

Let us study the dual, Slater's condition,  
strong duality, KKT conditions  
& application of all this through  
the example of LASSO

$$\min_{x \in \mathbb{R}^n} \|Ax - b\|_2^2 + \lambda \|x\|_1$$

where  $\lambda$  is fixed (not a Lagrange variable)

# Recap

For Lasso, it can be shown that for every  $\theta$  there exists a  $\lambda \geq 0$  s.t. following two problems are equivalent:

$$\min_x \|Ax - b\|_2^2 + \lambda \|x\|_1 \quad \textcircled{1} \dots \text{say soln is } x^* \text{ \& } \|x^*\|_1 = \beta$$

$$\begin{aligned} \min_x \|Ax - b\|_2^2 \\ \text{s.t. } \|x\|_1 \leq \theta \end{aligned}$$

$$\textcircled{2} \dots \text{say solution is } \hat{x}$$

Solution to  $\textcircled{2}$  with  $\theta = \beta = \|x^*\|_1$  is also  $x^*$ !

Solution to  $\textcircled{1}$  with  $\lambda$  as soln to  $A^T(b - Ax) = \lambda g_{\hat{x}}$  is also  $\hat{x}$ !

$$g_{\hat{x}} \in \partial \| \hat{x} \|_1$$

# What about dual of Lasso?

Dual of  $\min_{x \in \mathbb{R}^n} \frac{1}{2} \|y - Ax\|_2^2 + \lambda \|x\|_1$  is a constant!

Redefine primal as:

$$\min_{\substack{x \in \mathbb{R}^n \\ z \in \mathbb{R}^m}}$$
$$\frac{1}{2} \|y - z\|_2^2 + \lambda \|z\|_1$$

$$\text{s.t. } z = Ax$$

$$L^*(u) = \min_{\substack{z \in \mathbb{R}^n \\ z \in \mathbb{R}^m}} \frac{1}{2} \|y - z\|_2^2 + \lambda \|z\|_1 + u^T (z - Ax)$$
$$= \frac{1}{2} \|y\|_2^2 - \frac{1}{2} \|y - u\|_2^2 - \mathbb{I} \left( \left\| \frac{A^T u}{\lambda} \right\|_\infty \leq 1 \right)$$

$\therefore$  Lasso dual problem is

$$\max_{u \in \mathbb{R}^m} \frac{1}{2} (\|b\|_2^2 - \|b-u\|_2^2)$$

$$\text{s.t.} \quad \|A^T u\|_\infty \leq \lambda$$

$$\underline{\underline{\leq}} \quad \max_{u \in \mathbb{R}^m} \frac{1}{2} (\|u-b\|_2^2)$$
$$\text{s.t.} \quad \|A^T u\|_\infty \leq \lambda$$

Recall: Dual norm for  $p$  is  $q$  s.t.

$$\frac{1}{p} + \frac{1}{q} = 1$$

Here  $p=1$  &  $q=\infty$

∴ KKT conditions are necessary & sufficient  
(since Slater's condition satisfied)

① Primal constraints:  $\hat{z} = A\hat{x}$

② Dual constraints: NONE

③ Complementary slackness: NONE

④ Subdiff wrt primal variables = 0:

$$\partial_z L(\hat{x}, \hat{z}, \hat{\mu}) = \hat{z} - b - \hat{\mu} = 0$$

$$\partial_x L(\hat{x}, \hat{z}, \hat{\mu}) = \lambda \left[ \begin{array}{l} \text{sgn}(\hat{x}_i) \text{ if } \hat{x}_i \neq 0 \\ 0_i \in [-1, 1] \text{ if } \hat{x}_i = 0 \end{array} \right] + A^T \hat{\mu} = 0$$

Q: How to use the dual formulation & KKT?

