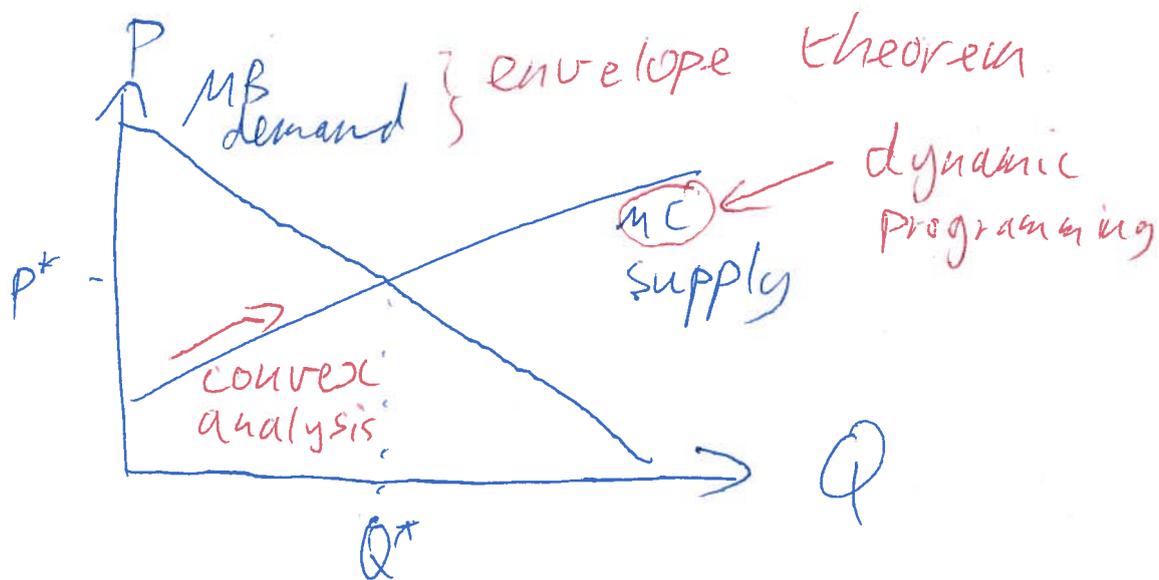


Micro 1

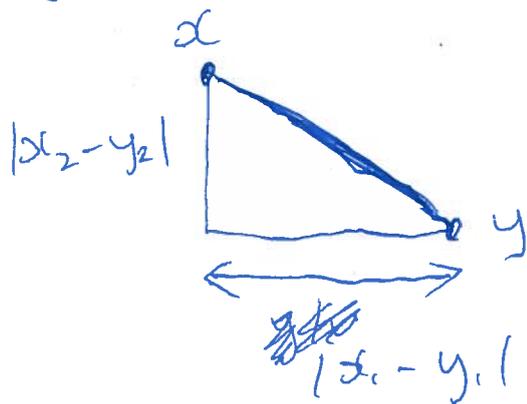
Andrew Clausen

<https://andrewclausen.net>



Bonus: Metric Spaces (CI)

Pythagoras:
$$d(x, y) = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$



Def (X, d) is a metric space if it consists of a point set X and distance metric $d: X \times X \rightarrow \mathbb{R}^+$ ← no negative numbers

two points in one number out

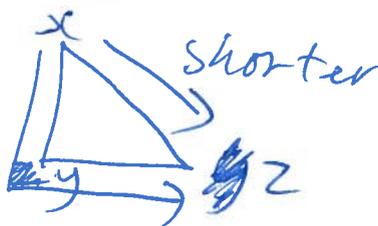
and ~~so~~ that it satisfies three properties:

(i) $d(x, y) = 0$ if and only if $x = y$ for all $x, y \in X$,

(ii) $d(x, y) = d(y, x)$ for all $x, y \in X$, and

(iii) $d(x, z) \leq \underbrace{d(x, y) + d(y, z)}_{\text{indirect route via } y}$ for all $x, y, z \in X$

— traditionally called "the triangle inequality"



or "no-shortcut property"

Examples: eg: $\mathbb{R}^2 = \{(x, y) : x \in \mathbb{R}, y \in \mathbb{R}\} = \{(0, 0), (0, 1), (0.5, 0.2), \dots\}$

* (\mathbb{R}^n, d_1) where $d_1(x, y) = \sum_{i=1}^n |x_i - y_i|$, "Manhattan metric"

