

# Econ 508-A

## Convex Sets and Concave Functions

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# CONVEX SETS IN $\mathbb{R}^N$

DEFINITION: **A set**  $X \subseteq \mathbb{R}^N$  **is convex** iff given any two points  $x', x'' \in X$ , the point

$$x_\alpha = \alpha x' + (1 - \alpha)x'' \in X \text{ for every } \alpha \in [0, 1].$$

$x_\alpha$  is called a **convex combination** of  $x'$  and  $x''$ .

## CONVEX SETS IN $\mathbb{R}^N$ (2)

Another ways of writing down the convex combination of two points is:

$$x_\alpha = x'' + \alpha(x' - x'')$$

The set of all convex combinations between any two points is the line segment that joins them.

Graphically, a set is convex if given any two points, the line that joins them is in the set.

[picture]

## CONVEX SETS IN $\mathbb{R}^N$ (3)

A set is **strictly convex** if the line joining any two points lies in the interior of the set:

For any  $x', x'' \in X$ , and  $\alpha \in [0, 1]$ ,

$$x_\alpha = \alpha x' + (1 - \alpha)x'' \in \text{int}(X)$$

[picture]

# PROPERTIES OF CONVEX SETS

**THEOREM:** Let  $X, Y \subseteq \mathbb{R}^N$  be convex sets and  $r \in \mathbb{R}$ . Then the sets:

1.  $rX = \{z \in \mathbb{R}^N : \exists x \in X \text{ s.t. } z = rx\}$
2.  $X + Y = \{z \in \mathbb{R}^N : \exists x \in X \text{ and } y \in Y \text{ s.t. } z = x + y\}$
3.  $X \cap Y$

Are convex sets.

**PROOF:** HW

# CONVEX COMBINATION

DEFINITION:  $\mathbf{x} \in \mathbb{R}^N$  is a **convex combination** of the vectors  $\mathbf{x}^1, \dots, \mathbf{x}^m \in \mathbb{R}^N$  if it can be written as

$$\mathbf{x} = \sum_{i=1}^m \alpha^i \mathbf{x}^i = \begin{bmatrix} \sum_{i=1}^m \alpha^i x_1^i \\ \vdots \\ \sum_{i=1}^m \alpha^i x_N^i \end{bmatrix}$$

with  $\alpha^i \in [0, 1]$  for all  $i$  and  $\sum_{i=1}^m \alpha^i = 1$

# CONVEX COMBINATION (2)

THEOREM: A set  $X$  is convex iff every convex combination of points in  $X$  lies in  $X$ .

PROOF: Rockafellar.

# CONVEX HULL

DEFINITION: Given a set  $X \subseteq \mathbb{R}^N$ , the **convex hull** of  $X$ , denoted  $\text{Conv}X$  is the smallest convex set that contains  $X$ .

[picture]

Alternative characterizations of  $\text{Conv}X$

1. The set containing all possible convex combinations of points in  $X$ :

$$\text{Conv}X = \left\{ \mathbf{y} : \mathbf{y} = \sum_{i=1}^m \alpha^i \mathbf{x}^i \text{ for some } m, \text{ with } \mathbf{x}^i \in X, \right. \\ \left. \alpha^i \in [0, 1] \text{ and } \sum_{i=1}^m \alpha^i = 1 \right\}.$$

2. The intersection of all the convex sets that contain  $X$ .



# CONVEX CONE

DEFINITION:  $K \subseteq \mathbb{R}^N$  is a **convex cone** if for any  $\mathbf{x}, \mathbf{x}' \in K$  and  $\alpha, \beta > 0$ ,  $\alpha\mathbf{x} + \beta\mathbf{x}' \in K$ .

Example: the nonnegative orthant of  $\mathbb{R}^N$ :

$$\mathbb{R}_+^N = \{\mathbf{x} = (x_1, x_2, \dots, x_N) \mid x_i \geq 0 \text{ for } i = 1, \dots, N\}$$

[picture]

# SEPARATION THEOREMS

DEFINITION: Given  $\mathbf{p} \in \mathbb{R}^N$ ,  $\mathbf{p} \neq \mathbf{0}$ ,  $c \in \mathbb{R}$ , the **hyperplane** generated by  $\mathbf{p}$  and  $c$  is the  $(N - 1)$  dimensional plane

$$H_{\mathbf{p},c} = \{\mathbf{x} \in \mathbb{R}^N : \mathbf{p} \cdot \mathbf{x} = c\}$$

The sets  $\{\mathbf{x} \in \mathbb{R}^N : \mathbf{p} \cdot \mathbf{x} \geq c\}$  and  $\{\mathbf{x} \in \mathbb{R}^N : \mathbf{p} \cdot \mathbf{x} \leq c\}$  are called the **half-space above** and the **half-space below** the hyperplane  $H_{\mathbf{p},c}$ .

