

2.1 Production functions

N goods

1 output

$N-1$ inputs

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

$$y = f(x)$$

↑ output ↑ inputs

} production
function

possible assumptions:

$$f(0) = 0$$

$$\uparrow \in \mathbb{R}^{N-1}$$

i.e. $\begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$



possibility of inaction

free disposal (or monotonicity).

If $x \leq x'$, i.e. $x_n \leq x'_n$
for all n , then $f(x) \leq f(x')$.

Smoothness f is twice
differentiable. Each
partial derivative

$$\frac{\partial f}{\partial x_i}$$

is called the marginal
productivity of good x_i .

decreasing marginal
productivity

$\frac{\partial f(x)}{\partial x_i}$ is decreasing in

x_i .

Weakly increasing returns

I.O. scale
For all $x \in \mathbb{R}_+^{N-1}$, and all
scales $t > 1$:

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

Constant returns to scale

For all $x \in \mathbb{R}_+^{N-1}$ and all $t > 0$,
 $f(tx) = tf(x)$.

Weakly decreasing returns to scale

For all $x \in \mathbb{R}_+^{N-1}$ and all $t > 1$,
 $f(tx) \leq tf(x)$.

Concavity

f is a concave function.

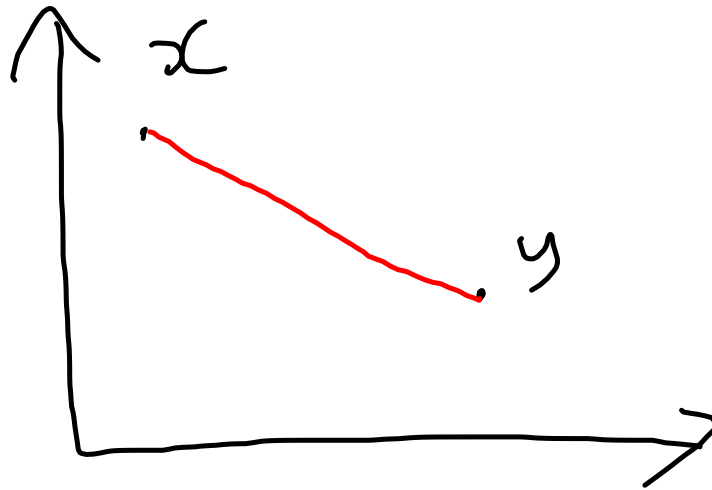
detour: A.3 (Geometry)

Def ~~A~~^{The} closed interval convex combination

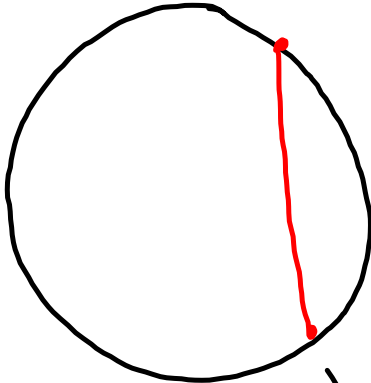
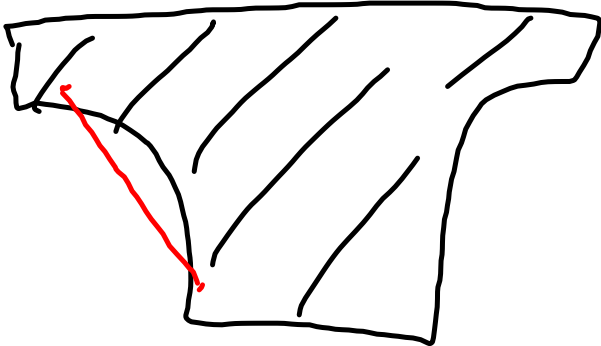
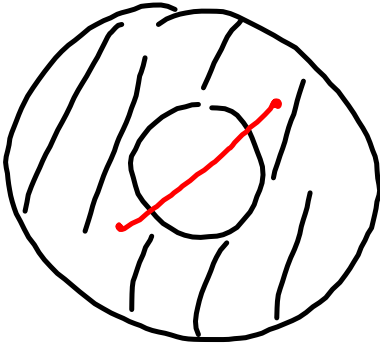
between $x, y \in \mathbb{R}^n$ is