* (\mathbb{R}^n, d_2) where $d_2(x, y) = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$, "Euclidean metric" (Pythagoras)

* (X, d_∞) where $X = \{f: [0, 1] \rightarrow [0, 1]\}$

and $d_\infty(f, g) = \sup_{x \in X} |f(x) - g(x)|$

supremum — like maximum

Problem with max:

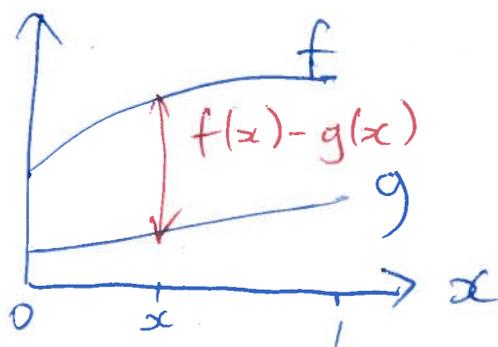
Q: What is $\max [0, 1)$?

$$\{x \in \mathbb{R} : x \geq 0 \text{ and } x < 1\}.$$

A: There is no maximum!

Solution: $\sup [0, 1) = 1$.

"smallest number that is at least as big as everything in the set."



C2 Convergence

Def. A sequence in the set X is any function with domain $\mathbb{N} = \{0, 1, 2, 3, \dots\}$ and co-domain X . Usually $f: \mathbb{N} \rightarrow X$.

But special notation: x_0, x_1, x_2, \dots

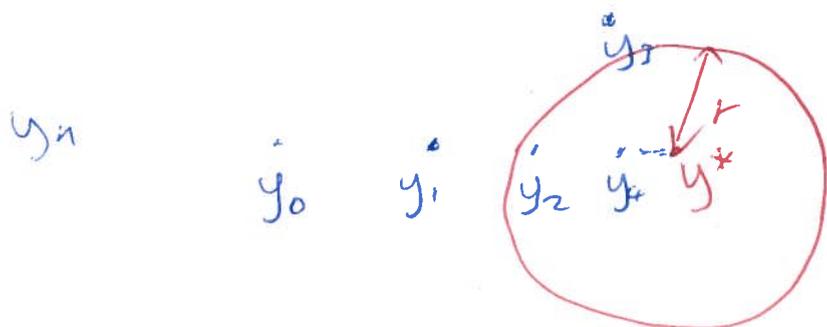
or $\{x_n\}_{n=0}^{\infty}$ or x_n .

eg: $x_n = 2n = 0, 2, 4, 6, 8, \dots$

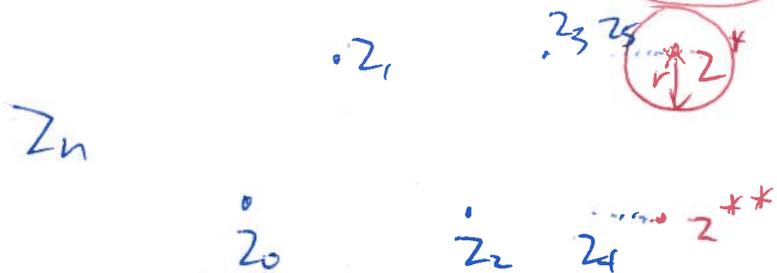
Def Suppose x_n is a sequence in a metric space (X, d) . We say that x_n converges to $x^* \in X$ (or write $x_n \rightarrow x^*$) if for every $r > 0$, there exists an $N \in \mathbb{N}$ such that $d(x_n, x^*) < r$ for every $n \geq N$.



need $N=3$

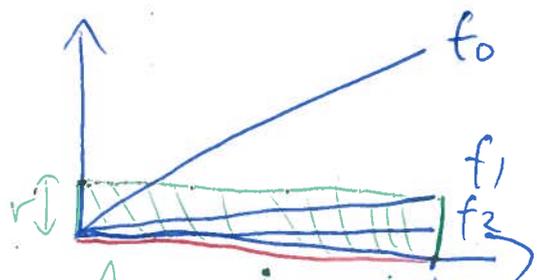


need $N=4$



NOT a convergent sequence

There is no N that excludes all of the "bottom" points



$f_n(x) = \frac{1}{n}x$
in (X, d_∞) from before
 $f_n \rightarrow 0$ (in red)

Any function g inside \square has $d(0, g) \leq r$.

2.1 Production functions

N types of goods (and services, etc.)