Ans: We know by Slater's condition & KKT that if  $\hat{x}, \hat{\mu}, \hat{z}$  satisfy KKT then

$$\frac{1}{2} \|A\hat{x} - b\|_2^2 + \lambda \|\hat{x}\|_1 = \min_{\substack{x \in \mathbb{R}^n \\ z \in \mathbb{R}^m}} \frac{1}{2} \|z - b\|_2^2 + \lambda \|x\|_1$$

s.t.  $Ax = z$

$$= \max_{\mu \in \mathbb{R}^m} \frac{1}{2} \|\mu - b\|_2^2$$

s.t.  $\|A^T \mu\|_\infty \leq \lambda$

$$= \frac{1}{2} \|\hat{\mu} - b\|_2^2$$

Iterative algo for primal  
Claim: If  $\bar{a}_j = 2 \sum_{i=1}^n [A_{ij}]^2$  &  $\bar{b}_j = 2 \sum_{i=1}^n A_{ij}(t_i - x_j^* A_{ij})$

Then:

$$x_j^* = \begin{cases} (\bar{b}_j + \lambda) / \bar{a}_j & \text{if } \bar{b}_j < -\lambda \\ 0 & \text{if } \bar{b}_j \in [-\lambda, \lambda] \\ (\bar{b}_j - \lambda) / \bar{a}_j & \text{if } \bar{b}_j > \lambda \end{cases}$$

So to satisfy this, Lasso iterates on  $x^k$  as follows:  $x^{(0)} \rightarrow \bar{b}^{(0)} \rightarrow x^{(1)} \dots$  until convergence. We can understand through following simplification where  $A=I$

Iterative algo for primal ...

Stopping criterion:

$x^{(k)} \rightarrow$  solve for  $\mu^{(k)}$  &  $z^{(k)}$  using KKT

check gap  $\left. \begin{aligned} & \frac{1}{2} \|Ax^{(k)} - b\|_2^2 + \lambda \|x^{(k)}\|_1 \\ & - \frac{1}{2} \|\mu^{(k)} - b\|_2^2 \end{aligned} \right\} \leq \epsilon$

$$\min_{x \in \mathcal{D}} f(x)$$

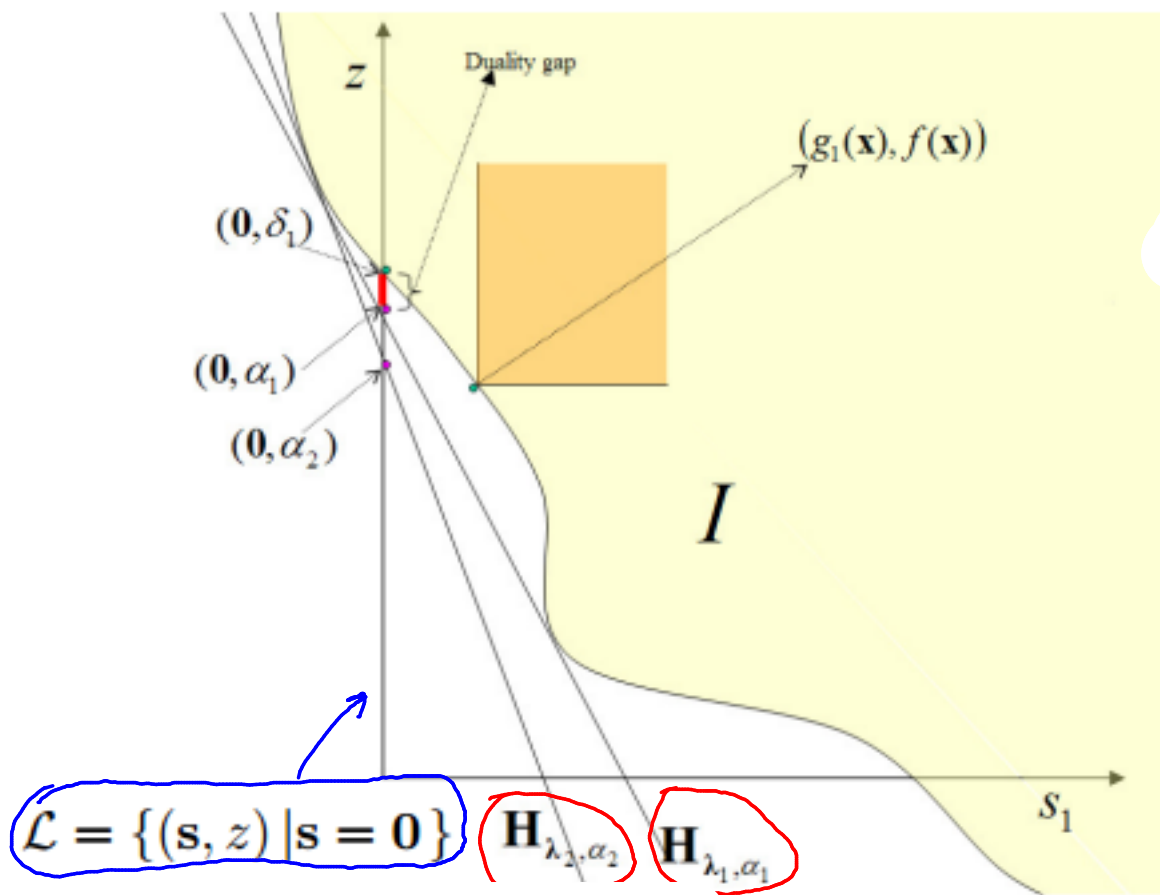
$$\text{s.t. } g_i(x) \leq 0 \quad i=1, \dots, m$$

