

Micro

(Wed Aug 23)

1. firm \rightarrow Varian.
2. Consumer Theory \rightarrow MWG
3. Choice under uncertainty.

Individual agent
↓
Single decision problem

① Firm Theory :-

Assumptions:-

1. firms are price takers
2. Tech is exogenous
3. firms maximize profits.

Technology :-

- n commodities, $y = (y_1, \dots, y_n)$ ($y \in \mathbb{R}^n$)
 - $y_i < 0$, i is an input
 - $y_i > 0$, i is an output.

- Production Set : $Y \subseteq \mathbb{R}^n$

what the firm can do.

$$n=3$$

$$y = (-2, 1, -3)$$

Firm's function : g units of something as input

3 units of something as input
1 Output

Properties of Production Sets! —

1. $\gamma \neq \emptyset$
2. γ is closed
3. No free lunch. If $y \in \gamma$ and $y \geq 0$, then $y = 0$

there are \downarrow as inputs. \downarrow then you can't make anything.

The following are sometimes assumed! —

4. Shutdown is possible: $0 \in \gamma$
5. Free disposal: If $y \in \gamma$ and $y' \leq y$, then $y' \in \gamma$.
(You can take more input, and not lead to increased output, say you get rid of it.)
6. Irreversibility: $y \in \gamma$, $y \neq 0$, then $-y \notin \gamma$
You can't take the output to make the input.