$N-1$ #inputs

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

$y = f(x)$ output

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

Possible assumptions:

* possibility of inaction: $f(0) = 0$

* free disposal (monotonicity):

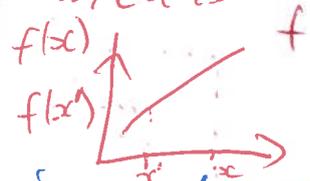
If $x \geq x'$ (i.e. if $x_n \geq x'_n$ for all $n \in \{1, \dots, N-1\}$)

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

music: monotone is the same note over and over

maths: monotone increasing means

"up and up and up"



We can also assume f is strictly (monotonically) increasing, i.e. if $x > x'$

($x \geq x'$ and for some n , $x_n > x'_n$)

then $f(x) > f(x')$.

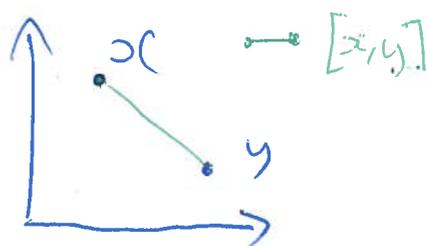
* Smoothness: f is twice continuously differentiable. $\frac{\partial}{\partial x_1} f(x)$ is the marginal productivity of input good 1.

1. Detour (to generalise \downarrow MP, \downarrow RTS)

D Convex geometry

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

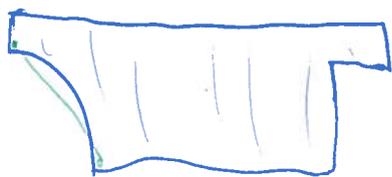
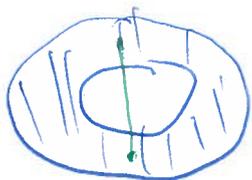
$$[x, y] = \{ ax + (1-a)y : a \in [0, 1] \}$$



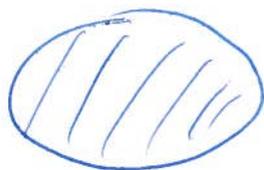
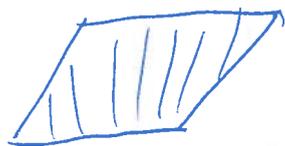
convex combination

Def $X \subseteq \mathbb{R}^n$ is a convex set if for all $x, y \in X$, the interval $[x, y] \subseteq X$.

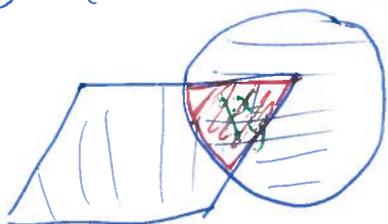
Non-convex sets



Convex sets:



Theorem The intersection of convex sets is convex.



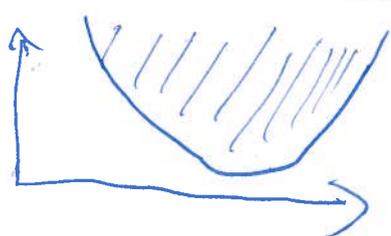
□ ~~convex~~ intersection

Proof: Suppose A and B are convex sets. We need to show that for every $x, y \in A \cap B$, the interval $[x, y] \subseteq A \cap B$.
 Since $x, y \in A \cap B$, we know $x, y \in A$.
 Since A is a convex set, we know $[x, y] \subseteq A$. Similarly, $[x, y] \subseteq B$.
 Combining, $[x, y] \subseteq A \cap B$. □

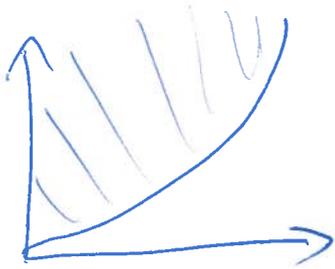
Def $f: X \xrightarrow{\subseteq \mathbb{R}^n} \mathbb{R}$ is a convex function if its hypergraph

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

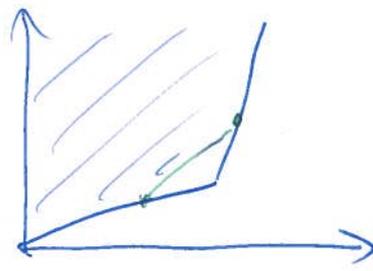
is a convex set



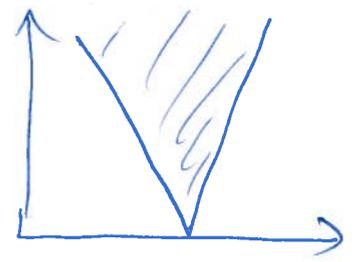
□ $\text{hyper}(f) \subseteq X \times \mathbb{R}$