(The general dual problem & its geometric interpretation)

pg 292, sec 4.4.3 of <http://www.cse.iitb.ac.in/~cs709/notes/BasicsOfConvexOptimization.pdf>

Consider the set:

$$\mathcal{I} = \{(s, z) \mid s \in \mathbb{R}^m, z \in \mathbb{R}, \exists x \in \mathcal{D} \text{ with } g_i(x) \leq s_i \forall 1 \leq i \leq m, f(x) \leq z\}$$



Kind of join between epigraphs of \$f\$ & \$g\_i\$'s.

$$\mathcal{H}_{\lambda, \alpha} = \{(s, z) \mid \lambda^T \cdot s + z = \alpha\}$$



$$\min_{x \in \mathcal{D}} f(x)$$

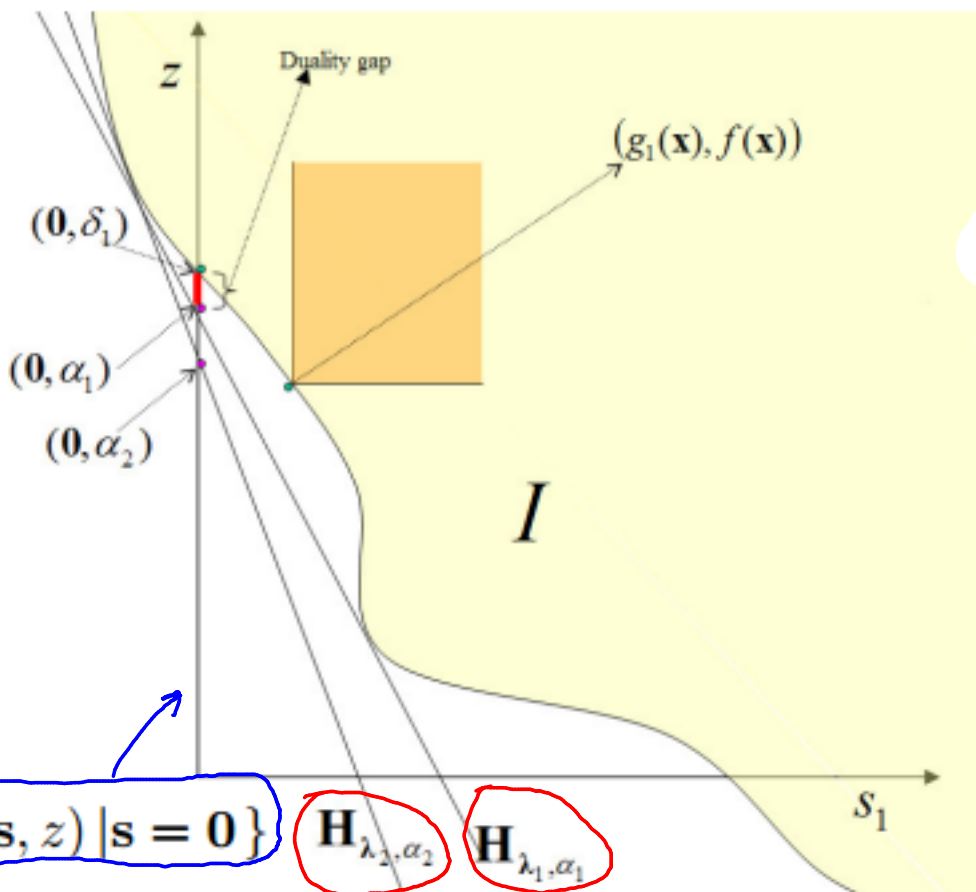
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Consider the set:

$$\mathcal{I} = \{(s, z) \mid s \in \mathbb{R}^m, z \in \mathbb{R}, \exists x \in \mathcal{D} \text{ with } g_i(x) \leq s_i \forall 1 \leq i \leq m, f(x) \leq z\}$$



Smallest  $z$  value in  $\mathcal{I}$  for  $s_i \leq 0$  will be for  $s_i = 0$

Since  $(s_1, z_1) \in \mathcal{I}$

$\Rightarrow (s_2, z_2) \in \mathcal{I}$

$\forall s_2 \geq s_1 \ \& \ z_2 \geq z_1$

$$\mathcal{L} = \{(s, z) \mid s = 0\}$$

$$\mathcal{H}_{\lambda_2, \alpha_2}$$

$$\mathcal{H}_{\lambda_1, \alpha_1}$$

$$\mathcal{H}_{\lambda, \alpha} = \{(s, z) \mid \lambda^T \cdot s + z = \alpha\}$$

Thus  $\Rightarrow$  is not possible!

$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & \mathcal{H}_{\lambda, \alpha}^+ \supseteq \mathcal{I} \end{aligned}$$

$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & \lambda^T \cdot \mathbf{s} + z \geq \alpha \quad \forall (\mathbf{s}, z) \in \mathcal{I} \end{aligned}$$

$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & \lambda^T \cdot \mathbf{s} + z \geq \alpha \quad \forall (\mathbf{s}, z) \in \mathcal{I} \\ & \lambda \geq \mathbf{0} \end{aligned}$$

(a)

If  $\exists \alpha_b$  soln to (b),  
 s.t.  $\lambda^T g(\alpha_b) + f(\alpha_b) = \alpha_b$   
 then  $s = g(\alpha_b)$   $z = f(\alpha_b)$   
 is soln to (a)

complete proof of equivalence

$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & \lambda^T \cdot \mathbf{g}(\mathbf{x}) + f(\mathbf{x}) \geq \alpha \quad \forall \mathbf{x} \in \mathcal{D} \\ & \lambda \geq \mathbf{0} \end{aligned}$$

(b)

