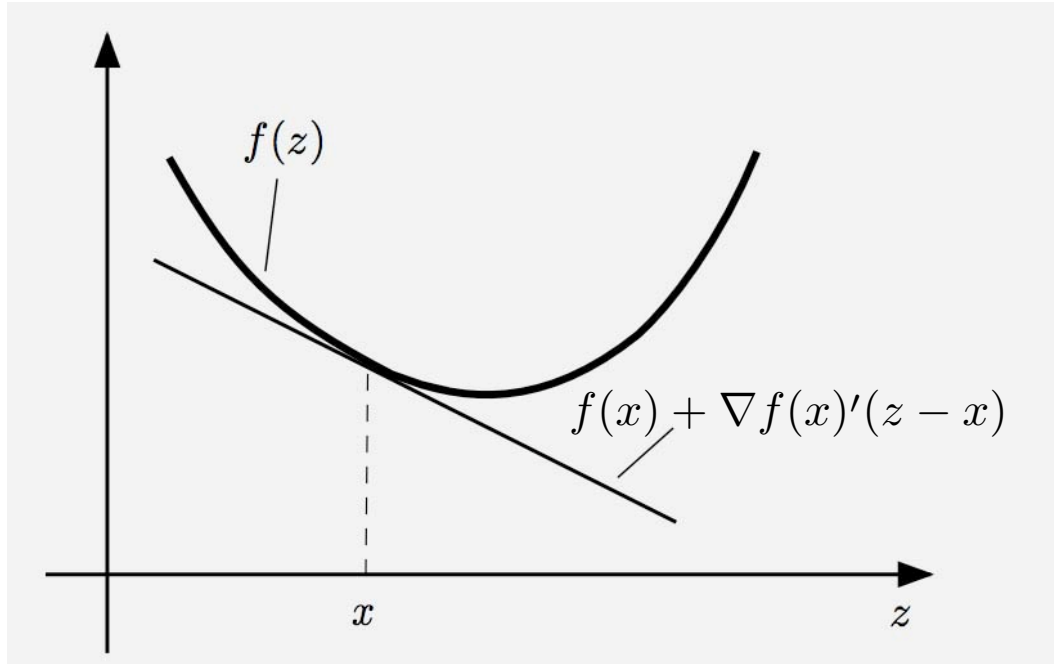


# LECTURE 3

## LECTURE OUTLINE

- Differentiable Convex Functions
- Convex and Affine Hulls
- Caratheodory's Theorem
- Relative Interior

# DIFFERENTIABLE CONVEX FUNCTIONS



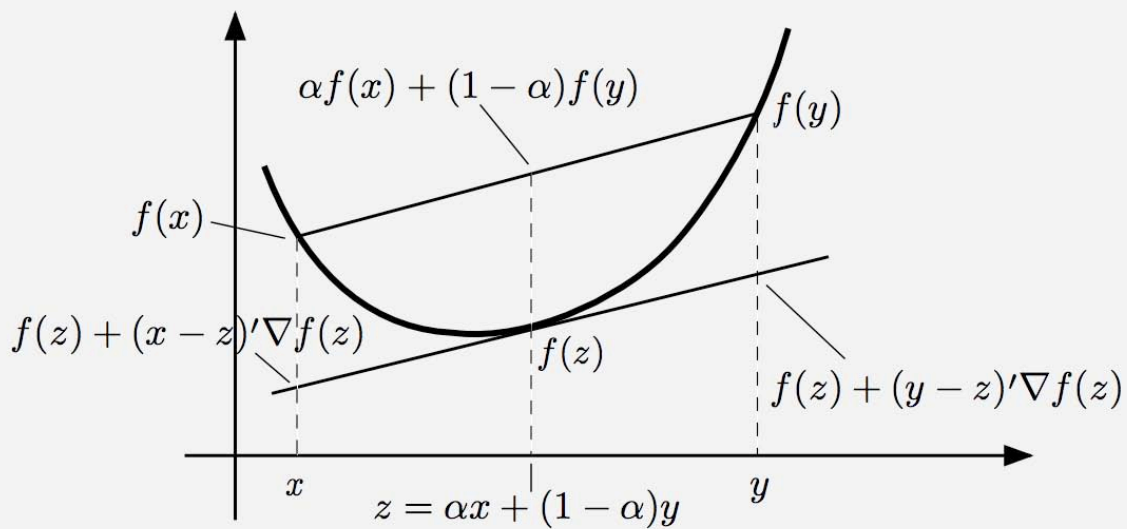
• Let  $C \subset \mathfrak{R}^n$  be a convex set and let  $f : \mathfrak{R}^n \mapsto \mathfrak{R}$  be differentiable over  $\mathfrak{R}^n$ .