[picture]

Hyperplanes and half spaces are convex sets.

## SEPARATION THEOREMS (2)

DEFINITION: Consider two nonempty sets  $X, Y \subseteq \mathbb{R}^N$ .

- ▶  $X$  and  $Y$  are **separated by a hyperplane**  $H_{p,c}$  iff  $\exists p \neq \mathbf{0} \in \mathbb{R}^N$  and  $c \in \mathbb{R}$  such that

$$p \cdot x \geq c \geq p \cdot y \quad \forall x \in X, y \in Y.$$

- ▶  $X$  and  $Y$  are **strictly separated by a hyperplane**  $H_{p,c}$  iff there exists  $p \neq \mathbf{0} \in \mathbb{R}^N$  and  $c \in \mathbb{R}$  such that

$$p \cdot x > c > p \cdot y \quad \forall x \in X, y \in Y.$$

[picture]

## SEPARATION THEOREMS (3)

**THEOREM: separating Hyperplane Theorem:** Let  $X, Y \subseteq \mathbb{R}^N$  be nonempty.

1. If  $X$  and  $Y$  are convex, and  $X \cap Y = \emptyset$ , then  $X$  and  $Y$  are separated by a hyperplane.
2. If  $X$  and  $Y$  are convex and closed, if  $X \cap Y = \emptyset$ , and if at least one of them is compact, then  $X$  and  $Y$  are strictly separated by a hyperplane.

PROOF: omitted.

# SEPARATION THEOREMS (4)

DEFINITION: Let  $X \subseteq \mathbb{R}^N$  be nonempty.

- ▶  $X$  is supported at  $x^*$  iff  $\exists p \neq 0 \in \mathbb{R}^N$  such that  $p \cdot x \geq p \cdot x^*$  for all  $x \in X$ .
- ▶  $X$  is strictly supported at  $x^*$  iff there exists  $p \neq 0 \in \mathbb{R}^N$  such that  $p \cdot x > p \cdot x^*$  for all  $x \in X, x \neq x^*$ .

[picture]

# SEPARATION THEOREMS (5)

**THEOREM: Supporting Hyperplane Theorem.** Suppose that  $X \subseteq \mathbb{R}^N$  is nonempty, closed and convex and that  $x^* \notin \text{Int}(X)$ . Then  $X$  is supported by a hyperplane at  $x^*$ .

[picture]

PROOF: omitted.

# DEFINITIONS

From now on, let  $C \subseteq \mathbb{R}^N$  be a convex set, and  $f : C \rightarrow \mathbb{R}$

The **hypograph** of  $f$  ( $\text{hyp } f$ ). is the set of points lying on and below the graph of  $f$ :

$$\text{hyp } f = \{(x, y) \in \mathbb{R}^{N+1} : x \in \mathbb{R}^N \text{ and } y \leq f(x)\}$$

Similarly, the **epigraph** of  $f$  ( $\text{epi } f$ ). is the set of points lying on and above the graph of  $f$ :

$$\text{epi } f = \{(x, y) \in \mathbb{R}^{N+1} : x \in \mathbb{R}^N \text{ and } y \geq f(x)\}$$

[pictures]

# CONCAVE FUNCTIONS

DEFINITION: Let  $x', x'' \in C$ , and define  $x^\alpha = \alpha x' + (1 - \alpha)x''$ , where  $\alpha \in [0, 1]$ .  $f$  is concave iff

$$f(x^\alpha) \geq \alpha f(x') + (1 - \alpha)f(x'')$$

[picture]

THEOREM:  $f$  is concave iff the set  $\text{hyp } f$  is convex.



# PROOF

(i)  $f$  concave  $\Rightarrow$  hyp  $f$  convex

Pick two arbitrary points in hyp  $f : (x', y'), (x'', y'')$ , and let  $(x^\alpha, y^\alpha) = \alpha(x', y') + (1 - \alpha)(x'', y'')$ .

Want to show that  $(x^\alpha, y^\alpha) \in \text{hyp } f$ .

First, we know that  $y' \leq f(x')$  and  $y'' \leq f(x'')$ .

So  $\alpha y' \leq \alpha f(x')$  and  $(1 - \alpha)y'' \leq (1 - \alpha)f(x'')$ .