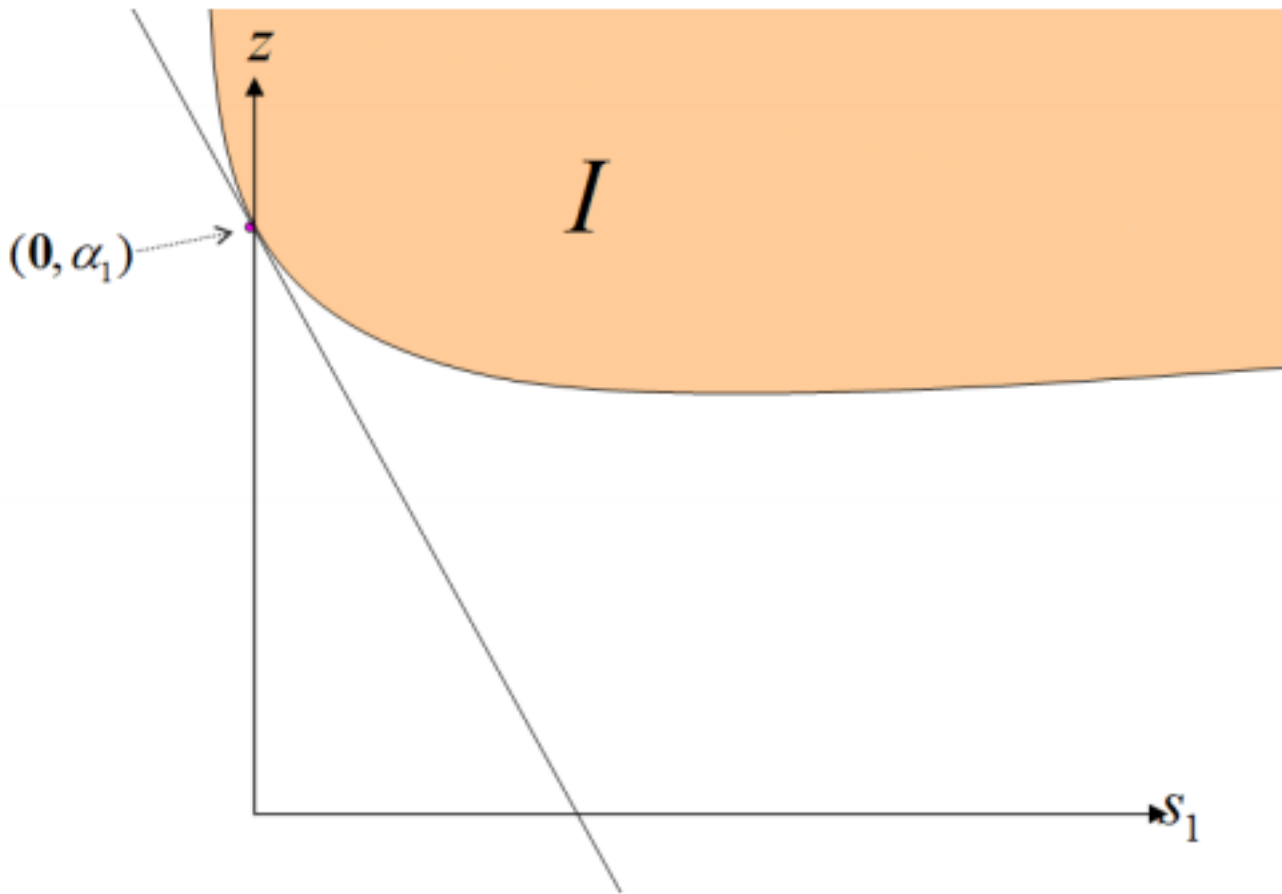
$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & L(\mathbf{x}, \lambda) \geq \alpha \quad \forall \mathbf{x} \in \mathcal{D} \\ & \lambda \geq \mathbf{0} \end{aligned}$$

Since,  $L^*(\lambda) = \min_{\mathbf{x} \in \mathcal{D}} L(\mathbf{x}, \lambda)$ , we can deal with the equivalent

$$\begin{aligned} \max \quad & \alpha \\ \text{subject to} \quad & L^*(\lambda) \geq \alpha \\ & \lambda \geq \mathbf{0} \end{aligned}$$

