \subset \mathbb{R}^n$ but g is not continuous.

4. Show: For $U \subset \mathbb{R}$ compact and f continuous, then $f : U \rightarrow \mathbb{R}$ is uniformly continuous.

*Any mistake in the solutions is of the exclusive my responsibility.

Solution: Let $B(x)$ an open ball in U , centered in x with radius $\delta(x)/2$. U is the union of a set of these open balls. For every point in the set, you can create one of these balls. However, as U is compact we can choose a finite set of points that serve as center for these balls, such that:

$$U = B(x_1) \cup B(x_2) \cup B(x_3) \dots \cup B(x_n)$$

If you choose δ as the lowest of the $\delta(x)/2$, we can be certain that we have satisfied the requirement (that the intersection is contained in the set). Now take p and $q \in U$, such that $d(p, q) < \delta$, for some x_i , you have that $p \in B(x_i)$. Then $d(p, x_i) < \delta(x_i)/2$. However,

$$d(p_i, q) \leq d(x_i, q) + d(p, q) < \delta(x_i)/2 + \delta \leq \delta(x_i)$$

Then we use $\delta = \min\{\delta(x_i)/2\}_{i \in \{1, \dots, n\}}$.

$$d(x_i, p) + d(x_i, q) < \delta(x_i)$$

Now using point continuity, given $\delta(x_i)$ we choose it, and we have $d(f(x_i), d(p)) < \frac{\epsilon}{2}$ y $d(f(x_i), f(q)) < \frac{\epsilon}{2}$.

$$d(f(p), f(q)) \leq d(f(x_i), f(p)) + d(f(x_i), f(q)) < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon$$

5. Show that \emptyset and \mathbb{R}^n are convex.

Solution: $\emptyset \rightarrow$ You cannot show that is not convex, as you cannot pick two elements in the set such that do not satisfy convexity. $\mathbb{R}^n \rightarrow$ Sea a y b $\in \mathbb{R}^n$, $t \in [0, 1]$ $ta + (1 - t)b$, because of the axioms of \mathbb{R}^n ta belongs to \mathbb{R}^n , and the summation as well, then it is convex (and a vector space!).

6. Given A and B convex, subsets of \mathbb{R}^n , show that:

$$A + tB := \{a + tb : a \in A, b \in B\}$$

is convex $\forall t \in \mathbb{R}$.

Solution: Let $a_i \in A$, $b_i \in B$ y $z_i = a_i + tb_i \in A + tB$ $i = 1, 2$. Let's show that $\{\lambda z_1 + (1 - \lambda)z_2\} \in A + tB$ $\lambda \in (0, 1)$ $\forall t \in \mathbb{R}$, $z_1, z_2 \in A + tB$

We know that

$$\lambda z_1 + (1 - \lambda)z_2 = \lambda a_1 + \lambda tb_1 + (1 - \lambda)a_2 + (1 - \lambda)tb_2 = \underbrace{\lambda a_1 + (1 - \lambda)a_2}_{\in A, \text{convex}} + t \underbrace{\lambda(b_1 + (1 - \lambda)b_2)}_{\in B, \text{convex}}$$

Then, it belongs to the set.

7. An arbitrary intersection of convex sets is convex.

Solution: Let a and b elements in the intersection of convex sets, then $(\lambda a + (1 - \lambda)b)$ belongs to each set, as these are convex. If it belongs to all, then it also belongs to the intersection. Then, the intersection is convex.

8. Show that a vector space is convex.

Solution: By definition, a vector space contains elements that can be added up and multiplied by a scalar. In particular, convex combinations of its elements belong to the same vector space.

Let a and b belong to a vector space, λa and $(1 - \lambda)b$ are well defined and belong to the vector space as $\lambda \in \mathbb{R}$, and particularly $\lambda \in (0, 1)$.

As the summation is closed in the vector space then the summation $\lambda a + (1 - \lambda)b$ belongs to the vector space, concluding the proof.

9. $C \subset \mathbb{R}^n$ convex \Rightarrow The closure is convex.

Solution: If C is closed, then it is trivial. Let C be not closed.

Let the closure to be not convex, while C is convex.

Let $a, b \in \bar{C} \setminus C$ $\lambda \in (0, 1)$, $\lambda a + (1 - \lambda)b \notin \bar{C}$ Then $\exists \{a_n\}_{n \in \mathbb{N}}, \{b_n\}_{n \in \mathbb{N}} \subset C$ such that $\{\lambda a_n + (1 - \lambda)b_n\} \in C \forall n$ A linear combination is continuous, so if I get close to a and b , the linear combination of these elements in the sequence must go through the closure, but C is convex, then contradiction.