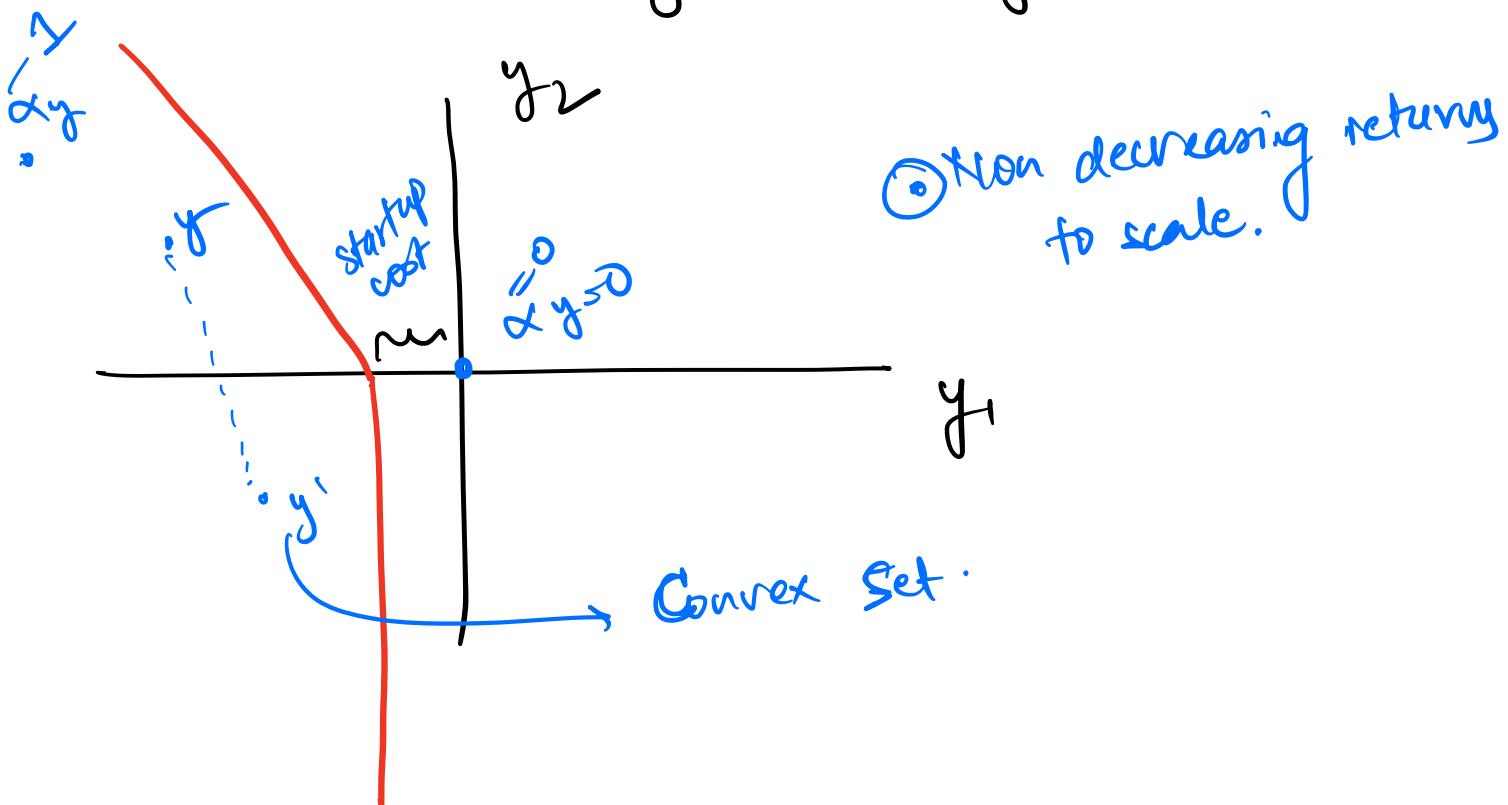
Returns to scale & Convexity

γ has:

- non increasing returns to scale if
 $y \in \gamma \Rightarrow \alpha y \in \gamma \text{ } \forall \alpha \in [0, 1]$
- non decreasing returns to scale if.
 $y \in \gamma \Rightarrow \alpha y \in \gamma \text{ } \forall \alpha \geq 1$
- Constant returns to scale
 $y \in \gamma \Rightarrow \alpha y \in \gamma \text{ } \forall \alpha \geq 0$

γ is convex. if :

$$y, y' \in \gamma \Rightarrow \alpha y + c(1-\alpha)y' \in \gamma \text{ & } \alpha \in [0, 1]$$



why do we "usually" assume non increasing returns?

- Modelling only inputs under our control.

Theorem

There is a constant returns production set

$$Y' \subseteq \mathbb{R}^{n+1} \text{ s.t. } Y = \{y \in \mathbb{R}^n : (y-1) \in Y'\}$$

Proof Define $Y' = \{y' \in \mathbb{R}^{n+1} : y' = \alpha(y - 1)$
for some $y \in Y$ & $\alpha > 0\}$

Check that Y' is constant returns.

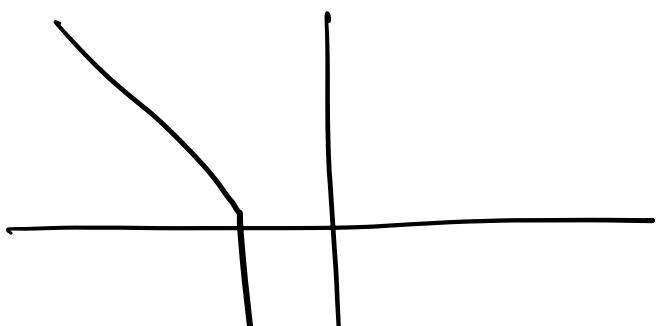
Efficiency :-

Technological efficiency:

A production plan y is technologically efficient if there does not exist

$$y' \in Y \text{ s.t. } \underbrace{y' > y}_{\downarrow}$$

$y'_i \geq y_i$ for all i & $y'_i > y_i$ for some i .



} efficient pts are
going to be on the

boundary.

Aug 28 (second class)

Transformation frontier.