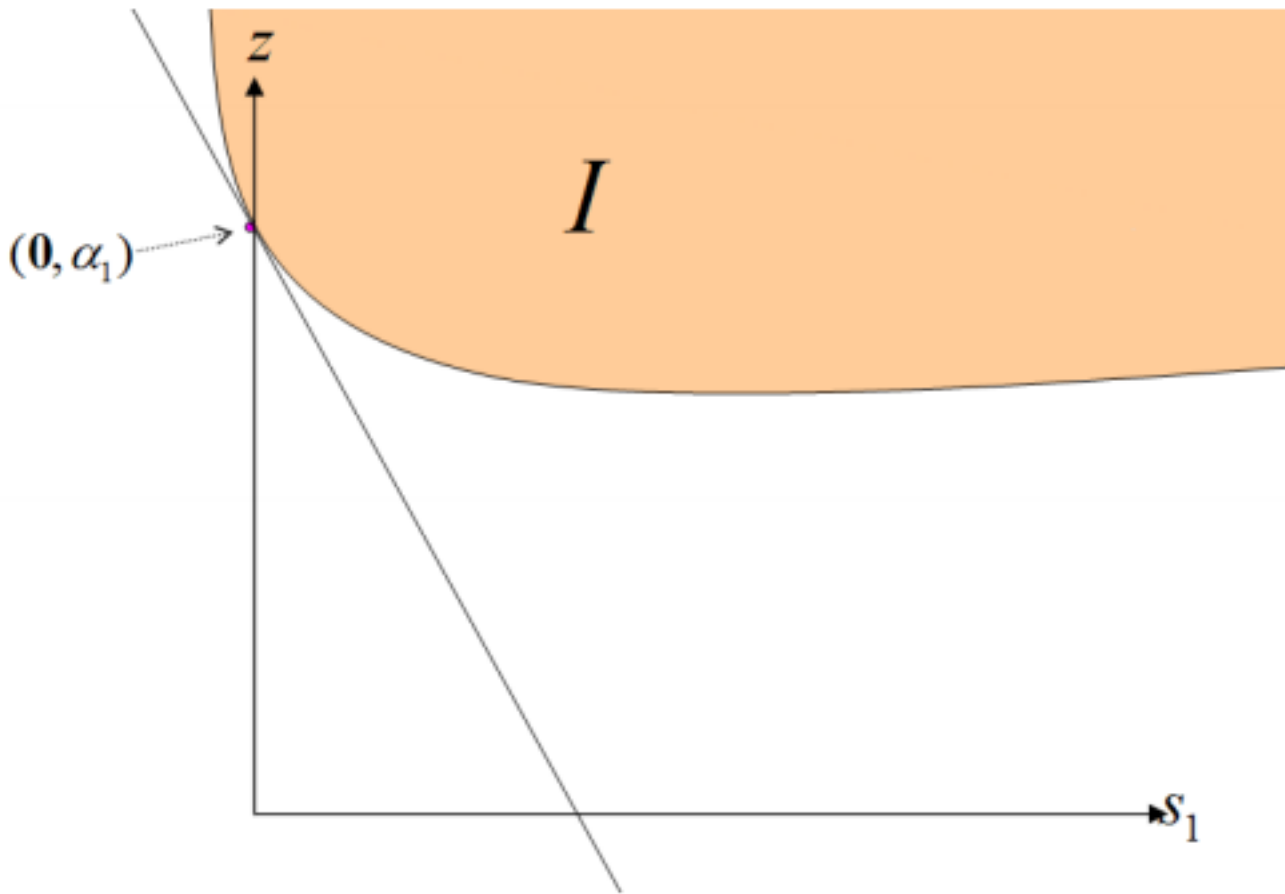
This problem can be restated as

$$\begin{aligned} \max \quad & L^*(\lambda) \\ \text{subject to} \quad & \lambda \geq \mathbf{0} \end{aligned}$$



Q: What is desirable of the set  $I$  for zero duality gap?

- ①  $I$  should be convex i.e.  $f$  &  $g_i$ 's are convex
- ②  $(0, \delta_1)$  should exist i.e.  $I$ 's intersection with  $s=0$  should be closed



Q: What is desirable of the set  $I$  for zero duality gap?

Ans:  $\exists (0, \alpha) \in I$  and  $\lambda$  s.t.  $\lambda^T s + z \geq \alpha \quad \forall (s, z) \in I$

& Intersection of  $I$  with  $z$  axis is closed below

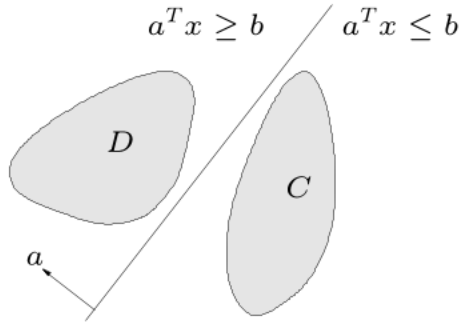
$\Leftrightarrow \exists$  a supporting hyperplane to  $I$  at  $(0, \alpha)$   
& Intersection of  $I$  with  $z$  axis is closed below (with  $(0, \alpha)$  being boundary pt)