10. There is a market with J assets traded by subjects that maximize their wealth. Each subject can create a portfolio $\theta = (\theta_1, \dots, \theta_J) \in \mathbb{R}^J$, where $\theta_j > 0$ means that the subject bought θ_j units of asset j , while $\theta_j < 0$ means that the subject promised in the future to pay the market value of θ_j units of asset J (or, *shorted* θ_j units of the asset j). The value of each unit of asset j is given by $q_j \geq 0$. Assume there is uncertainty regarding the future price of the assets. Consider that there are S different possible states of nature. On state $s \in \{1, \dots, S\}$ the asset $j \in \{1, \dots, J\}$ has a value of $R_{s,j}$ per unit. In sum, a portfolio $\theta \in \mathbb{R}^J$ will be worth, for a specific state of nature $s \in \{1, \dots, S\}$ $\sum_{j=1}^J R_{s,j} \theta_j$.

As said, the subjects try to maximize their wealth, it is expectable that there are no positions that could generate unlimited wealth without taking some risks. Formally, let's say that *there is no arbitrage* if there is no portfolio $\theta \in \mathbb{R}^J$ such that $\sum_{j=1}^J q_k \theta_j \leq 0$ and, for each state of nature $s \in \{1, \dots, S\}$, $\sum_{j=1}^J R_{s,j} \theta_j \geq 0$, with at least one of the inequalities being strict. Put in words, there is no way of (i) get more money today without sacrificing (expected) future wealth; or (ii) without paying nothing today increasing future wealth.

The following steps are suggested:

- (a) Let

$$A = \begin{pmatrix} -q_1 & \dots & -q_J \\ R_{1,1} & \dots & R_{1,J} \\ \vdots & & \vdots \\ R_{S,1} & \dots & R_{S,J} \end{pmatrix}$$

Show that, without arbitrage, the set $C := \{z \in \mathbb{R}^{S+1} : \exists \theta \in \mathbb{R}^J, z = A\theta\}$ is disjoint with $C_\epsilon := \{z \in \mathbb{R}_+^{S+1} : \|z\| \in [\epsilon, 2]\}$, $\epsilon > 0$.

Solution: $C :=$

$$z = \begin{pmatrix} \sum_{i=1}^J -q_i \theta_i \\ \sum_{i=1}^J R_{1,i} \theta_i \\ \vdots \\ \sum_{i=1}^J R_{S,i} \theta_i \end{pmatrix}$$

$$\begin{aligned}
\rightsquigarrow \quad z_1 &= -q_1\theta_1 - q_2\theta_2 - \dots - q_J\theta_J \\
&\vdots \\
z_i &= R_{i1}\theta_1 + R_{i2}\theta_2 - \dots + R_{iJ}\theta_J \\
&\vdots \\
z_s &= R_{s1}\theta_1 + R_{s2}\theta_2 - \dots + R_{sJ}\theta_J
\end{aligned}$$

If there is no arbitrage then:

$$\nexists \theta \in \mathbb{R}^J \text{ such that } \sum_{j=1}^J q_j \theta_j \leq 0 \text{ and } s \in \{1 \dots S\} \sum_{j=1}^J R_{sj} \theta_j \geq 0$$

If you could find some θ such that happens, then $z \in \mathbb{R}_+^{s+1}$. As it doesn't happen, it can be zero, or with some negative coordinate, then $z \in \mathbb{R}^{s+1} \setminus \mathbb{R}_+^{s+1}$. It follows that $C_\epsilon \cap C = \emptyset$ as $C_\epsilon \subset \mathbb{R}_+^{s+1}$ and therefore they are disjoint.

(b) Show that C and C_ϵ are convex, non empty and closed. Show that C_ϵ is compact.

Solution:

The empty set case is trivial because $0 \in C$ and given $z \in C_\epsilon$, in particular a vector such that $\{c \in \mathbb{R}_+^{s+1} : 0 \leq \|c\| \leq 2\} \Rightarrow \sqrt{\sum_{i=1}^{s+1} (c_i)^2} \leq 2$

$$\sum_{i=1}^{s+1} (c_i)^2 \leq 4$$

For example $c_j = 2$ and all the others 0 satisfies the requirement.

Convexity,

The set C_ϵ is a closed ball in \mathbb{R}_+^{s+1} , that is a convex set. Then C_ϵ is convex.