Firm, production set \mathcal{Y}

The transformation function $T: \mathbb{R}^n \rightarrow \mathbb{R}$

$T(y) < 0 \Leftrightarrow y$ is inefficient

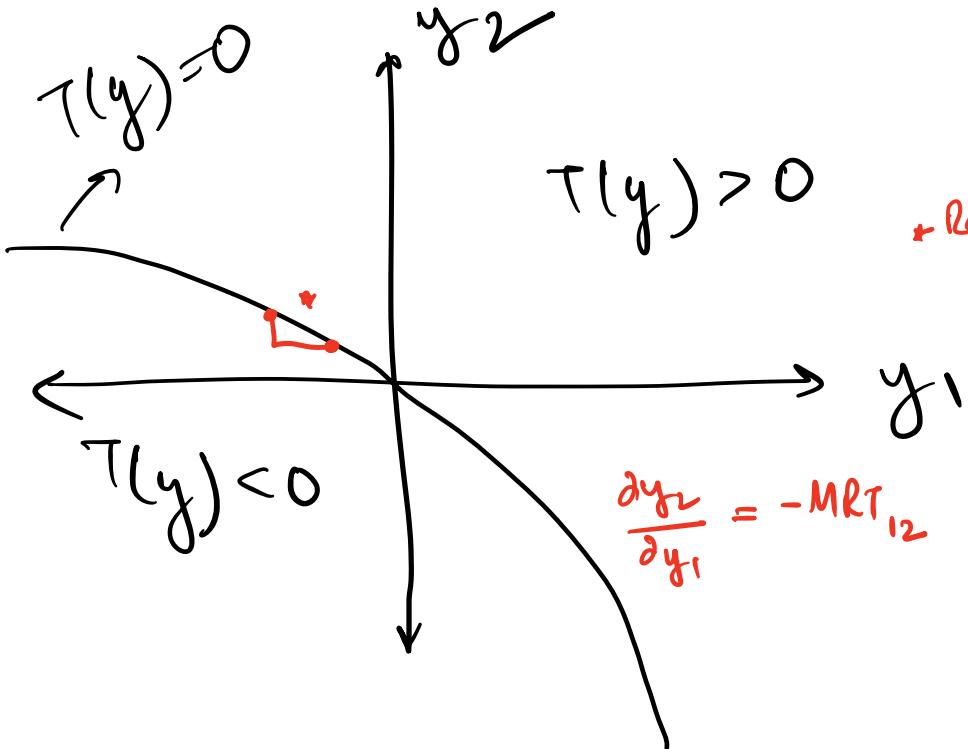
$T(y) = 0 \Leftrightarrow y$ is efficient

$T(y) > 0 \Leftrightarrow y$ is infeasible.

$\Rightarrow \mathcal{Y} = \{y \in \mathbb{R}^n : T(y) \leq 0\}$ | ~~To check~~

Production frontier

$\{y \in \mathbb{R}^n : T(y) = 0\}$



+ Rate of substitution.

Let's fix $\bar{y}_3, \dots, \bar{y}_n$. Consider y_1, y_2, \dots

Define a function $y_2(y_1)$ by

$$T(y_1, y_2(y_1), \bar{y}_3, \dots, \bar{y}_n) = 0 \quad \text{constants.}$$

implicit fn of y_1

Pick some y_1 , what y_2 gives me $T(y) = 0$

Let's differentiate both sides w.r.t y_1

$$\frac{\partial T}{\partial y_1} + \frac{\partial T}{\partial y_2} \cdot \frac{\partial y_2}{\partial y_1} = 0$$

slope in the picture

$$\frac{\partial y_2}{\partial y_1} = - \frac{\partial T / \partial y_1}{\partial T / \partial y_2}$$

Works for any 2 goods

(Marginal Rate of Transformation)

$$\frac{\partial y_k}{\partial y_j} = - \frac{\partial T / \partial y_j}{\partial T / \partial y_k} \Big|_{y=\bar{y}} = - MRT_{jk}(\bar{y})$$

My
Ques:

Special: Many inputs, one output

Care

Inputs $(x_1, \dots, x_m) \geq 0$

Output: $y \geq 0$

Production function: $f(x)$

Production set: $Y = \{ \underbrace{(-x_1, -x_2, \dots, -x_m, y)}_m : y \leq f(x_1, \dots, x_m) \}$

$x_i \geq 0 \forall i \in \{1, \dots, m\}$

$y \leq f(x_1, \dots, x_m)$

$\tau(z) = y - f(x), z = (-x, y)$

Efficient frontier is

$$\tau(z) = 0$$

$$\rightarrow y - f(x) = 0$$

$$\rightarrow y = f(x)$$

Example

Cobb-Douglas

- 2 inputs, capital (x_K) + labor (x_L)
- Output y
- $f(x_K, x_L) = x_K^\alpha x_L^\beta$, $\alpha + \beta = 1$
- $T = \{(-x_K, -x_L, y) : x_K, x_L \geq 0, y \leq x_K^\alpha x_L^\beta\}$
- $T = (-x_K, -x_L, y) = y - x_K^\alpha x_L^\beta$

MRT between capital + output.

$$MRT_{(-x_K), y} = \frac{\partial T / \partial (-x_K)}{\partial T / \partial y}$$

$$= - \frac{\partial T / \partial x_K}{\partial T / \partial y}$$

$$= - \frac{-\alpha x_K^{\alpha-1} x_L^\beta}{1}$$

$$= \alpha \left(\frac{x_L}{x_K} \right)^\beta$$

Marginal product of capital, MPK.

$$\begin{aligned} \text{MPK} &= f_{x_K}(x_K, x_L) \rightarrow \text{from undergrad.} \\ &= \frac{\partial f}{\partial x_K} \\ &= \alpha x_K^{\alpha-1} x_L^\beta \\ &= \alpha \left(\frac{x_L}{x_K} \right)^\beta \end{aligned}$$

The method now is more a general way of finding sol'n.