Therefore,  $y^\alpha = \alpha y' + (1 - \alpha)y'' \leq \alpha f(x') + (1 - \alpha)f(x'')$

By concavity of  $f$ ,  $\alpha f(x') + (1 - \alpha)f(x'') \leq f(x^\alpha)$

Then  $y^\alpha \leq f(x^\alpha)$  and, hence,  $(x^\alpha, y^\alpha) \in \text{hyp } f$

## PROOF (2)

(ii)  $f$  concave  $\Leftrightarrow$  hyp  $f$  convex

Given any two points  $\mathbf{x}', \mathbf{x}''$  in the domain of  $f$ ,  $(\mathbf{x}', f(\mathbf{x}'))$  and  $(\mathbf{x}'', f(\mathbf{x}''))$  are in hyp  $f$ .

Let  $\mathbf{x}^\alpha = \alpha\mathbf{x}' + (1 - \alpha)\mathbf{x}''$  and  $f^\alpha = \alpha f(\mathbf{x}') + (1 - \alpha)f(\mathbf{x}'')$

Want to show that  $f(\mathbf{x}^\alpha) \geq f^\alpha$

By convexity of hyp  $f$ , the point  $(\mathbf{x}^\alpha, f^\alpha)$  is also in hyp  $f$

Which means that  $f^\alpha \leq f(\mathbf{x}^\alpha)$

So  $\alpha f(\mathbf{x}') + (1 - \alpha)f(\mathbf{x}'') \leq f(\mathbf{x}^\alpha)$ , and, hence,  $f$  is concave. ■

# CONVEX FUNCTIONS

DEFINITION:  $f$  is **convex** iff  $-f$  is concave.

THEOREM: the following are equivalent:

1.  $f$  is convex.
2.  $\text{epi } f$  is a convex set.
3. For any  $x', x'' \in C$ ,  $x^\alpha = \alpha x' + (1 - \alpha)x''$ ,

$$f(x^\alpha) \leq \alpha f(x') + (1 - \alpha)f(x'').$$

PROOF: HW.

# STRICT CONCAVITY AND CONVEXITY

DEFINITION:

1.  $f$  is **strictly concave** iff for any  $\mathbf{x}', \mathbf{x}'' \in C$ ,  $\mathbf{x}' \neq \mathbf{x}''$ , and any  $\alpha \in (0, 1)$

$$f(\mathbf{x}^\alpha) > \alpha f(\mathbf{x}') + (1 - \alpha)f(\mathbf{x}'')$$

2.  $f$  is **strictly convex** iff  $-f$  is strictly concave.

## SOME USEFUL THEOREMS

**THEOREM:** Let  $f_1, \dots, f_k$  be concave functions, each defined on  $C$ . Let  $a_1, \dots, a_k \in \mathbb{R}_+$ . Then  $a_1 f_1 + \dots + a_k f_k$  is a concave function on  $C$ .

**PROOF:** HW.

**THEOREM:** Let  $f : C \rightarrow \mathbb{R}$  where  $C$  is open and convex. If  $f$  is concave on  $C$ , it is continuous on  $C$ . More generally,  $f$  is continuous on the interior of  $C$ .

## OTHER USEFUL PROPERTIES

- a. If  $f$  is concave and  $F$  is concave and increasing, then  $U(\mathbf{x}) = F(f(\mathbf{x}))$  is concave.
- b. If  $f$  is convex and  $F$  is convex and increasing, then  $U(\mathbf{x}) = F(f(\mathbf{x}))$  is convex.
- c. Let  $f$  and  $g$  be concave functions. Then  $h(\mathbf{x}) = \min\{f(\mathbf{x}), g(\mathbf{x})\}$  is concave.
- d. Let  $f$  and  $g$  be convex functions. Then  $h(\mathbf{x}) = \max\{f(\mathbf{x}), g(\mathbf{x})\}$  is convex.

# QUASICONCAVE FUNCTIONS

DEFINITION:

1.  $f$  is **quasiconcave** iff

$$\forall x', x'' \in C, x^\alpha = \alpha x' + (1 - \alpha)x'', \alpha \in [0, 1].$$

$$f(x^\alpha) \geq \min\{f(x'), f(x'')\}.$$

2.  $f$  is **strictly quasiconcave** iff

$$\forall x', x'' \in C, x^\alpha = \alpha x' + (1 - \alpha)x'', \alpha \in [0, 1]$$

$$f(x^\alpha) > \min\{f(x'), f(x'')\}$$

The function  $f$  is quasiconvex iff  $-f$  is quasiconcave  
 strictly quasiconvex iff  $-f$  is strictly quasiconcave.