$\Leftarrow I$  is closed &  $\exists$  a supporting hyperplane to  $I$  at every boundary point

## Separating hyperplane theorem

if  $C$  and  $D$  are disjoint convex sets, then there exists  $a \neq 0, b$  such that

$$a^T x \leq b \text{ for } x \in C, \quad a^T x \geq b \text{ for } x \in D$$



the hyperplane  $\{x \mid a^T x = b\}$  separates  $C$  and  $D$

strict separation requires additional assumptions (e.g.,  $C$  is closed,  $D$  is a singleton)

Convex sets

2-19

consequence

## Supporting hyperplane theorem

**supporting hyperplane** to set  $C$  at boundary point  $x_0$ :

$$\{x \mid a^T x = a^T x_0\}$$

where  $a \neq 0$  and  $a^T x \leq a^T x_0$  for all  $x \in C$



**supporting hyperplane theorem:** if  $C$  is convex, then there exists a supporting hyperplane at every boundary point of  $C$

Convex sets

2-20

Proof (from separating hyperplane theorem):

(a)  $\text{interior}(C) \neq \emptyset$  (so that  $\text{int}(C) \cap \{x_0\} = \emptyset$ )

Apply separating hyperplane theorem

to sets  $C' = \{x_0\}$  and  $D' = \text{interior}(C)$

$$\exists a, b \text{ s.t. } a^T x \geq b \quad \forall x \in D'$$

$$a^T x_0 \leq b$$

This inequality extends to all boundary (limit pts) leading to  $a^T x_0 \geq b$

$$\text{combined: } a^T x_0 = b$$

Strict separation not applicable since  $\text{int}(C)$  is open

(b)  $\text{int}(C) = \emptyset$

$\Rightarrow$  Any hyperplane containing that affine set contains  $C$  &  $x_0$  (H/w: Prove)

$\Rightarrow$  This hyperplane is a trivial supporting hyperplane

## Some topological concepts: Topological set is set with concept of nbrhood

- ① A set  $U$  is called an open set if it does not contain any of its boundary pts. If  $S$  is a metric space (eg an inner product space) with distance metric  $d(x, y)$ , then a subset  $U$  of  $S$  is called open if, given any  $x \in U$ ,  $\exists \epsilon > 0$  such that given any  $y \in S$  with  $d(x, y) < \epsilon$ ,  $y \in U$ .
- ② A set  $V \subseteq S$  is called closed if its complement  $S \setminus V$  is an open set.



③  $x \in S$  is called an interior point of  $S$  if there exists a neighborhood of  $x$  contained in  $S$ . If  $S$  is a metric space, then  $x \in S$  is an interior pt if  $\exists \epsilon > 0$  s.t.  $\forall y$  s.t.  $d(x, y) < \epsilon, y \in S$

The set of all interior pts of  $S$  form the interior of  $S$ . Thus, if  $S$  is a metric space!

$$\text{int}(S) = \left\{ x \mid \exists \epsilon > 0 \text{ s.t. } \forall y \text{ s.t. } d(x, y) < \epsilon, y \in S \right\}$$

interior of  $S$

What can I say if  $\text{interior}(C) = \emptyset$

eg: sufficient conditions

- ①  $C \subseteq$  hyperplane. In particular, if  $S$  is affine this is necessary & sufficient
- ② eg: a shell

③ eg:  $C = \partial K$  in topological space  $S$   
 (see next page for defn of boundary of set  $X$  denoted by  $\partial X$ )

④ The set of pts of a set  $S$  s.t. every neighborhood of a point from the set consists of at least one point in  $S$  and one point not in  $S$  is called the **boundary**  $\partial S$  of  $S$ . If  $S$  is a metric space

$$\partial S = \left\{ x \in S \mid \forall \epsilon > 0, \exists y \text{ s.t. } d(x, y) < \epsilon \text{ \& } y \in S \text{ and } \exists y' \text{ s.t. } d(x, y') < \epsilon \text{ \& } y' \notin S \right\}$$

⑤ Let  $S$  be a subset of a topological space  $X$ . A point  $x \in X$  is a limit point of  $S$  if every neighborhood of  $x$  contains at least one point of  $S$  different from  $x$  itself.

