

Chapter 2 Production

2.1 Production functions

N goods

N-1 inputs

1 output

$x \in \mathbb{R}_+^{N-1}$ input

$y = f(x)$ output; production function

$$f: \mathbb{R}_+^{N-1} \rightarrow \mathbb{R}_+$$

Possible assumptions on f :

* possibility of inaction: $f(0) = 0$
 \uparrow $(0, 0, \dots, 0)$

$\underbrace{\hspace{10em}}_{N-1 \text{ 0's.}}$

* free disposal (monotonicity):

if $x \geq x'$ (i.e. $x_n \geq x'_n$ for all n)

then $f(x) \geq f(x')$.

* smoothness: f is twice differentiable.

$\frac{\partial f}{\partial x_i}$ is called the marginal productivity of x_i .

* decreasing marginal productivity:

e.g. $\frac{\partial f(x)}{\partial x_i}$ decreases as x_i increases

and all other ~~the~~ inputs are held constant.

* weakly increasing returns to scale:
For all $x \in \mathbb{R}_+^{N-1}$, and all $t > 1$, $f(tx) \geq tf(x)$.

* constant returns to scale:
For all $x \in \mathbb{R}_+^{N-1}$ and all $t > 1$, $f(tx) = tf(x)$.

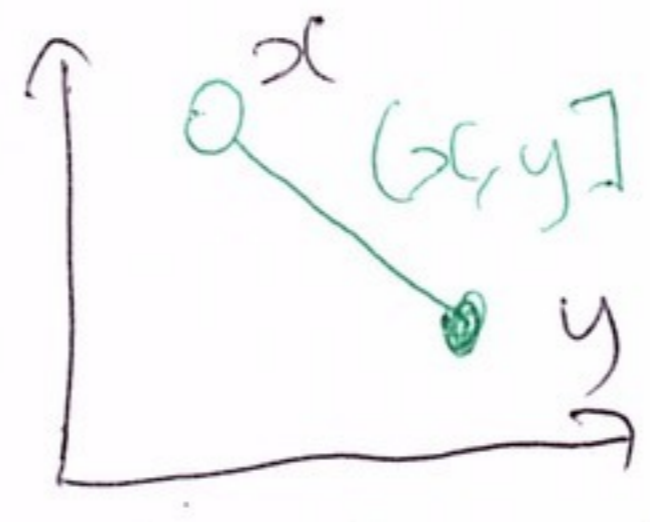
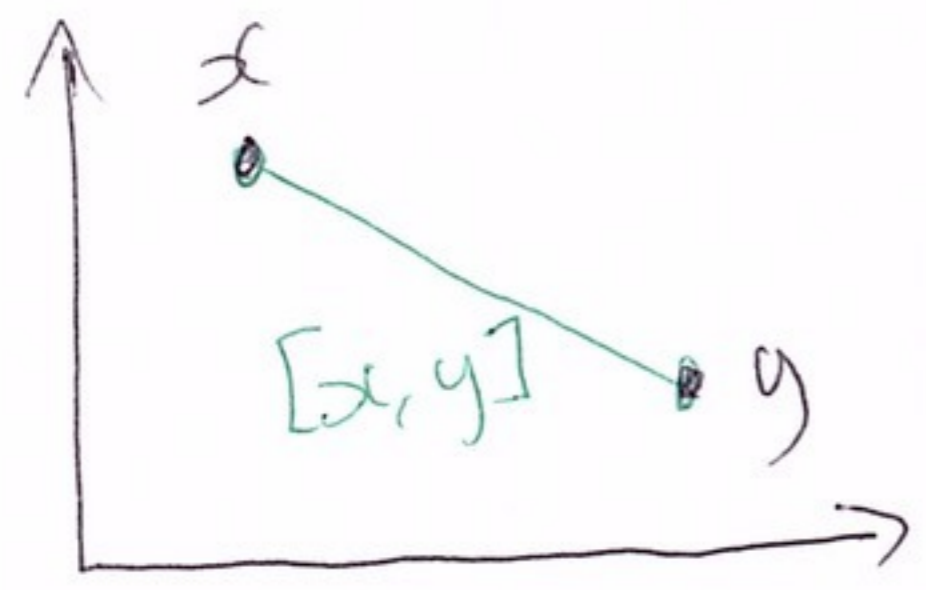
* ^(weakly) Decreasing returns to scale:
 For all ~~all~~ $x \in \mathbb{R}^{n-1}_+$, and all $t > 1$, $f(tx) \leq tf(x)$

Detour: D convex geometry

Def D.1: A closed interval between $x, y \in \mathbb{R}^n$ is defined as

$$[x, y] = \{ \underbrace{ax + (1-a)y}_{\text{convex combination}} : a \in [0, 1] \}$$

e.g.: $x = (1, 0)$
 $y = (0, 1)$
 $\frac{2}{3}x + \frac{1}{3}y = (\frac{2}{3}, \frac{1}{3})$