C : Let a and $b \in C$, $\lambda \in [0, 1]$

$$\begin{aligned}
\lambda a + (1 - \lambda)b &= \lambda A\theta_c + (1 - \lambda)A\theta_b \\
&= A\theta_b + \lambda(A\theta_c - A\theta_b) = A[\theta_b + \lambda(\theta_c - \theta_b)] \\
&= A \underbrace{(\theta_b(1 - \lambda) + \lambda\theta_c)}_{\in \mathbb{R}^J} \Rightarrow \exists \theta \in \mathbb{R}^J \text{ such that the set is convex}
\end{aligned}$$

Closedness,

C_ϵ is closed in \mathbb{R}_+^{s+1} . It is enough to show that \mathbb{R}_+^{s+1} is open if we exclude C_ϵ , that is, to show that $\{z \in \mathbb{R}_+^{s+1} : \|z\| < 2\}$, but the function $\|\cdot\|$ is continuous, and its codomain in the set is $(2, +\infty)$, then the set is open. By definition of a closed set, C_ϵ is closed.

C is closed, because z is a linear transformation of a vector space. The linear transformation of a vector space is itself a vector space, and a vector space is always closed.

That C_ϵ is compact is trivial, as the norm of all of its element is bounded by 2. Then, as we have shown, C_ϵ is closed and bounded, then it is compact.

- (c) Show that there is $p \gg 0$ such that $pz \leq 0$, for each $z \in C$. [Hint: Lookout the **hyperplane separation theorem**]

Solution:

From the previous result, we can apply the hyperplane separation theorem. In particular we have

$$pa < c < pb \quad \forall (a, b) \in C \times C_\epsilon, \text{ for some } c \in \mathbb{R}$$

Where we can chose p .

Of course we cannot set $p = 0$ as $0 < 0$ is false. The elements in C_ϵ have at least one positive coordinate. Then, we can bound it from below with 0. However 0 is feasible in C , and we know that $pz \leq 0$. As z has some negative coordinate, the it could be that if p would have negative coordinates, we could obtain something “greater or equal” than 0. Finally, as p cannot be 0, then $p \gg 0$.

- (d) Show that $pz = 0$ for each $z \in C$ (remember that C is a vector subspace).

Solution: Let's assume that $pz < 0$. If that is true, then $p(-z) > 0$, but as C is a vector subspace, $-z$ must belong to the vector subspace, but $pz < 0 \quad \forall z \in C$. Contradiction. The only possibility is that $pz = 0$.

- (e) Conclude the proof.

Solution: Given that $pz = 0$ and that $p \gg 0$ then:

$$\begin{aligned} pz &= 0 \\ -p_1 \sum_{j=1}^J q_j \theta_j + p_2 \sum_{j=1}^J R_{1,j} \theta_j + \dots + p_{s+1} \sum_{j=1}^J R_{S,j} \theta_j &= 0 \\ p_2 \sum_{j=1}^J R_{1,j} \theta_j + \dots + p_{s+1} \sum_{j=1}^J R_{S,j} \theta_j &= p_1 \sum_{j=1}^J q_j \theta_j \\ \frac{p_2}{p_1} \sum_{j=1}^J R_{1,j} \theta_j + \dots + \frac{p_{s+1}}{p_1} \sum_{j=1}^J R_{S,j} \theta_j &= \sum_{j=1}^J q_j \theta_j \end{aligned}$$

If we redefine, $\frac{p_i}{p_1} = \gamma_{i-1}$ then we have:

$$\gamma_1 \sum_{j=1}^J R_{1,j} \theta_j + \dots + \gamma_S \sum_{j=1}^J R_{S,j} \theta_j = \sum_{j=1}^J q_j \theta_j$$

The this is equivalent to:

$$\begin{aligned} \gamma^T R \theta &= q^T \theta \\ \gamma^T R &= q^T \end{aligned}$$

The for each element i we have that:

$$\sum_{j=1}^J \gamma_i R_{i,j} = q_i$$

For the next questions, consider that a function is said to be **quasiconcave** if $f(\lambda x_1 + (1 - \lambda)x_2) \geq \lambda f(x_1) + (1 - \lambda)f(x_2)$

11. Show that any function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ as $f(x) = ax + b$, where $a \in \mathbb{R}^n$ and $b \in \mathbb{R}$, is quasiconcave.