(a) The function  $f$  is convex over  $C$  iff

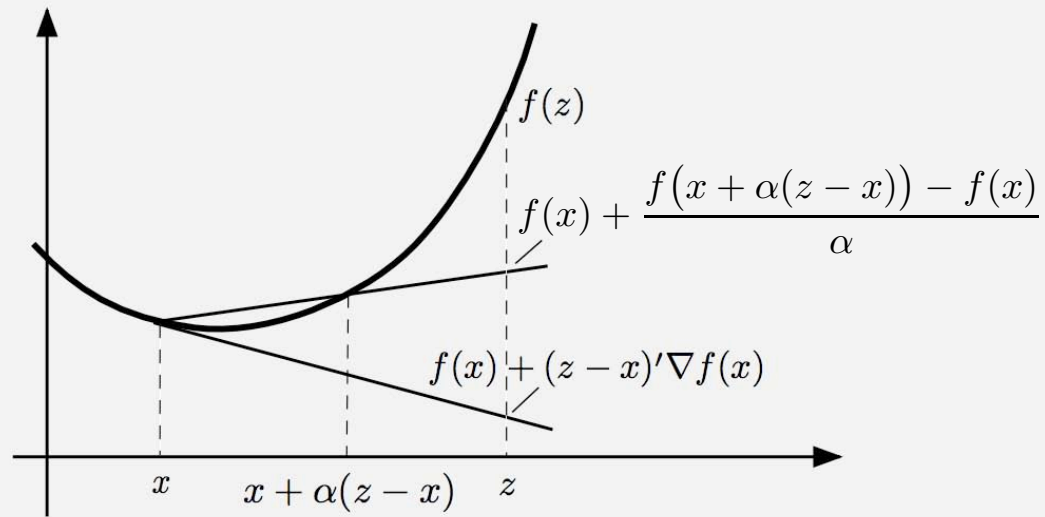
$$f(z) \geq f(x) + (z - x)' \nabla f(x), \quad \forall x, z \in C$$

(b) If the inequality is strict whenever  $x \neq z$ , then  $f$  is strictly convex over  $C$ .

# PROOF IDEAS



(a)



(b)

## OPTIMALITY CONDITION

• Let  $C$  be a nonempty convex subset of  $\mathfrak{R}^n$  and let  $f : \mathfrak{R}^n \mapsto \mathfrak{R}$  be convex and differentiable over an open set that contains  $C$ . Then a vector  $x^* \in C$  minimizes  $f$  over  $C$  if and only if

$$\nabla f(x^*)'(z - x^*) \geq 0, \quad \forall z \in C.$$

**Proof:** If the condition holds, then

$$f(z) \geq f(x^*) + (z - x^*)' \nabla f(x^*) \geq f(x^*), \quad \forall z \in C,$$

so  $x^*$  minimizes  $f$  over  $C$ .

Converse: Assume the contrary, i.e.,  $x^*$  minimizes  $f$  over  $C$  and  $\nabla f(x^*)'(z - x^*) < 0$  for some  $z \in C$ . By differentiation, we have

$$\lim_{\alpha \downarrow 0} \frac{f(x^* + \alpha(z - x^*)) - f(x^*)}{\alpha} = \nabla f(x^*)'(z - x^*) < 0$$

so  $f(x^* + \alpha(z - x^*))$  decreases strictly for sufficiently small  $\alpha > 0$ , contradicting the optimality of  $x^*$ . **Q.E.D.**

# TWICE DIFFERENTIABLE CONVEX FNS

- Let  $C$  be a convex subset of  $\mathbb{R}^n$  and let  $f : \mathbb{R}^n \mapsto \mathbb{R}$  be twice continuously differentiable over  $\mathbb{R}^n$ .
  - (a) If  $\nabla^2 f(x)$  is positive semidefinite for all  $x \in C$ , then  $f$  is convex over  $C$ .
  - (b) If  $\nabla^2 f(x)$  is positive definite for all  $x \in C$ , then  $f$  is strictly convex over  $C$ .
  - (c) If  $C$  is open and  $f$  is convex over  $C$ , then  $\nabla^2 f(x)$  is positive semidefinite for all  $x \in C$ .