MRT for capital & labor

$$\begin{aligned} \text{MRT}(-x_K), (-x_L) &= \frac{\frac{\partial T}{\partial (-x_K)}}{\frac{\partial T}{\partial (-x_L)}} \\ &= \frac{\frac{\partial T}{\partial x_K}}{\frac{\partial T}{\partial x_L}} \\ &= \frac{-\alpha x_K^{\alpha-1} x_L^\beta}{-\beta x_K^\alpha x_L^{\beta-1}} \\ &= \frac{\alpha}{\beta} \left(\frac{x_L}{x_K} \right) \end{aligned}$$

$$\frac{\partial (-x_L)}{\partial (-x_K)} = -MRT_{(-x_K, -x_L)}$$

$$\frac{\partial x_L}{\partial x_K} = -\frac{\alpha}{\beta} \frac{x_L}{x_K}$$

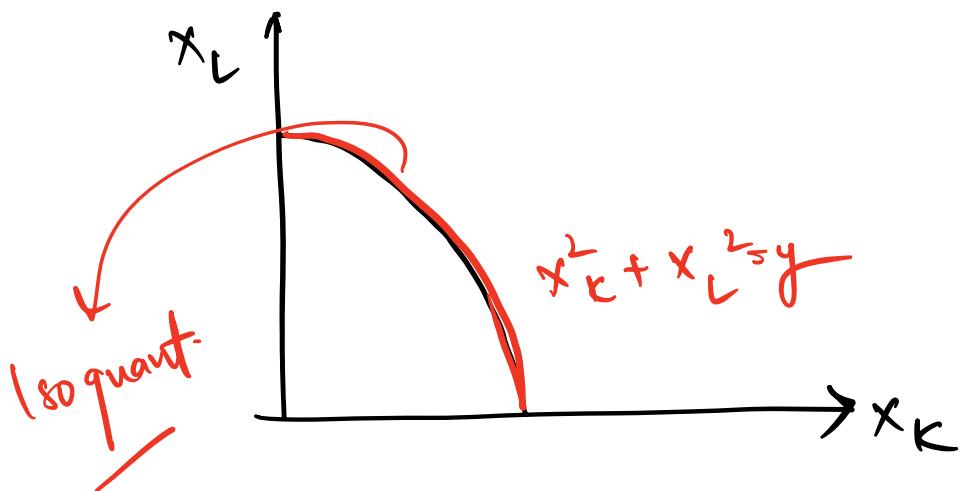
single output, fix y

$$V(y) = \{x \in \mathbb{R}^n : f(x) \geq y\} \rightarrow \begin{matrix} \text{Input} \\ \text{requirement} \\ \text{set.} \end{matrix}$$

$$Q(y) = \{x \in \mathbb{R}^n : f(x) = y\} \rightarrow \text{Isquant.}$$

Example:

$$f(x_K, x_L) = x_K^2 + x_L^2 = y$$



This is all fine as the inputs here are considered positive as we are only working with inputs.

With a single output, we can define the marginal rate of technical substitution between two inputs

$$MRTS_{x_K, x_L} = \frac{\frac{\partial f}{\partial x_K}}{\frac{\partial f}{\partial x_L}}$$

$$\frac{\partial x_L}{\partial x_K} = -MRTS_{x_K, x_L} = -\frac{f'_{x_K}}{f'_{x_L}}$$

In the example

$$\frac{\partial x_L}{\partial x_K} = -\frac{\frac{\partial f}{\partial x_K}}{\frac{\partial f}{\partial x_L}} = -\frac{x_K}{x_L}$$

This is a funny example because the slope is increasing (MRTS is usually decreasing). This is because $V(y)$ is not convex.

