

A.5 Sufficient Conditions for a Maximum (pp. 513-520)

Def: f is concave on $[a, b]$ if

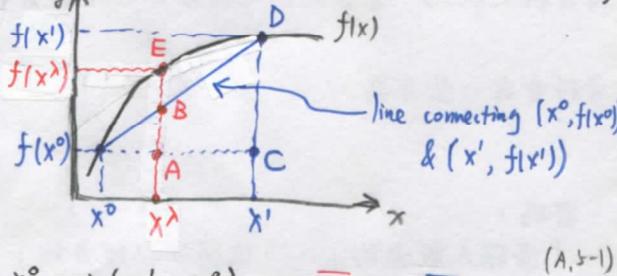
$$\text{For } x^*, x' \in [a, b], \quad x^\lambda = (1-\lambda)x^* + \lambda x', \quad 0 < \lambda < 1 \\ f(x^\lambda) \geq (1-\lambda)f(x^*) + \lambda f(x')$$

f is strictly concave if

$$f(x^\lambda) > (1-\lambda)f(x^*) + \lambda f(x').$$

Note: This means that function f between x^* & x'

lies above the line connecting $(x^*, f(x^*))$ & $(x', f(x'))$.



(A.5-1)

$$\text{Why? } x^\lambda - x^* = \lambda(x' - x^*) \Rightarrow \overline{AB} = \lambda \cdot \overline{CD}. \quad (\text{linear})$$

For f concave, we have

$$\frac{\overline{AB}}{\overline{AE}} \geq \frac{\overline{CD}}{\overline{CD}} \quad \text{#} \quad \text{#}$$

$$\frac{f(x^\lambda) - f(x^*)}{\overline{f(x^*) - f(x^*)}} \geq \frac{\overline{f(x^*) - f(x^*)}}{\overline{f(x^*) - f(x^*)}} = 1 \quad \text{#}$$

(pt) By Prop. A.5-1, $\frac{df}{dx}(x^*) \cdot (x^\lambda - x^*) \geq f(x^\lambda) - f(x^*)$
 f strictly concave $\Rightarrow f(x^\lambda) - f(x^*) > \lambda \cdot [f(x') - f(x^*)]$
 $\text{Hence, } \frac{df}{dx}(x^*) \cdot (x^\lambda - x^*) > \lambda \cdot [f(x') - f(x^*)] \\ = \left(\frac{x^\lambda - x^*}{x' - x^*} \right) \cdot [f(x') - f(x^*)] \\ \therefore \frac{df}{dx}(x^*) \cdot (x' - x^*) > f(x') - f(x^*).$

Moreover, if f is diff. at $x' > x^*$, by Prop. A.5-1,

$$\frac{df}{dx}(x^*) \cdot (x' - x^*) > f(x') - f(x^*) \geq \frac{df}{dx}(x') \cdot (x' - x^*).$$

Note: This means that all tangent lines lie above the graph of f .

In fact, this is "iff":

Prop. A.5-3: A diff. function f is concave on interval X iff

$$\text{for } x^*, x' \in X, \quad f(x') \leq f(x^*) + \frac{df}{dx}(x^*) \cdot (x' - x^*). \quad (\text{A.5-3})$$

(pt) \Rightarrow By Prop. A.5-2.

\Leftrightarrow For $x^*, x' \in X, \quad x^\lambda = (1-\lambda)x^* + \lambda \cdot x'$, by (A.5-3):

$$\begin{cases} f(x^*) \leq f(x^\lambda) + \frac{df}{dx}(x^\lambda) \cdot (x^* - x^\lambda) \dots \textcircled{1} \\ f(x') \leq f(x^\lambda) + \frac{df}{dx}(x^\lambda) \cdot (x' - x^\lambda) \dots \textcircled{2} \end{cases}$$

$$\textcircled{1} \cdot (1-\lambda) + \textcircled{2} \cdot \lambda: \quad (1-\lambda)f(x^*) + \lambda \cdot f(x') \leq f(x^\lambda) + 0. \quad \text{#}$$

Note: This also means the $\frac{df}{dx} \downarrow$ as $x \uparrow$, or $\frac{d^2f}{dx^2} \leq 0$.