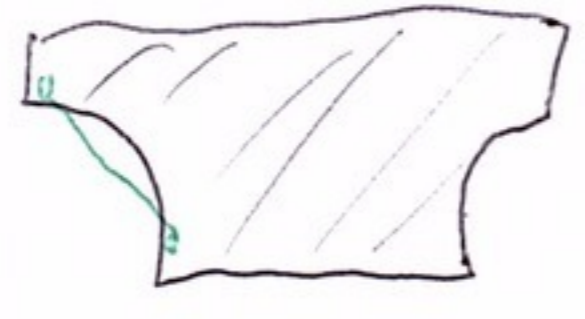
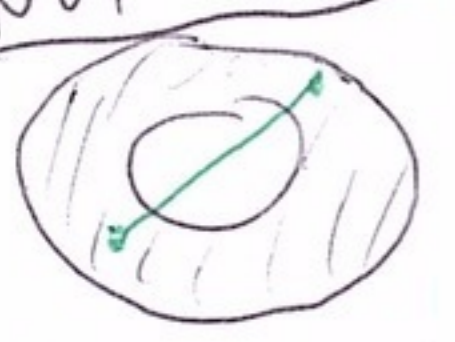


Similar definitions for $[x, y]$, etc.

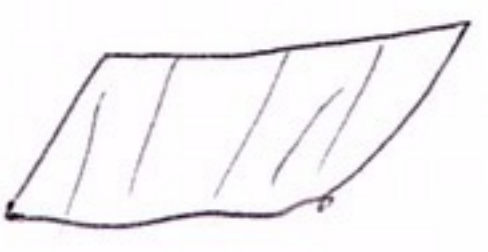
Def D.2: $X \subseteq \mathbb{R}^n$ is a convex set

if for all $x, y \in X$, $[x, y] \subseteq X$.
 "It is impossible to escape from" X by taking convex combinations."
 ← or drawing a closed interval

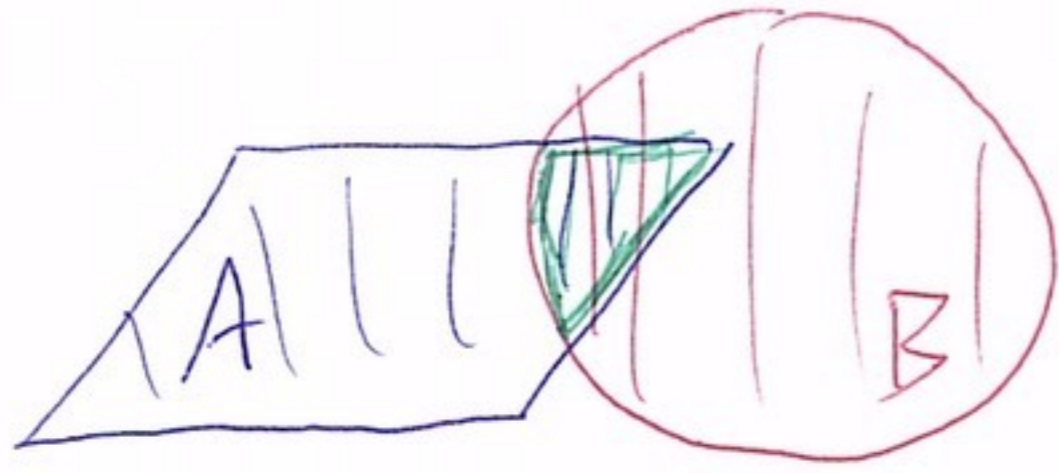
NOT CONVEX:



CONVEX:



Theorem If A and B are convex sets, then $A \cap B$ is a convex set



\square $A \cap B$ is convex

Proof Suppose $x, y \in A \cap B$. We want to prove that $[x, y] \subseteq A \cap B$. Since A is convex and $x, y \in A$, we know $[x, y] \subseteq A$. Similarly, $[x, y] \subseteq B$. We conclude $[x, y] \subseteq A \cap B$. \square

Def $f: X \rightarrow \mathbb{R}$ is a convex function

if its epigraph,

$$\text{hyper}(f) = \{ (x, y) : x \in X, y \geq f(x) \}$$

is convex set.



f is a convex function



f is NOT a convex function



f is not a convex function

Def $f: X \rightarrow \mathbb{R}$ is a concave function if $g(x) = -f(x)$ is a convex function.
Or equivalently, if its hypograph is convex, where $\text{hyp}(f) = \{(x, y) : y \leq f(x)\}$.

Caution: there is no such thing as a concave set.

Theorem If $f: X \rightarrow \mathbb{R}$ is a convex function and $X \subseteq \mathbb{R}^n$ is an open set, then f is continuous.

Theorem Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is differentiable. Then f is convex if and only if f' is weakly ~~increasing~~ increasing.

Theorem Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is twice differentiable. Then f is convex if and only if $f''(x) \geq 0$ for all x .

eg: $f(x) = x^2$ is a convex function
 $f'(x) = 2x$
 $f''(x) = 2$