**Proof:** (a) By mean value theorem, for  $x, y \in C$

$$f(y) = f(x) + (y-x)' \nabla f(x) + \frac{1}{2} (y-x)' \nabla^2 f(x + \alpha(y-x)) (y-x)$$

for some  $\alpha \in [0, 1]$ . Using the positive semidefiniteness of  $\nabla^2 f$ , we obtain

$$f(y) \geq f(x) + (y-x)' \nabla f(x), \quad \forall x, y \in C$$

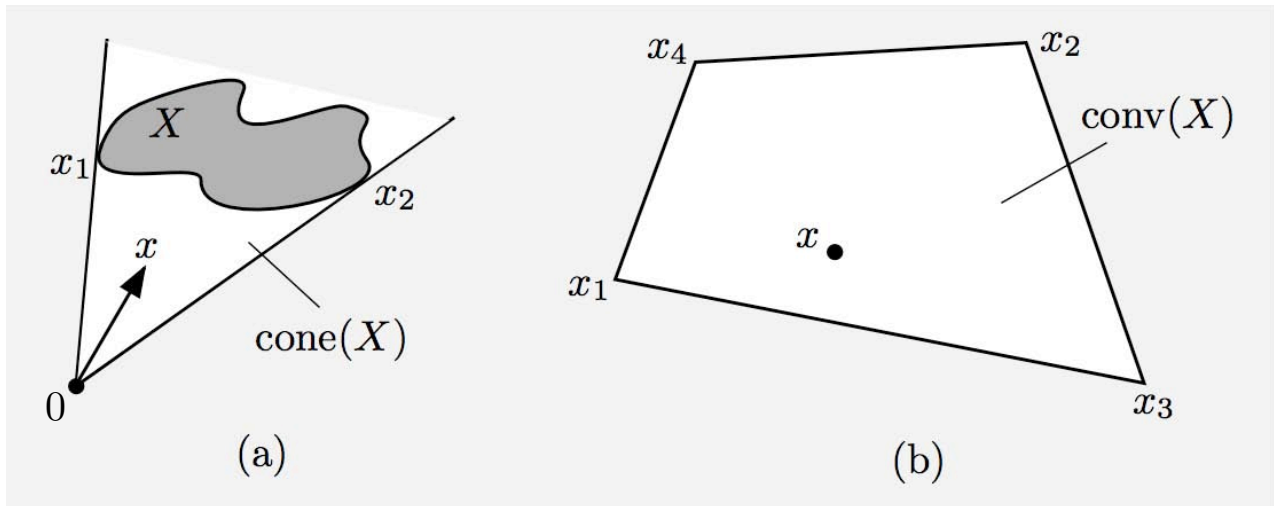
From the preceding result,  $f$  is convex.

- (b) Similar to (a), we have  $f(y) > f(x) + (y-x)' \nabla f(x)$  for all  $x, y \in C$  with  $x \neq y$ , and we use the preceding result.
- (c) By contradiction ... similar.