Solution:

$$\begin{aligned}
 f(\lambda x_1 + (1 - \lambda)x_2) &= a\lambda(x_1 - x_2) + f(x_2) + \lambda b - \lambda b \\
 &= \lambda f(x_1) - \lambda f(x_2) + f(x_2) \\
 &= \lambda f(x_1) + (1 - \lambda)f(x_2) \\
 \Rightarrow f(\lambda x_1 + (1 - \lambda)x_2) &= \lambda f(x_1) + (1 - \lambda)f(x_2) \\
 \text{si, } f(x_1) &> f(x_2) \\
 &\geq f(x_2) = \min f(x_1), f(x_2) \\
 \text{si, } f(x_1) &< f(x_2) \\
 &\geq f(x_1) = \min f(x_1), f(x_2) \\
 \Rightarrow f(\lambda x_1 + (1 - \lambda)x_2) &\geq \min f(x_1), f(x_2)
 \end{aligned}$$

12. A monotone function (increasing or decreasing) is always quasiconcave.

Solution: Having $x_1 \geq x_2$ is clear that $\lambda x_1 + (1 - \lambda)x_2 \geq x_2$, Then:

$$\begin{aligned}
 f(\lambda x_1 + (1 - \lambda)x_2) &\geq f(x_2), \text{ if } f \text{ is increasing.} \\
 \Rightarrow f(x_2) &= \min\{f(x_1), f(x_2)\} \\
 f(\lambda x_1 + (1 - \lambda)x_2) &\geq f(x_1), \text{ if } f \text{ is decreasing.} \\
 \Rightarrow f(x_1) &= \min\{f(x_1), f(x_2)\}
 \end{aligned}$$

In both cases $f(\lambda x_1 + (1 - \lambda)x_2) \geq \min\{f(x_1), f(x_2)\}$, then f is quasiconcave.

13. Any concave function is quasiconcave.

Solution: If:

$$f(x_1) < f(x_2) \Rightarrow \lambda f(x_1) + (1 - \lambda)f(x_2) \geq \lambda f(x_1) + (1 - \lambda)f(x_1) = f(x_1) = \min\{f(x_1), f(x_2)\}$$

The proof in the opposite case follows in the same way.

14. Given function $f : U \subset \mathbb{R}^n \rightarrow \mathbb{R}$, where U is convex, f is quasiconcave in U if and only if, for each a in \mathbb{R} the set $U_a = \{x \in U : f(x) > a\}$ is convex.

Solution:

(\Rightarrow) hypothesis:

f is quasiconcave in $U, a \in \mathbb{R}$, for when $U = \emptyset$, is trivial. Let $U \neq \emptyset$.

Let $x_1, x_2 \in U$ $a \in \mathbb{R}$, $\lambda \in (0, 1)$.

$$z_\lambda = \lambda x_1 + (1 - \lambda)x_2 \in U, \text{ because } U \text{ is convex}$$

using f 's quasiconcavity

$$f(\lambda x_1 + (1 - \lambda)x_2) \geq \min\{f(x_1), f(x_2)\} \rightarrow \text{(easy to see by the def. of the set)}$$

$$\min\{f(x_1), f(x_2)\} > a \text{ then}$$

$$f(\lambda x_1 + (1 - \lambda)x_2) > a \Rightarrow f(\lambda x_1 + (1 - \lambda)x_2) \in U_a$$

$$\Rightarrow U_a \text{ is convex}$$

(\Leftarrow) Now the hypothesis is: U_a is convex for each $a \in \mathbb{R}$. Let $x_1, x_2 \in U$, $\{x_1, x_2\} \subset U_{\min\{f(x_1), f(x_2)\}}$ then for each $\lambda \in (0, 1)$

$$\lambda x_1 + (1 - \lambda)x_2 \in \min\{f(x_1), f(x_2)\}$$

$$\Rightarrow f(\lambda x_1 + (1 - \lambda)x_2) \geq \min\{f(x_1), f(x_2)\} \quad \forall \lambda \in (0, 1)$$

Concluding the proof.

15. Given $(\alpha, \beta) >> 0$, $(x, y) \in \mathbb{R}_+$, the function $f(x, y) = x^\alpha y^\beta$ is strictly quasiconcave.

Solution: Using the result in exercise (14), if $a = 0$, then it is easy to see that f is quasiconcave. (\mathbb{R}_+^2 is convex).