In fact, this is also "iff":

Prop. A.5-4: A twice diff. function f is concave on the interval X

iff for $x^*, x' \in X, \quad \frac{d^2f}{dx^2}(x) \leq 0 \quad \forall x \in X$.

(pt) \Rightarrow By Prop. A.5-2.

\Leftrightarrow Consider $g(\lambda) = f(x^* + \lambda(x' - x^*))$, by chain rule,

$$\frac{dg}{d\lambda}(\lambda) = (x' - x^*) \cdot \frac{df}{dx}(x^*), \quad \frac{d^2g}{d\lambda^2}(\lambda) = (x' - x^*)^2 \cdot \frac{d^2f}{dx^2}(x^*).$$

Since $\frac{d^2f}{dx^2} \leq 0 \quad \forall x \in X, \quad \frac{d^2g}{d\lambda^2} \leq 0 \quad \& \quad \frac{dg}{d\lambda}$ decreasing on $[0, 1]$

$$\therefore \frac{dg}{d\lambda}(0) = (x' - x^*) \cdot \frac{df}{dx}(x^*) = \int_0^1 \frac{dg}{d\lambda}(0) d\lambda \\ \leq \int_0^1 \frac{dg}{d\lambda}(\lambda) \cdot d\lambda = g(1) - g(0) = f(x') - f(x^*)$$

So by Prop. A.5-3, f is concave. #

Prop. A.5-5 (Sufficient Conditions for a Maximum)

Suppose f is concave on $[a, b]$ & for some $x^* \in (a, b)$, $\frac{df}{dx}(x^*) = 0$. Then, $f(x) \leq f(x^*) \quad \forall x \in [a, b]$.

(pt) By Prop. A.5-3, $\forall x \in [a, b], \quad f(x) \leq f(x^*) + \frac{df}{dx}(x^*) \cdot (x - x^*) = f(x^*)$
 since $\frac{df}{dx}(x^*) = 0$. #

Note: FOC is enough (if f is concave).

Corollary A.5-2: If f is strictly concave on interval X &

diff. at $x^* \in X$, then $\frac{df}{dx}(x^*) \cdot (x' - x^*) > f(x') - f(x^*)$ (pt)

Moreover, if f is also diff. at $x' > x^*$, then $\frac{df}{dx}(x^*) > \frac{df}{dx}(x')$.

A.5 Sufficient Conditions for a Maximum (Continued)

Concavity might be too strong. Can relax to quasi-concavity.

Def: (Local Property) f satisfies property P locally at x^* if property P holds in some neighborhood of x^* .

Ex: (Local Maximum) By Corollary A.5-2, if $\frac{df}{dx}(x^*) = 0$ (FOC holds at x^*) & f is locally strictly concave at x^* , then $f(x^*) > f(x')$ $\forall x'$ in a neighborhood of x^* . i.e. f has a local maximum at x^* .

Def: f is quasi-concave over $[a, b]$ if for $x^*, x' \in [a, b]$, $x^\lambda = (1-\lambda)x^* + \lambda x'$, $0 < \lambda < 1$, $f(x^\lambda) \geq \min\{f(x^*), f(x')\}$.

Note: Quasi-concavity implies that a local maximum is also a global maximum.

Prop. A.5-6 (Quasi-concavity & Sufficient Conditions for a Max.)

If $\frac{df}{dx}(x^*) = 0$ & f is locally strictly concave at $x^* \in [a, b]$, then f has a local maximum at x^* .

If f is quasi-concave on $[a, b]$, then $f(x) \leq f(x^*) \forall x \in [a, b]$.

(Pf) Ex. gives the first statement: \exists neighborhood $N(x^*, \delta)$ s.t. $f(x) < f(x^*) \quad \forall x \in N(x^*)$.