# CONVEX AND AFFINE HULLS

- Given a set  $X \subset \mathbb{R}^n$ :
- A *convex combination* of elements of  $X$  is a vector of the form  $\sum_{i=1}^m \alpha_i x_i$ , where  $x_i \in X$ ,  $\alpha_i \geq 0$ , and  $\sum_{i=1}^m \alpha_i = 1$ .
- The *convex hull* of  $X$ , denoted  $\text{conv}(X)$ , is the intersection of all convex sets containing  $X$ . (Can be shown to be equal to the set of all convex combinations from  $X$ ).
- The *affine hull* of  $X$ , denoted  $\text{aff}(X)$ , is the intersection of all affine sets containing  $X$  (an affine set is a set of the form  $x + S$ , where  $S$  is a subspace).
- A *nonnegative combination* of elements of  $X$  is a vector of the form  $\sum_{i=1}^m \alpha_i x_i$ , where  $x_i \in X$  and  $\alpha_i \geq 0$  for all  $i$ .
- The *cone generated by*  $X$ , denoted  $\text{cone}(X)$ , is the set of all nonnegative combinations from  $X$ :
  - It is a convex cone containing the origin.
  - It need not be closed!
  - If  $X$  is a finite set,  $\text{cone}(X)$  is closed (non-trivial to show!)

# CARATHEODORY'S THEOREM



- Let  $X$  be a nonempty subset of  $\mathbb{R}^n$ .
  - (a) Every  $x \neq 0$  in  $\text{cone}(X)$  can be represented as a positive combination of vectors  $x_1, \dots, x_m$  from  $X$  that are linearly independent (so  $m \leq n$ ).
  - (b) Every  $x \notin X$  that belongs to  $\text{conv}(X)$  can be represented as a convex combination of vectors  $x_1, \dots, x_m$  from  $X$  with  $m \leq n + 1$ .

# PROOF OF CARATHEODORY'S THEOREM

(a) Let  $x$  be a nonzero vector in  $\text{cone}(X)$ , and let  $m$  be the smallest integer such that  $x$  has the form  $\sum_{i=1}^m \alpha_i x_i$ , where  $\alpha_i > 0$  and  $x_i \in X$  for all  $i = 1, \dots, m$ . If the vectors  $x_i$  were linearly dependent, there would exist  $\lambda_1, \dots, \lambda_m$ , with

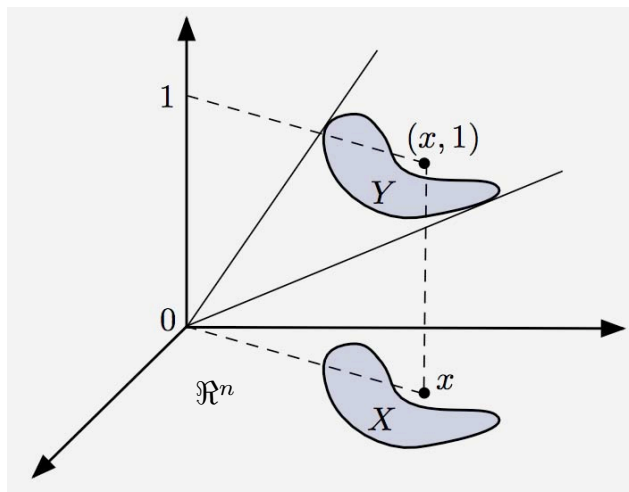
$$\sum_{i=1}^m \lambda_i x_i = 0$$

and at least one of the  $\lambda_i$  is positive. Consider

$$\sum_{i=1}^m (\alpha_i - \gamma \lambda_i) x_i,$$

where  $\gamma$  is the largest  $\gamma$  such that  $\alpha_i - \gamma \lambda_i \geq 0$  for all  $i$ . This combination provides a representation of  $x$  as a positive combination of fewer than  $m$  vectors of  $X$  – a contradiction. Therefore,  $x_1, \dots, x_m$ , are linearly independent.

(b) Use “lifting” argument: apply part (a) to  $Y = \{(x, 1) \mid x \in X\}$ .





# AN APPLICATION OF CARATHEODORY

- The convex hull of a compact set is compact.

**Proof:** Let  $X$  be compact. We take a sequence in  $\text{conv}(X)$  and show that it has a convergent subsequence whose limit is in  $\text{conv}(X)$ .