# QUASICONCAVITY AND UPPER CONTOUR SETS

Given a point  $a \in \mathbb{R}$ , the upper contour set of  $f$  is the set

$$U_a = \{x \in C : f(x) \geq a\}$$

**THEOREM:**  $f$  is quasiconcave iff the upper contour sets are convex.



# PROOF

(i)  $f$  is quasiconcave  $\Rightarrow U_y$  convex.

If  $U_a = \emptyset$ , it is convex trivially.

If  $U_a \neq \emptyset$ , take any two points  $x', x'' \in U_a$ .

By definition of  $U_a$ , both  $f(x') \geq a$  and  $f(x'') \geq a$ .

By quasiconcavity of  $f$ ,  $f(x^\alpha) \geq \min \{f(x'), f(x'')\} \geq a$

So  $x^\alpha \in U_a$ , and, hence,  $U_a$  is convex.

(ii)  $f$  is quasiconcave  $\Leftarrow U_a$  convex: HW.

[picture: examples  $\mathbb{R}, \mathbb{R}^2$ ]

Similarly,  $f$  is quasiconvex iff the lower contour sets,  
 $L_a = \{x \in C : f(x) \leq a\}$ , are convex.

# USEFUL THEOREMS

THEOREM: let  $f$  be quasiconcave, and  $g : \mathbb{R} \rightarrow \mathbb{R}$  be a weakly increasing function defined on an interval  $I$  that contains  $f(C)$ . Then the composite function  $g \circ f(x)$  is quasiconcave in  $C$ .

PROOF:

Pick any two points  $x', x'' \in C$ . WLOG, assume that  $f(x'') \geq f(x')$

By quasiconcavity of  $f$ ,  $f(x^\alpha) \geq f(x')$ .

Since  $g$  is nondecreasing,  $g(f(x'')) \geq g(f(x'))$  and  $g(f(x^\alpha)) \geq g(f(x'))$

Hence,  $g \circ f(x^\alpha) \geq \min\{g(f(x'')), g(f(x'))\}$ . ■

REMARK: nondecreasing transformations preserve quasiconcavity, NOT concavity.

## USEFUL THEOREMS (2)

THEOREM: if  $f$  is concave, then  $f$  is quasiconcave.

PROOF:

Let  $f$  be concave, with  $f(\mathbf{x}'') \geq f(\mathbf{x}')$ .

By definition of concavity,  $f(\mathbf{x}^\alpha) \geq \alpha f(\mathbf{x}'') + (1 - \alpha)f(\mathbf{x}')$

Since  $f(\mathbf{x}'') \geq f(\mathbf{x}')$ ,

$$\alpha f(\mathbf{x}'') + (1 - \alpha)f(\mathbf{x}') \geq \alpha f(\mathbf{x}') + (1 - \alpha)f(\mathbf{x}') = f(\mathbf{x}').$$

Then  $f(\mathbf{x}^\alpha) \geq f(\mathbf{x}') = \min\{f(\mathbf{x}''), f(\mathbf{x}')\}$ .

So  $f$  is quasiconcave. ■

The converse is not true.

# EXAMPLE 1

1.  $f(x) = e^x$  :

Quasiconcave?

For any  $y > 0$  :  $U_y = [\ln y, \infty)$ , which is a convex set.

For  $y \leq 0$  :  $U_y = \mathbb{R}$ , which is also convex.

So  $f(x) = e^x$  is quasiconcave

Concave?

No, convex.      [picture]

## EXAMPLE 2

2.  $f(x) = e^{-x^2}$  (similar to the standard normal density function)

Quasiconcave?

For any  $y \in (0, 1]$ ,  $U_y = \left[-\sqrt{\ln(1/y)}, \sqrt{\ln(1/y)}\right]$ , which is a convex set.

For any  $y > 1$ ,  $U_y = \emptyset$ , which is also convex

For any  $y \leq 0$ ,  $U_y = \mathbb{R}$ , which is also convex

Concave?

No. [picture]

## EXAMPLE 3

3. Consider the following Cobb Douglas utility function:

$$u : \mathbb{R}_{++} \rightarrow \mathbb{R}, \quad u(x_1, x_2) = x_1^{\frac{1}{3}} x_2^{\frac{2}{3}}$$

It's both concave and quasiconcave

But  $\hat{u} = [u(x_1, x_2)]^3 = x_1 x_2$  is not concave anymore.

EXERCISE: (HW)

Show that the Cobb Douglas function  $f : C \rightarrow \mathbb{R}$ ,  $C \subseteq \mathbb{R}^N$

$$f(\mathbf{x}) = A \prod_{i=1}^N x_i^{\alpha_i} \quad \text{where } \alpha_i > 0 \quad \forall i$$

is quasiconcave for all  $\mathbf{x} > \mathbf{0}$ .