16. Given $a \in \mathbb{R}^n$, $f(x) = -\|x - a\|$ is strictly quasiconcave.

Solution: Let $x, y \in \mathbb{R}^n$

$$z = \lambda x + (1 - \lambda)y$$

$$g(x) = -f(x)$$

$$g(z) = \|\lambda x + (1 - \lambda)y - a\|$$

$$= \|\lambda(x - a) + (1 - \lambda)(y - a)\|$$

$$\leq \lambda \|x - a\| + (1 - \lambda) \|y - a\|$$

Assume an x further from a than from y .

$$g(z) < g(x)$$

$$g(\lambda x + (1 - \lambda)y) < g(x)$$

$$\|\lambda x + (1 - \lambda)y - a\| < \|x - a\| \quad / - 1$$

$$- \|\lambda x + (1 - \lambda)y - a\| > - \|x - a\|$$

$$f(z) > f(x)$$

Because we assumed x further from a than from y ,

$$g(x) = \max\{g(x), g(y)\}$$

$$\Rightarrow f(x) = \min\{f(x), f(y)\}$$

Then $f(z) > \min\{f(x), f(y)\}$, concluding the proof.

1 Bonus Track, Proof fo the Caratheodory's Theorem

Definition 1.1. The covenxhull of a set S is the set that contains all the convex combinations of a finite number of elements of S .

If some $x \in \mathbb{R}^d$ is in the convexhull of P ($co(P)$), then, there is a subset $P' \subset P$ that has $d + 1$ or less elements such that x is in the $co(P)$.

Solution:

Let $x \in co(P)$. Then x is a convex combination of a finite number of elements of P :

$$x = \sum_{j=1}^k \lambda_j x_j$$

where each $x_j \in P$, each $\lambda_j \geq 0$ and it holds that

$$\sum_{j=1}^k \lambda_j = 1$$

Assume that $k > d + 1$ (if not, there is nothing to prove). Then, the elements $(x_2 - x_1), \dots, (x_k - x_1)$ are linearly dependent, that is there are scalars μ_2, \dots, μ_k such that:

$$\begin{aligned} \sum_{j=2}^k \mu_j (x_j - x_1) &= 0 \\ \sum_{j=2}^k \mu_j x_j - \sum_{j=2}^k \mu_j x_1 &= 0 \\ \sum_{j=2}^k \mu_j x_j &= \sum_{j=2}^k \mu_j x_1 \end{aligned}$$

If we define

$$\mu_1 := - \sum_{j=2}^k \mu_j$$

$$\begin{aligned} \sum_{j=2}^k \mu_j x_j &= \sum_{j=2}^k \mu_j x_1 \quad / + \mu_1 x_1 \\ \sum_{j=1}^k \mu_j x_j &= x_1 \sum_{j=1}^k \mu_j \end{aligned}$$

And noting that:

$$x_1 \sum_{j=1}^k \mu_j = \mu_1 x_1 + \sum_{j=2}^k \mu_j x_1 = -x_1 \sum_{j=2}^k \mu_j + x_1 \sum_{j=2}^k \mu_j = x_1 \left[-\sum_{j=2}^k \mu_j + \sum_{j=2}^k \mu_j \right] = 0$$

Then

$$\sum_{j=1}^k \mu_j x_j = x_1 \sum_{j=1}^k \mu_j = 0$$

$$\sum_{j=1}^k \mu_j x_j = 0$$

$$\sum_{j=1}^k \mu_j = 0$$

And not all the μ_j are zero. Then, at least some $\mu_j > 0$. It follows that,

$$x = \sum_{j=1}^k \lambda_j x_j - \alpha \sum_{j=1}^k \mu_j x_j = \sum_{j=1}^k (\lambda_j - \alpha \mu_j) x_j$$

for some $\alpha \in \mathbb{R}$. In particular, the inequality will hold if α is defined as,

$$\alpha := \min_{1 \leq j \leq k} \left\{ \frac{\lambda_j}{\mu_j} : \mu_j > 0 \right\} = \frac{\lambda_i}{\mu_i}$$

Note that $\alpha > 0$ and for j between 1 and k ,

$$\lambda_j - \alpha \mu_j \geq 0$$

in particular $\lambda_j - \alpha \mu_j = 0$, by definition of α . Then

$$x = \sum_{j=1}^k (\lambda_j - \alpha \mu_j) x_j$$

where each $(\lambda_j - \alpha \mu_j)$ is non negative, adds up to 1 and even further, $\lambda_i - \alpha \mu_i = 0$. In other words x is a convex combination of at the most $k - 1$ elements of P .

This process can be repeated until x is represented as a convex combination of at the most $d + 1$ elements of P .