By Caratheodory, a sequence in  $\text{conv}(X)$  can be expressed as  $\left\{ \sum_{i=1}^{n+1} \alpha_i^k x_i^k \right\}$ , where for all  $k$  and  $i$ ,  $\alpha_i^k \geq 0$ ,  $x_i^k \in X$ , and  $\sum_{i=1}^{n+1} \alpha_i^k = 1$ . Since the sequence

$$\left\{ (\alpha_1^k, \dots, \alpha_{n+1}^k, x_1^k, \dots, x_{n+1}^k) \right\}$$

is bounded, it has a limit point

$$\left\{ (\alpha_1, \dots, \alpha_{n+1}, x_1, \dots, x_{n+1}) \right\},$$

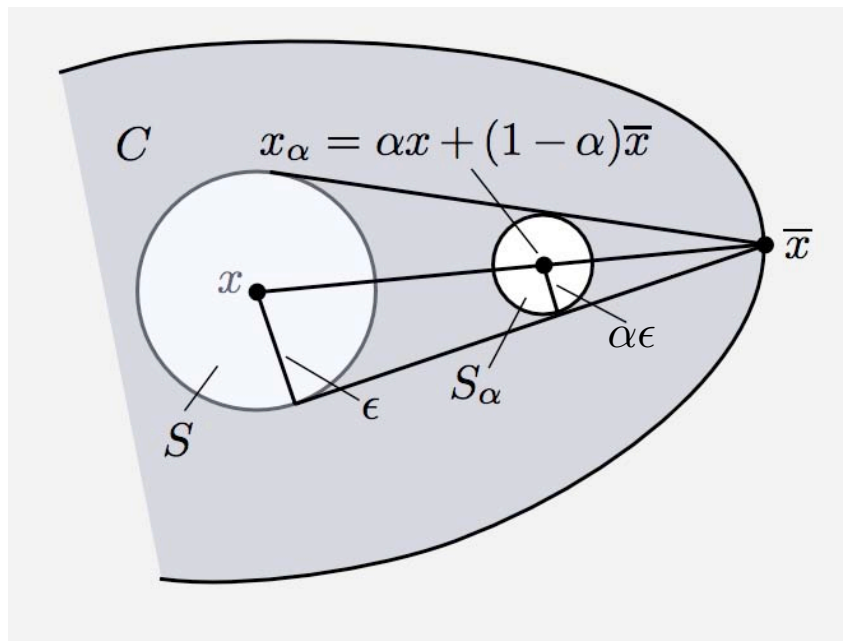
which must satisfy  $\sum_{i=1}^{n+1} \alpha_i = 1$ , and  $\alpha_i \geq 0$ ,  $x_i \in X$  for all  $i$ .

The vector  $\sum_{i=1}^{n+1} \alpha_i x_i$  belongs to  $\text{conv}(X)$  and is a limit point of  $\left\{ \sum_{i=1}^{n+1} \alpha_i^k x_i^k \right\}$ , showing that  $\text{conv}(X)$  is compact. **Q.E.D.**

- Note that the convex hull of a closed set need not be closed!

## RELATIVE INTERIOR

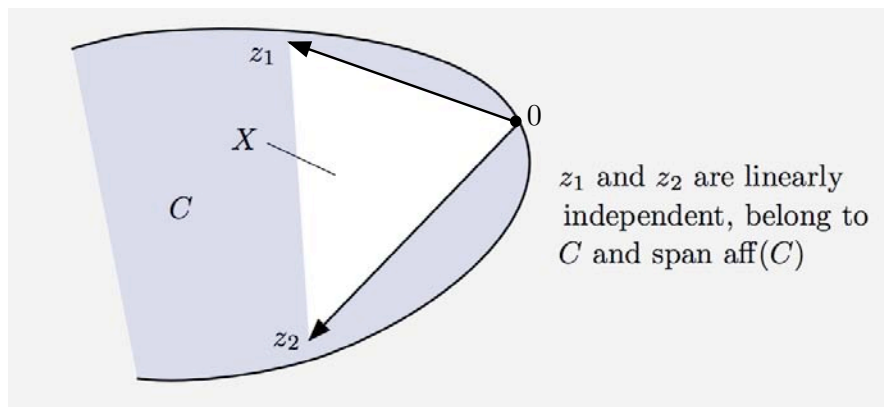
- $x$  is a *relative interior point* of  $C$ , if  $x$  is an interior point of  $C$  relative to  $\text{aff}(C)$ .
- $\text{ri}(C)$  denotes the *relative interior of  $C$* , i.e., the set of all relative interior points of  $C$ .
- **Line Segment Principle:** If  $C$  is a convex set,  $x \in \text{ri}(C)$  and  $\bar{x} \in \text{cl}(C)$ , then all points on the line segment connecting  $x$  and  $\bar{x}$ , except possibly  $\bar{x}$ , belong to  $\text{ri}(C)$ .