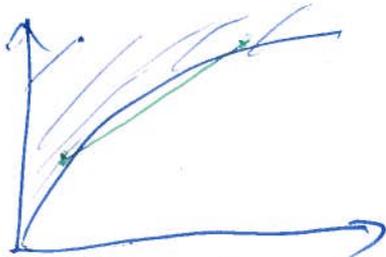
convex



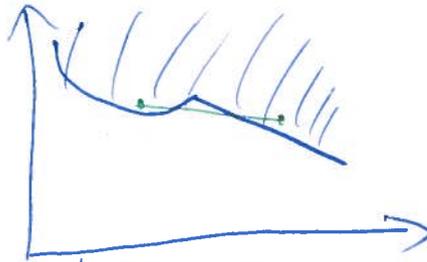
convex



convex



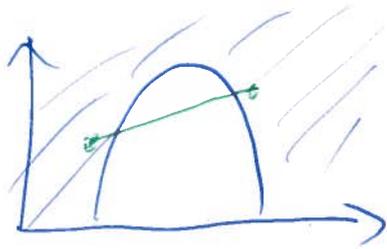
NOT convex



NOT convex



NOT convex



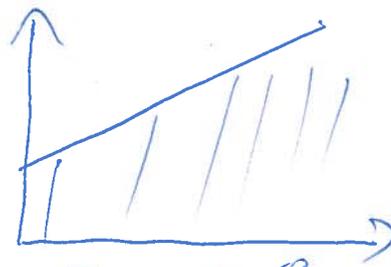
NOT convex

Def $f: X \rightarrow \mathbb{R}$ is a concave function if $g(x) = -f(x)$ is a convex function (or equivalently, its hypograph, $\text{hypo}(f) = \{(x, y) : f(x) \geq y\}$ is a convex set).

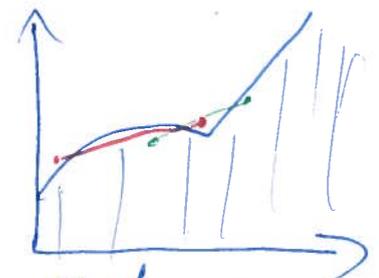
Confusion: ① convex functions and sets
② ~~convex~~ no such thing as a concave set.



concave



concave
(and convex)



not concave
(and not convex)

Bonus

⊛ 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.

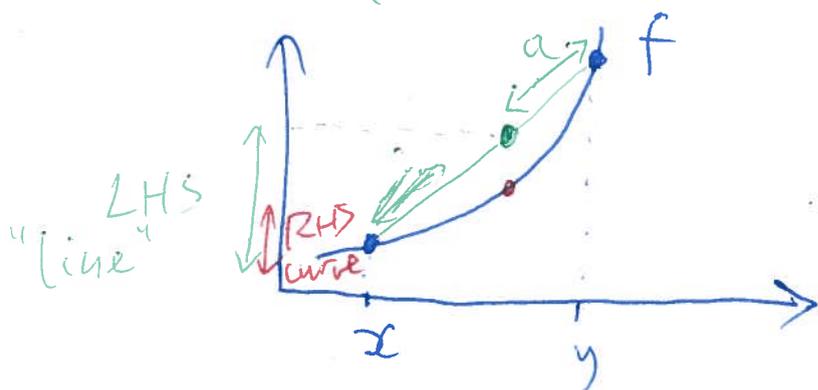
Theorem Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is differentiable. Then f is convex if and only if its derivative f' is (weakly) monotonically increasing.

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 $x \in \mathbb{R}$.

Theorem A function $f: X \rightarrow \mathbb{R}$ is convex if and only if X is convex and for all $x, y \in X$ and all $a \in (0, 1)$
 weight

$$a f(x) + (1-a) f(y) \geq f(ax + (1-a)y)$$

line curve



2.1 Production Functions (again)

* decreasing marginal productivity:

~~eg~~ for good 1: $\frac{\partial}{\partial x_1} f(x)$ is decreasing as x_1 increases (and x_2, \dots, x_{N-1} are held fixed)

$\Leftrightarrow g(x_1) = f(x_1; x_2, \dots, x_{N-1})$ is concave for all (x_2, \dots, x_{N-1}) .