Consider $[a, b]$, if there exists $x' \in [a, b]$ s.t. $f(x') > f(x^*)$, then if f is quasi-concave, $f(x^\lambda) \geq \min\{f(x'), f(x^*)\} = f(x^*)$ for all $x^\lambda = (1-\lambda)x^* + \lambda x'$, $0 < \lambda < 1$. But this means that $\exists x^\lambda \in N(x^*, \delta)$ s.t. $f(x^\lambda) \geq f(x^*)$, contradiction. $\#$

By setting $g = -f$, we get the flip-side conditions for minimum:

Def: f is convex on $[a, b]$ if for $x^*, x' \in [a, b]$, $x^\lambda = (1-\lambda)x^* + \lambda x'$, $0 < \lambda < 1$, we have $f(x^\lambda) \leq (1-\lambda)f(x^*) + \lambda \cdot f(x')$

(i.e. $g = -f$ is concave!)

f is strictly convex if

$$f(x^\lambda) < (1-\lambda)f(x^*) + \lambda \cdot f(x')$$

Prop. A.5-7 (Sufficient Conditions for a Minimum)

Suppose f is convex on $D_f = \mathbb{R}$ & for some x^* , $\frac{df}{dx}(x^*) = 0$.

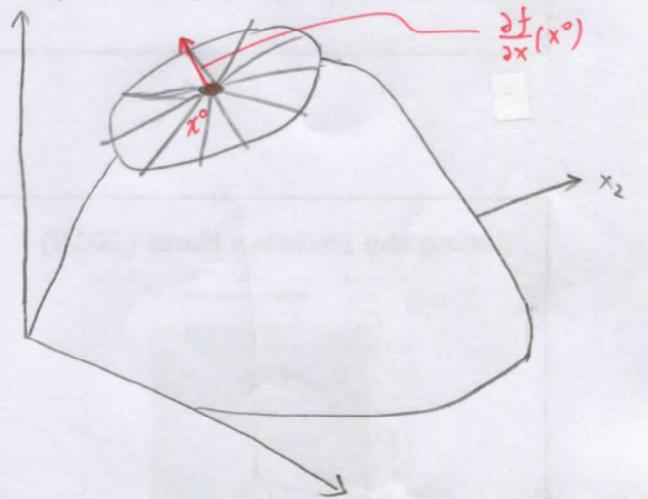
Then, $f(x) \geq f(x^*) \quad \forall x \in \mathbb{R}$.

(Pf) Set $g = -f$ & use Prop. A.5-5. $\#$

B.2 Concave Functions of Vectors (pp. 532-)

Def: f is concave on convex set $X \subseteq \mathbb{R}^n$ if for $x^*, x' \in X$, $x^\lambda = (1-\lambda)x^* + \lambda x'$, $0 < \lambda < 1$ $f(x^\lambda) \geq (1-\lambda)f(x^*) + \lambda f(x')$ (x^*, x' , x^λ are all vectors in \mathbb{R}^n)
 f is strictly concave if $f(x^\lambda) > (1-\lambda)f(x^*) + \lambda f(x')$

Def: $H = \left\{ x \mid \frac{\partial f}{\partial x}(x^*) \cdot (x - x^*) = 0 \right\}$ is the tangent hyperplane of f at x^* .



Prop. B.2-3 (All tangent hyperplanes lie above the graph of f)

A diff. function f is concave on convex set $X \subseteq \mathbb{R}^n$ iff for $x^*, x' \in X$, $f(x') \leq f(x^*) + \frac{\partial f}{\partial x}(x^*) \cdot (x' - x^*)$.

(Pf) \Leftrightarrow (Copy Prop. A.5-3)

$$f(x^*) \leq f(x^\lambda) + \frac{\partial f}{\partial x}(x^\lambda) \cdot (x^\lambda - x^*) \quad \dots \textcircled{1}$$

$$f(x') \leq f(x^\lambda) + \frac{\partial f}{\partial x}(x^\lambda) \cdot (x' - x^\lambda) \quad \dots \textcircled{2}$$

$$\textcircled{1} \cdot (1-\lambda) + \textcircled{2} \cdot \lambda : (1-\lambda)f(x^*) + \lambda f(x') \leq f(x^\lambda) + 0. \quad \#$$

(\Rightarrow) Consider $g(\lambda) = f(x^\lambda) = f(x^* + \lambda(x' - x^*)) \geq (1-\lambda)f(x^*) + \lambda f(x')$

$$g'(\lambda) = \frac{\partial f}{\partial x}(x^\lambda) \cdot (x' - x^*) \quad (\text{f concave})$$