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

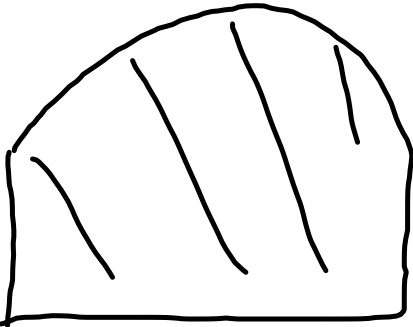
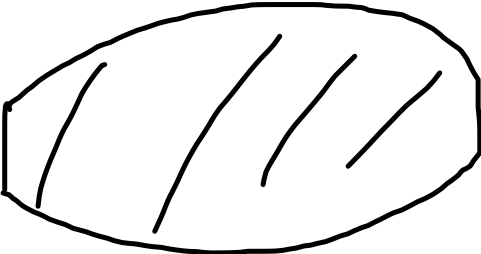
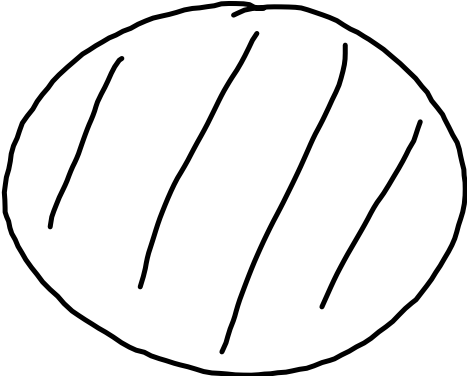
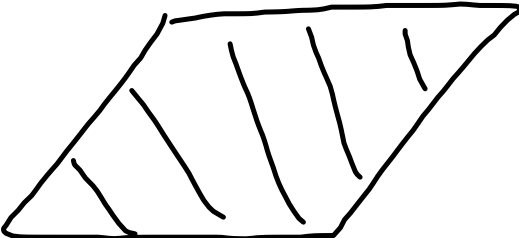
Similar definitions for $[x, y)$, (x, y) , etc.



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



(boundary)



Theorem The intersection of convex sets is convex.

Proof Suppose A and B are convex sets. We need to show that if $x, y \in A \cap B$ then $[x, y] \subseteq A \cap B$.

If $x, y \in A \cap B$ then $x, y \in A$.

Since A is convex,

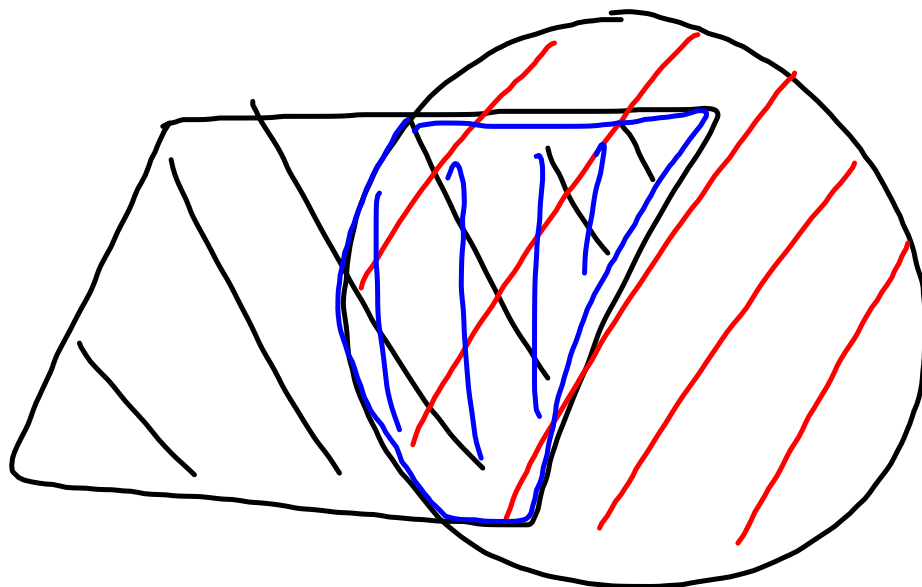
$$[x, y] \subseteq A.$$

By the same logic,

$$[x, y] \subseteq B.$$

Therefore $[x, y] \subseteq A \cap B$.