Theorem: If γ is convex, then $v(y)$ is also convex.

$$(P \Rightarrow Q \Leftrightarrow \text{not } Q \Rightarrow \text{not } P)$$

Proof

$$(-x, y) \in \gamma$$

$$(-x', y) \in \gamma$$

$$\gamma \text{ convex} \Rightarrow t(-x, y) + (1-t)(-x', y) \in \gamma$$

$\underbrace{_{\text{linear combination.}}}_{\in \gamma}$

$$\Rightarrow (-tx - (1-t)x', ty + (1-t)y) \in \gamma$$

$$\Rightarrow (-tx - (1-t)x', y) \in \gamma$$

$$\Rightarrow (-x'', y) \in \gamma \text{ where } x'' = tx + (1-t)x'$$

$$\Rightarrow x'' \in v(y)$$

$$\Rightarrow v(y) \text{ is } \underline{\text{convex.}}$$

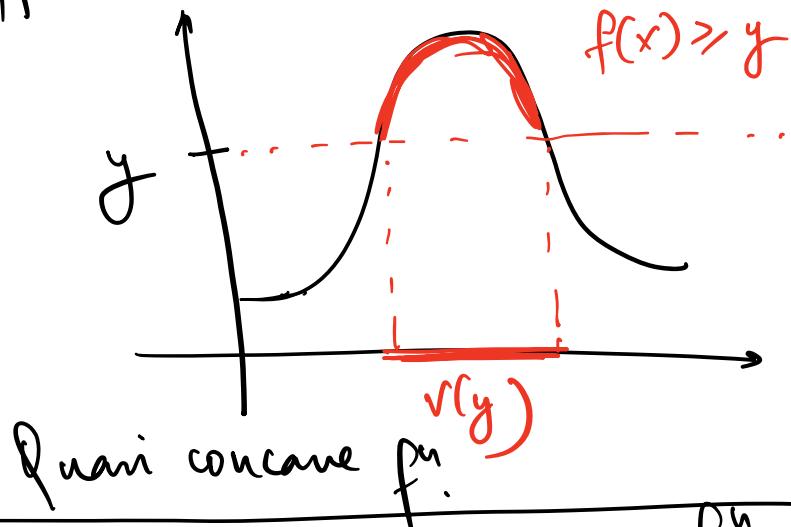
Theorem : $V(y)$ is convex iff $f(x)$ is quasiconcave.

Proof : $V(y) = \{x \in \mathbb{R}^n : f(x) \geq y\}$

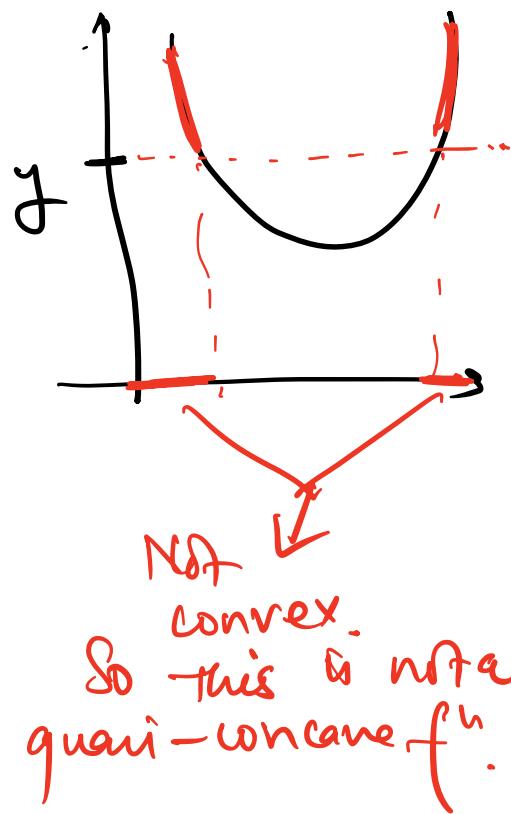
(It's more of a defⁿ)

↑ upper contour set of f .

Definition of a quasi concave f^n that has convex upper contour sets.



Concave f^n are quasi-concave f^n .