Since $g(0) = f(x^*)$, $g(\lambda) - g(0) \geq \lambda \cdot [f(x') - f(x^*)]$

$$\text{Or, } \frac{g(\lambda) - g(0)}{\lambda - 0} \geq f(x') - f(x^*)$$

$$\underset{(\lambda \rightarrow 0)}{\underbrace{g'(0)}} = \frac{\partial f}{\partial x}(x^*) \cdot (x' - x^*) \quad \#$$

Prop. B.2-4

A twice diff. function f is concave on the convex set $X \subseteq \mathbb{R}^n$

iff for $x^*, x \in X$, the quadratic form $g(x)$ is negative semi-definite.

$$g(x) = \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x^*) (x_i - x^*)(x_j - x^*)$$

Ex: In \mathbb{R}^2 , this requires $\frac{\partial^2 f}{\partial x_i^2} \leq 0, j=1,2$ & $\frac{\partial^2 f}{\partial x_1 \partial x_2} - \left(\frac{\partial^2 f}{\partial x_1 \partial x_2}\right)^2 \geq 0$

B.2 Concave Functions of Vectors (Continued)

(pf of Prop. B.2-4)

For $x^0, x^1 = x^0 + \vec{z}$, define $x^\theta = x^0 + \theta \vec{z}$ & $g(\theta) = f(x^0 + \theta \vec{z})$.

$\Rightarrow f$ is concave $\Rightarrow g(\theta)$ is concave.

(pf) Consider θ^0, θ^1 & $\theta^\lambda = (1-\lambda)\theta^0 + \lambda\theta^1$,

$$\begin{aligned} g(\theta^\lambda) &= f(x^0 + (1-\lambda)\theta^0 \vec{z} + \lambda\theta^1 \vec{z}) = f((1-\lambda)(x^0 + \theta^0 \vec{z}) + \lambda(x^1 + \theta^1 \vec{z})) \\ &\geq (1-\lambda) f(x^0 + \theta^0 \vec{z}) + \lambda f(x^1 + \theta^1 \vec{z}) \quad (\because f \text{ is concave}) \\ &= (1-\lambda) g(\theta^0) + \lambda g(\theta^1). \end{aligned}$$

g is concave $\Rightarrow g''(0) \leq 0$

$$\text{But } g'(0) = \frac{\partial f}{\partial x}(x^0 + \theta^0 \vec{z}) \cdot \vec{z} = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x^0 + \theta^0 \vec{z}) \cdot \vec{z}_i$$

$$\Rightarrow g''(0) = \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x^0 + \theta^0 \vec{z}) \cdot \vec{z}_i \cdot \vec{z}_j$$

$$\text{Hence, } g''(0) = \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x^0) \cdot (x_i - x_i^0) \cdot (x_j - x_j^0) \leq 0. \quad *$$

$(\Leftarrow) g'(x) \leq 0 \Rightarrow g''(0) \leq 0 \Rightarrow g'(0)$ decreasing on $[0, 1]$
 $\Rightarrow g'(0) \leq g'(1) \quad \forall \theta \in [0, 1]$

$$\begin{aligned} \therefore f(x') - f(x^0) &= g(1) - g(0) = \int_0^1 g'(1) d\theta \leq \int_0^1 g'(0) d\theta \\ &= g'(0) = \frac{\partial f}{\partial x}(x^0) \cdot (x' - x^0). \end{aligned}$$

Hence, by Prop. B.2-3, f is concave. $(x, p \geq 0)$

Exercise: Show that the Cobb-Douglas function $f(x) = x_1^\alpha x_2^\beta, x \geq 0$ is concave if $\alpha + \beta \leq 1$.

Prop. B.2-5 (Sum of Concave Functions)

f, g concave $\Rightarrow (f+g)$ is concave. EX: $f(x) = \sum_j b_n x_j$ is concave

Prop. B.2-6 (Concave Function of a Function)

$h(x) = g(f(x))$ is concave if g is concave and either

(a) g is increasing & f is concave, or (b) f is linear.

EX: $f(x) = (x_1^\alpha + x_2^\beta)^r, \alpha, \beta, r \in (0, 1], x \geq 0$ is concave.