We conclude that $A \cap B$
is a convex set. \square



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

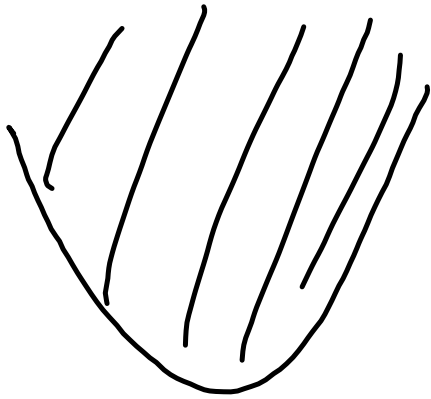
a convex function

if its hypergraph

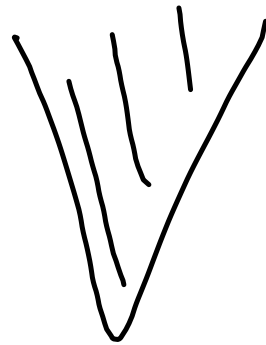
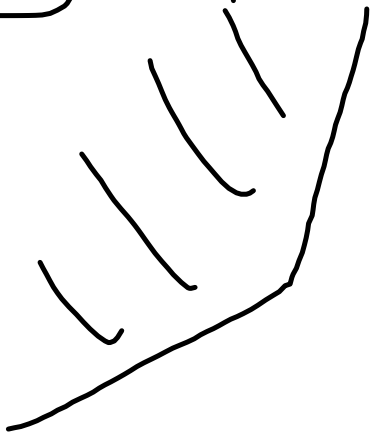
$$\{(x, y) : x \in X, y \geq f(x)\}$$

is a convex set.

Note: a function can only be convex if its domain is convex.