- Proof of case where  $x \in C$ : See the figure.
- Proof of case where  $x \notin C$ : Take sequence  $\{x_k\} \subset C$  with  $x_k \rightarrow x$ . Argue as in the figure.

## ADDITIONAL MAJOR RESULTS

- Let  $C$  be a nonempty convex set.
  - (a)  $\text{ri}(C)$  is a nonempty convex set, and has the same affine hull as  $C$ .
  - (b) **Prolongation Lemma:**  $x \in \text{ri}(C)$  if and only if every line segment in  $C$  having  $x$  as one endpoint can be prolonged beyond  $x$  without leaving  $C$ .



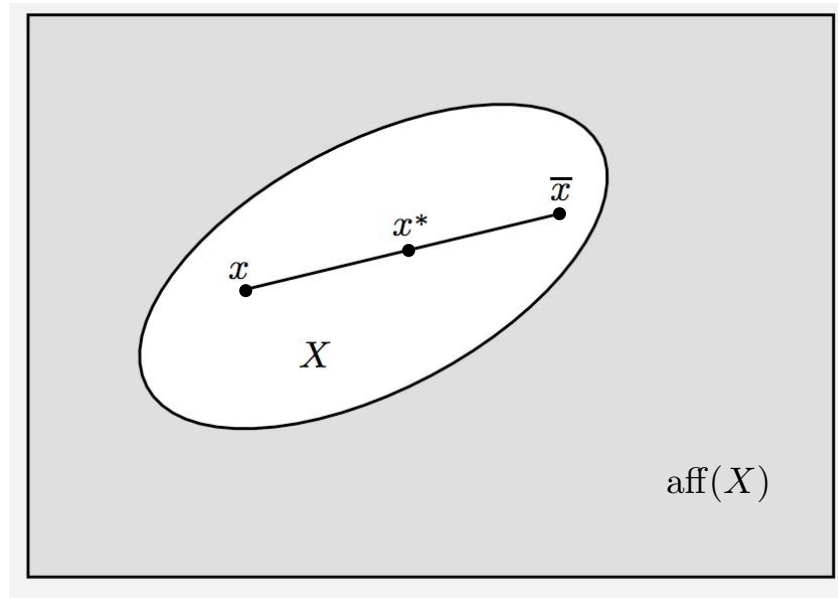
**Proof:** (a) Assume that  $0 \in C$ . We choose  $m$  linearly independent vectors  $z_1, \dots, z_m \in C$ , where  $m$  is the dimension of  $\text{aff}(C)$ , and we let

$$X = \left\{ \sum_{i=1}^m \alpha_i z_i \mid \sum_{i=1}^m \alpha_i < 1, \alpha_i > 0, i = 1, \dots, m \right\}$$

(b)  $\Rightarrow$  is clear by the def. of rel. interior. Reverse: take any  $x \in \text{ri}(C)$ ; use Line Segment Principle.

# OPTIMIZATION APPLICATION

- A concave function  $f : \mathfrak{R}^n \mapsto \mathfrak{R}$  that attains its minimum over a convex set  $X$  at an  $x^* \in \text{ri}(X)$  must be constant over  $X$ .



**Proof:** (By contradiction) Let  $x \in X$  be such that  $f(x) > f(x^*)$ . Prolong beyond  $x^*$  the line segment  $x$ -to- $x^*$  to a point  $\bar{x} \in X$ . By concavity of  $f$ , we have for some  $\alpha \in (0, 1)$

$$f(x^*) \geq \alpha f(x) + (1 - \alpha)f(\bar{x}),$$

and since  $f(x) > f(x^*)$ , we must have  $f(x^*) > f(\bar{x})$  - a contradiction. **Q.E.D.**

- **Corollary:** A linear function can attain a minimum only at the boundary of a convex set.

MIT OpenCourseWare  
<http://ocw.mit.edu>

6.253 Convex Analysis and Optimization  
Spring 2010

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.