* increasing returns to scale:

for all $x \in \mathbb{R}_+^{N-1}$ and all $t > 1$,

$$f(tx) \geq t f(x).$$

eg $t=2$: double inputs double outputs

* constant ~~RTS~~ RTS: for all $x \in \mathbb{R}_+^{N-1}$ and all $t > 0$, $f(tx) = t f(x)$.

* decreasing RTS: for all $x \in \mathbb{R}_+^{N-1}$ and all $t > 1$, $f(tx) \leq t f(x)$.

~~Assump~~ This assumption is sometimes made to capture non-tradeable resources — but this goes against CTE philosophy / motivation.

* concave: f is a concave function.

- if f is smooth & concave, then f has decreasing marginal productivity
- if f is concave and has the possibility of inaction, then f has decreasing (constant) RTS.

Proof: Want to show: $f(tx) \leq tf(x)$,
~~that~~ for all $t > 1$. Let $a = \frac{1}{t} \in (0, 1)$.

Recall f is concave

$$\Leftrightarrow \underbrace{af(X) + (1-a)f(X')} \leq f(aX + (1-a)X')$$

Let $X = tx$ and $X' = 0$.

~~the~~ inequality becomes: ~~inaction~~

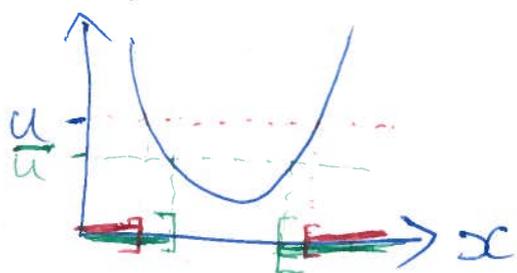
$$\frac{1}{t} f(tx) + (1 - \frac{1}{t}) f(0) \leq f(\frac{1}{t}tx + (1 - \frac{1}{t})0)$$

$$\Leftrightarrow \frac{1}{t} f(tx) + 0 \leq f(x)$$

$$\Leftrightarrow f(tx) \leq tf(x). \quad \square$$

Detour: back to convex analysis

Def The upper contour set of a function $f: X \rightarrow \mathbb{R}$ at level u is $\{x: f(x) \geq u\}$.



— $\{x: f(x) \geq u\} \subseteq X$
 — $\{x: f(x) \geq u\}$

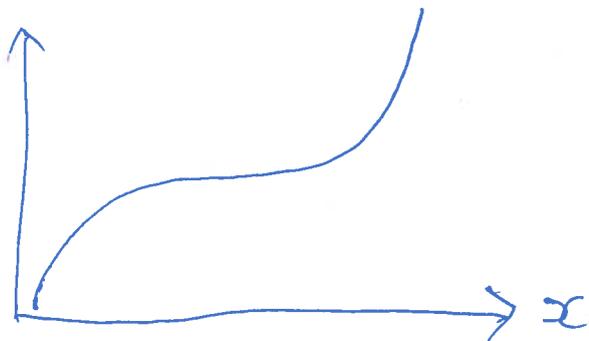
Confusion:

* superficially similar to hypergraphs, but details are totally different!

Similarly, the lower contour set of f at level u is $\{x \in X: f(x) \leq u\}$.

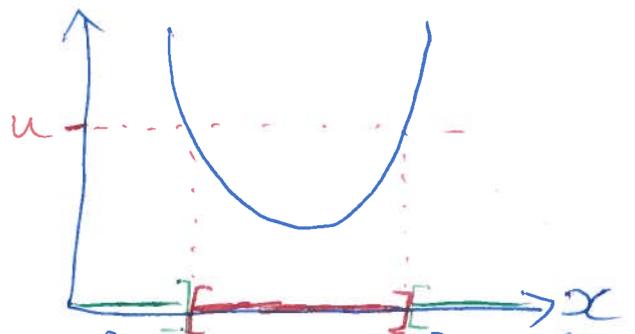
Def: $f: X \rightarrow \mathbb{R}$ is a quasi-convex function if all of its lower ~~contours~~ sets (i.e. at each level) are convex sets.

Similarly, f is a quasi-concave function if its upper contour sets are convex sets.