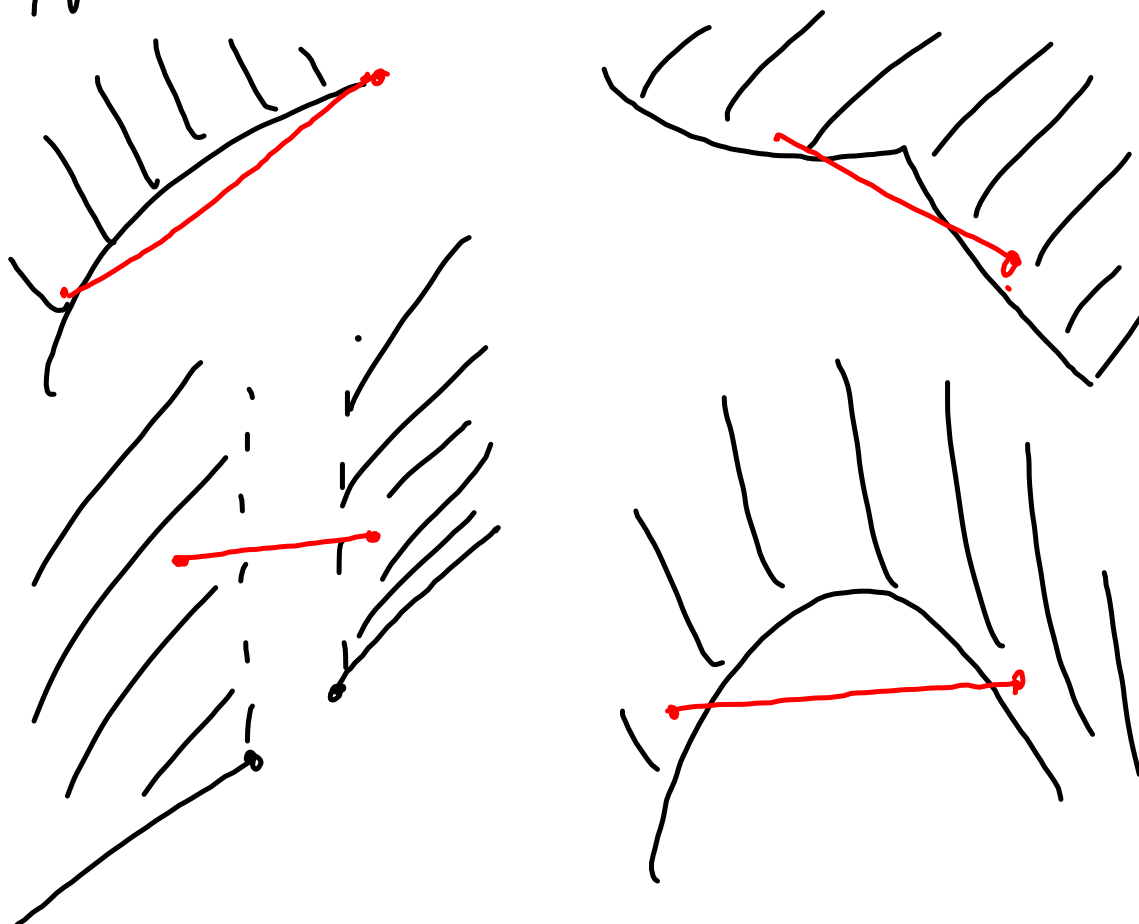


iii hypergraph



are convex functions

NOT convex:



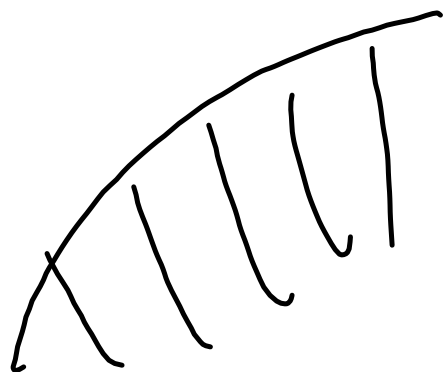
Def $f: X \rightarrow \mathbb{R}$ is a
concave function if

its hypograph

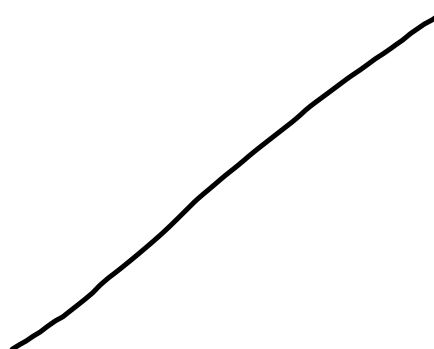
$$\{(x, y) : x \in X, y \leq f(x)\}$$

is a convex set.

Theorem f is a concave function
if and only if $-f$ is a
convex function.



concave



convex and
concave

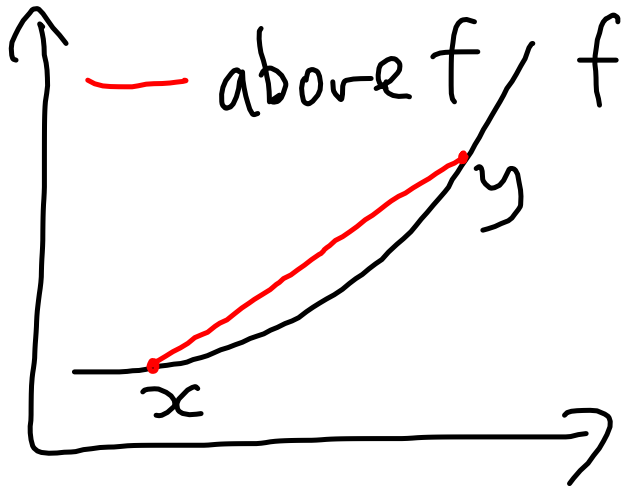


neither convex
nor concave

Theorem If $f: X \rightarrow \mathbb{R}$ is a convex function and X is an open set in (\mathbb{R}^n, d_2) , then f is continuous.

Theorem Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is differentiable. Then f is a convex function if and only if its derivative f' is weakly 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 for all $x, y \in X$, and $a \in (0, 1)$

$$\underbrace{af(x) + (1-a)f(y)}_{\text{line}} \geq \underbrace{f(ax + (1-a)y)}_{\text{curve}}$$

Back to production functions.