Homogeneity + Homotheticity

A function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is homogeneous of degree k

if $f(tx) = t^k f(x)$ & $t \neq 0$.

$$k=0 : f(tx) = f(x)$$

$$k=1 : f(tx) = tf(x)$$

Cobb-Douglas :

$$f(x_k, x_L) = x_k^\alpha x_L^\beta \quad \alpha + \beta = 1.$$

$$\begin{aligned}
 f(tx_k, tx_L) &= (tx_k)^\alpha (tx_L)^\beta \\
 &= t^{\alpha+\beta} x_k^\alpha x_L^\beta \\
 &= t x_k^\alpha x_L^\beta \quad \xrightarrow{\text{when } \alpha + \beta = 1.} = tf(x)
 \end{aligned}$$

What about MRTS?

$$\begin{aligned}
 \text{MRTS} &= \frac{\alpha}{\beta} \frac{x_L}{x_K} \\
 &= \frac{\alpha}{\beta} \frac{(tx_L)}{(tx_K)} = \frac{\alpha}{\beta} \frac{x_L}{x_K}
 \end{aligned}$$

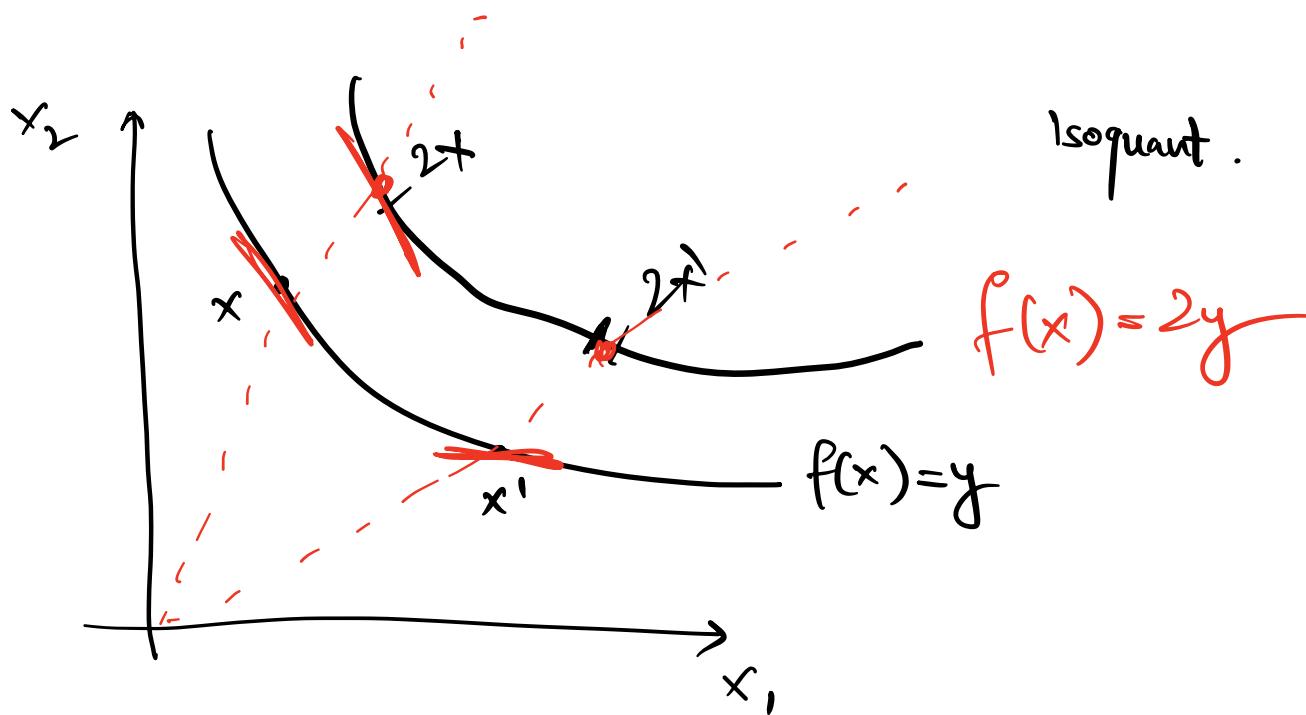
MRTS is homogeneous of degree 0.

Theorem: If f is $hd-k$, then the partial derivative f_{x_i} is $hd-(k-1)$.

Proof: $f(tx) = t^k f(x)$

$$t f_{x_i}(tx) = t^k f_{x_i}(x)$$

$$f_{x_i}(tx) = t^{k-1} f_{x_i}(x)$$



Say f is $hd-1$

$$f(2x) = 2f(x) = 2y$$

The slopes would be the same too -

Every isoquant is just a blown up version of the first one.