Quasi-convex? ✓

Quasi-concave? ✓



Quasi-convex? ✓ -

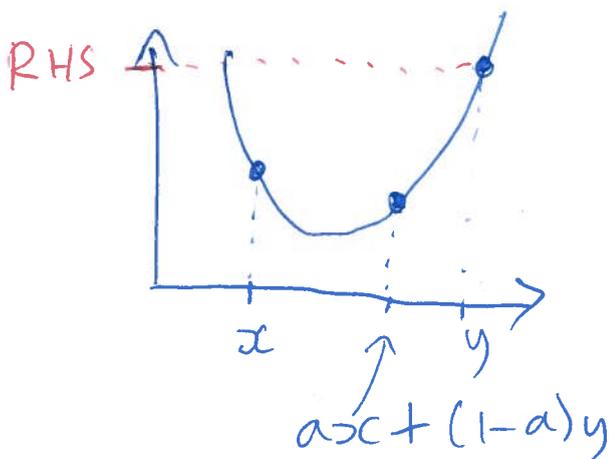
Quasi-concave X -

Theorem If f is a convex function, then f is a quasi-convex function.

Theorem $f: X \rightarrow \mathbb{R}$ is a quasi-convex function if and only if X is convex and for all $x, y \in X$ and $a \in (0, 1)$

$$f(ax + (1-a)y) \leq \max \{f(x), f(y)\}$$

the curve



return from detour:

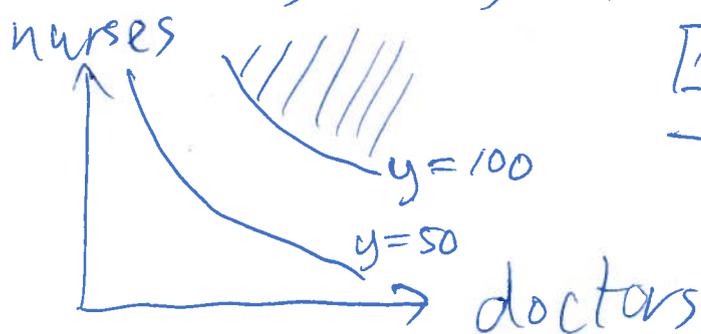
* f is a quasi-concave function.

Rationale: Let y be a target output level. The upper contour set at level y is

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

i.e. the set of inputs that meet the production target.

Amounts to assuming that it is possible to "mix" different ways of meeting any production target.



 upper contour set
 isoquant (same output quantity along the isoquant)

2.2 Profit maximisation

Only consider perfect competition.

Firms can't influence ~~the~~ prices
 $p \in \mathbb{R}_+$ (output) and $w \in \mathbb{R}_+^{N-1}$ (input).

e.g. wages (and equipment, etc.)

The firm's profit function is:

$$\pi(p; w) = \max_{x \in \mathbb{R}_+^{N-1}} \underbrace{p f(x)}_{\text{revenue}} - \underbrace{w \cdot x}_{\text{costs}}$$

$$= p f(\underbrace{x(p; w)}) - w \cdot \underbrace{x(p; w)}$$

factor demand function

First-order condition with respect to x_i :

$$p \frac{\partial f(x^*)}{\partial x_i} = w_i.$$

Divide FOC - x_i by FOC - x_j

$$\frac{\frac{\partial f(x^*)}{\partial x_i}}{\frac{\partial f(x^*)}{\partial x_j}} = \frac{w_i}{w_j}.$$

Example: Suppose music recordings are produced from musician labour and technician labour.

Answer:

r royalties (sales price)

l_m musician labours

l_t technician labour

w_m, w_t wages

$f(l_m, l_t)$ song output

$$\pi(r; w_m, w_t) = \max_{l_m, l_t} r f(l_m, l_t) - w_m l_m - w_t l_t.$$

Example Glycerine is a by-product of diesel production, produced from waste organic material.

Answer:

w waste input

$g(w)$ glycerine output

$d(w)$ diesel output

p^w, p^g, p^d prices of waste, glycerine, diesel

$$\pi(p^g, p^d; p^w) = \max_w \underbrace{p^g g(w) + p^d d(w)}_{\text{revenue}} - \underbrace{p^w w}_{\text{cost}}$$

Example PET (polyethylene)

is made from ethylene, which is made from crude oil. Model a vertically integrated firm that buys oil and sells plastic.

Answer

x crude input

$e = f(x)$ ethylene output

$y = g(e)$ plastic output

P^x, P^y prices of ~~oil~~ crude and

$$\pi(P^y; P^x) = \max_x P^y g(f(x)) - P^x x$$

HW: Generalise to buy/sell ethylene.