Assumption f is a concave function. *and smooth*

Claim If f is concave, then f has decreasing marginal productivity.

claim If f is smooth, concave, and has the possibility of inaction, then f has weakly decreasing returns to scale.

Proof

We want to show

$$f(tx) \leq tf(x)$$

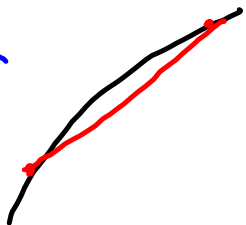
for all $x \in \mathbb{R}_+^{N-1}$ and all $t > 1$.

Let $a = \frac{1}{t}$. So $a \in (0, 1)$.

We will apply the theorem to the points $0 \in \mathbb{R}^{N-1}$ and tx with $a = \frac{1}{t}$.

$$\underbrace{a f(tx) + (1-a)f(0)}_{\text{line}} \leq f(atx + (1-a)0)$$

(Note: In the original image, the terms $(1-a)f(0)$ and $(1-a)0$ are crossed out with blue lines. A red arrow labeled "weight" points to the a coefficient, and a blue arrow labeled "=0" points to the $f(0)$ term.)



$$\Rightarrow a f(tx) \leq f(atx).$$

$$\Rightarrow \frac{1}{t} f(tx) \leq f\left(\frac{1}{t} tx\right)$$

$$\Rightarrow \frac{1}{t} f(tx) \leq f(x)$$

$$\Rightarrow f(tx) \leq t f(x) \quad \square$$

Upper contour sets are
Convex (another assumption)
 f is concave \nearrow
(skipped)

Compact \Rightarrow closed, bounded

\uparrow any metric space

closed, bounded

\Rightarrow compact

\uparrow only in (\mathbb{R}^n, d_2) .

If (X, d) and (Y, d)
are metric spaces
and $K \subseteq X \cap Y$, then
if K is compact in X ,
then K is also compact
in Y .

Proof

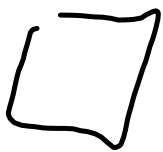
K is compact in X

\Leftrightarrow every $x_n \in K$

has a subsequence y_n

s.t. $y_n \xrightarrow{d} y^* \in K$

$\Leftrightarrow K$ is compact in Y .



2.2 Profit maximization

$p > 0$ output price

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

(w including wages,
raw materials, parts,
& rent)

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

$$= pf(x(p; w)) - w \cdot \underbrace{x(p; w)}_{\text{factor demand}}$$

F.O.C. for optimal x_i :

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

FOC x_i / x_j :