Def: f is quasi-concave on convex set X if

$x^0, x^1 \in X, x^\lambda = (1-\lambda)x^0 + \lambda x^1, 0 < \lambda < 1,$

$f(x^\lambda) \geq \min\{f(x^0), f(x^1)\}.$

f is quasi-convex if $f(x^\lambda) \leq \min\{f(x^0), f(x^1)\}$.

Prop. B.2-7 f is concave $\Rightarrow f$ is quasi-concave.

Prop. B.2-8 f is concave, g is increasing $\Rightarrow h(x) = g(f(x))$ is

EX: $f(x) = x_1^{\alpha_1} x_2^{\alpha_2} \cdots x_n^{\alpha_n}, x \geq 0$ is quasi-concave!

Def: f is homogeneous of degree 1 if $f(mx) = m f(x) \quad \forall x \in \mathbb{R}_+^n, m > 0$

Prop. B.2-9 f is quasi-concave & homogeneous of degree 1 $\Rightarrow f$ concave.

(pf) For x^0, x^1 , find θ s.t. $\frac{f((1-\lambda)x^0)}{f(x^0)} = \theta \frac{f(\lambda x^1)}{f(x^1)}$

$$\Rightarrow (1-\lambda)f(x^0) + \lambda f(x^1) = (1+\theta) f(\lambda x^1)$$

$$\text{Consider } \frac{\theta}{1+\theta} \frac{(1-\lambda)x^0}{f(x^0)} + \frac{1}{1+\theta} \frac{\theta \lambda x^1}{f(x^1)} = \frac{\theta}{1+\theta} \cdot [(1-\lambda)x^0 + \lambda x^1],$$

f is quasi-concave \Rightarrow

$$f\left(\frac{\theta}{1+\theta} \cdot [(1-\lambda)x^0 + \lambda x^1]\right) \geq \min\{f((1-\lambda)x^0), f(\lambda x^1)\} = f(\theta x^1)$$

$$\Rightarrow \frac{\theta}{1+\theta} f((1-\lambda)x^0 + \lambda x^1) \geq \theta f(\lambda x^1)$$

$$\Rightarrow f((1-\lambda)x^0 + \lambda x^1) \geq (1+\theta) f(\lambda x^1) = (1-\lambda)f(x^0) + \lambda f(x^1). \quad *$$

Ex: $f(x) = x_1^{\alpha_1} x_2^{\alpha_2} \cdots x_n^{\alpha_n}, x \geq 0, \alpha_1 + \cdots + \alpha_n = 1 \Rightarrow$ homogeneous of degree 1.

$\Rightarrow f$ is concave (since f is quasi-concave).

Def: $C(x^0) = \{x \mid f(x) = f(x^0)\}$ is the contour set of f .

$C_U(x^0) = \{x \mid f(x) \geq f(x^0)\}$ is the upper contour set.

Prop. B.2-11 f is quasi-concave iff the upper contour set of f is convex.

(pf) \Rightarrow f is quasi-concave. For any \hat{x} & $x^0, x^1 \in C_U(\hat{x})$,

$$f(x^0) \geq f(\hat{x}), f(x^1) \geq f(\hat{x}) \text{ for } x^\lambda = (1-\lambda)x^0 + \lambda x^1,$$

$$f(x^\lambda) \geq \min\{f(x^0), f(x^1)\} \geq f(\hat{x}),$$

$$\text{i.e. } x^\lambda \in C_U(\hat{x}). \quad *$$

\Leftarrow For any x^0, x^1 , Without loss of generality, assume $f(x^1) \geq f(x^0)$

$\therefore x^1 \in C_U(x^0)$. By convex upper contour set,

$$x^\lambda = (1-\lambda)x^0 + \lambda x^1 \in C_U(x^1)$$

$$\Rightarrow f(x^\lambda) \geq f(x^0) = \min\{f(x^0), f(x^1)\} \Rightarrow f \text{ quasi-concave}$$

Prop. B.2-12 $f \in C^1$ is quasi-concave iff

$$f(x) \geq f(x^0) \Rightarrow \frac{\partial f}{\partial x}(x^0) \cdot (x - x^0) \geq 0$$