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

Example 2.1

r royalties for each song

l_m musician input

l_t technician input

w_m } wages

w_t } output of songs

$f(l_m, l_t)$

$\Pi(r; w_m, w_t)$

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

Example 2.2

w waste input

$g(w)$ glycerine output

$d(w)$ diesel output

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

$$\pi(P^g, P^d; P^w) = \max_w P^g g(w) + P^d d(w) - P^w w.$$

Example 2.3

x crude oil input

$e = f(x)$ ethylene output
 " allocated to plastic

$y = g(E)$ plastic output

p^x, p^e, p^y prices of oil, ethylene, plastic

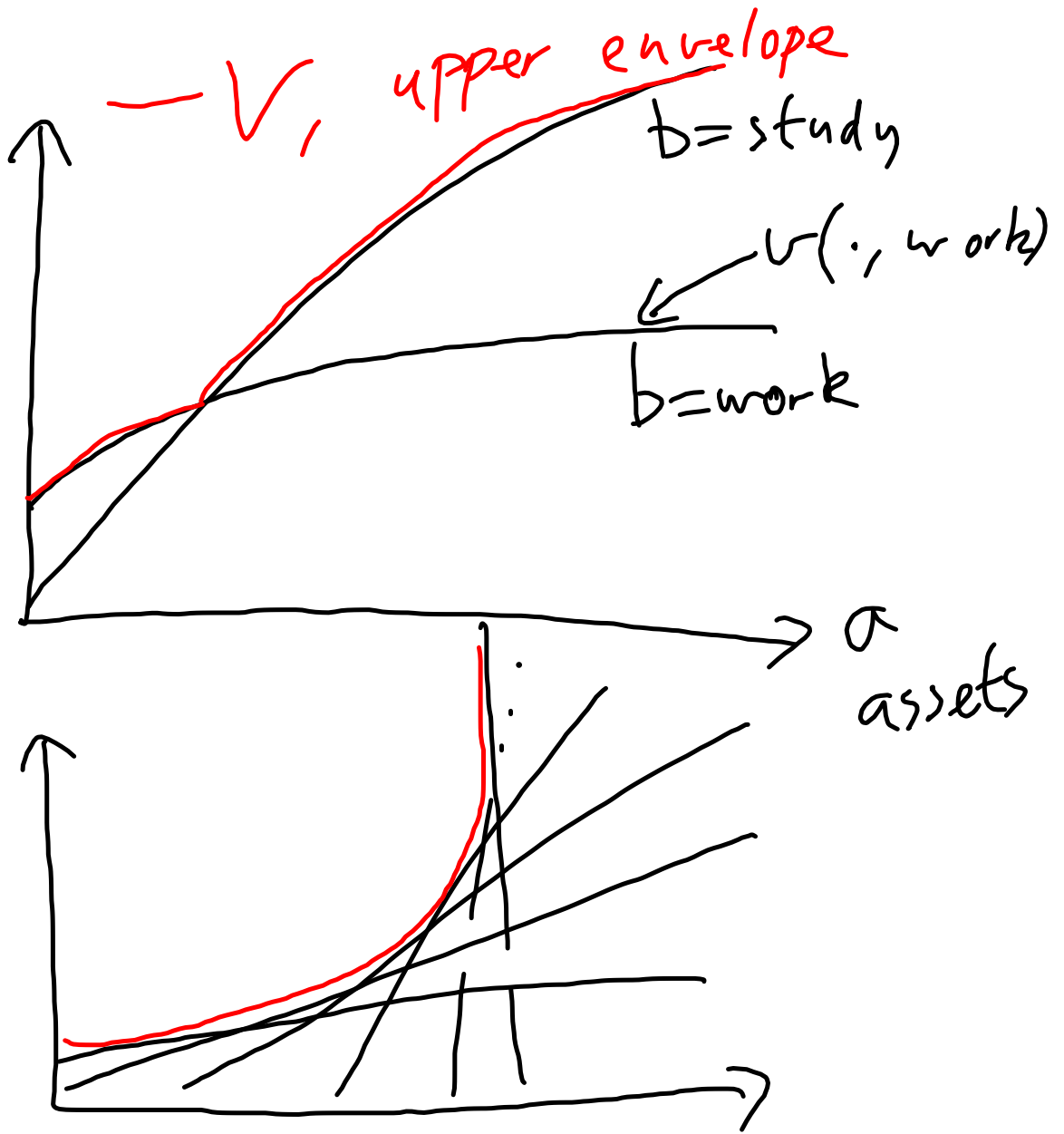
$$\pi(p^x, p^e, p^y) = \max_{x, E} p^e (f(x) - E) + p^y g(E) - p^x x.$$

ethylene sold

2.3 Upper envelopes and value functions

$$V(a) = \max_{b \in B} v(a, b)$$

$V(a)$ is the value function, where a is the state variable.
 $\max_{b \in B}$ indicates the choice set B , where b is the choice variable.
 $v(a, b)$ is the objective function.



Theorem Let $v: \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}$

be a differentiable function,

$$V(a) = \max_{b \in \mathbb{R}^m} v(a, b)$$

and $b(a) \in \arg \max_{b \in \mathbb{R}^m} v(a, b)$.

If V is differentiable at a ,
then

$$V'(a) = \left. \frac{\partial v(a, b)}{\partial a} \right|_{b=b(a)}$$

$$\text{or } V'(a) = v_a(a, b(a)).$$

Note: could have tried to differentiate $a \mapsto v(a, b(a))$, which would involve $b'(a)$.