→ can be relaxed to open neighborhoods without loss

If  $S$  happens to have an associated metric  $d$ , and  $A \subseteq S$ , then  $x \in S$  is a limit point of  $A$  iff:

$$\forall \epsilon > 0 : \{ x \in A \text{ s.t. } 0 < d(x, a) < \epsilon \} \neq \emptyset$$

Informally speaking,  $x$  is a limit point of  $A$  if there are points in  $A$  that are different from  $x$  but arbitrarily close to it

[Note:  $x$  need not belong to  $A$ ]

⑥ Closure of  $S$   $cl(S) = S \cup \{ \text{limit points of } S \}$

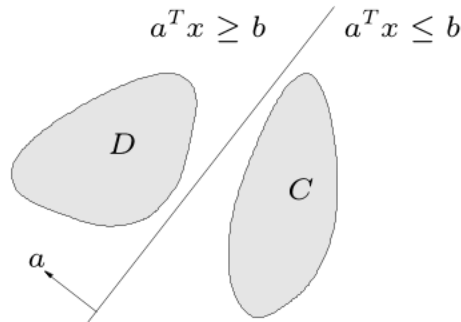
Some standard results that we will regularly invoke for topological spaces

- ① Intersection of (even uncountable) closed sets is closed
- ② Union of (even uncountable) open sets is open
- ③ Intersection of finite number of open sets is open
- ④ Union of finite number of closed sets is closed
- ⑤  $S$  is closed iff  $S^c$  is open

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**Proof:** Let  $S = \{x - y \mid x \in C, y \in D\}$ .

Now we can prove (see <http://www.cse.iitb.ac.in/~cs709/notes/eNotes/ExtraProblems-1.pdf>, Q1) that  $S$ , being a sum of two convex sets, is convex.

Since  $C \cap D = \emptyset$ ,  $0 \notin S$ .

(a) Suppose  $0 \notin \text{cl}(S)$ : Consider the sets  $\{0\}$  and  $\text{cl}(S)$ . We will prove that  $\exists a \neq 0$  s.t.

$$a^T z > 0 \quad \forall z \in \text{cl}(S) \quad \& \quad a^T w = 0 \quad \text{for } w \in \{0\}$$

Q: How to choose 'a'?

complete proof given in class: H/w

obvious

i.e.  $\exists a$  s.t.  $a^T(x-y) > 0 \quad \forall x-y \in S$

i.e.  $a^T x > a^T y \quad \forall x \in C \text{ \& } y \in D$

Let  $b = \inf_{x \in C} a^T x$ . Then we proved existence

of  $a$  &  $b$  s.t.

$$a^T x \geq b \quad \forall x \in C \quad \& \quad a^T y \leq b \quad \forall y \in D$$

⑥ suppose  $0 \in \text{cl}(S)$ . Since  $0 \notin S$ ,  $0 \in \text{bdry}(S)$   
if  $\text{interior}(S) = \emptyset$  (empty),  $S$  must be  $\subseteq \{z \mid a^T z = b\}$

& the hyperplane must include  $0$  on  $\text{bdry}(S)$  ↓  
A hyperplane

$\Rightarrow b = 0$ . i.e.  $a^T x = a^T y \quad \forall x \in C \text{ \& } y \in D$

$\Rightarrow$  we have a trivial separating hyperplane

limit by  $\bar{a}$ , we have

$$a(\epsilon_k)^T z > 0 \quad \forall z \in S_{-\epsilon_k}$$

for all  $k$  & therefore

$$\bar{a}^T z > 0 \quad \forall z \in \text{interior}(S)$$

and

$$\bar{a}^T z \geq 0 \quad \forall z \in S \quad \leftarrow \text{proof by contradiction}$$

that is

$$\bar{a}^T x \geq \bar{a}^T y$$

$$\forall x \in C \text{ \& } y \in D$$

(use the property that a convex set is connected!)

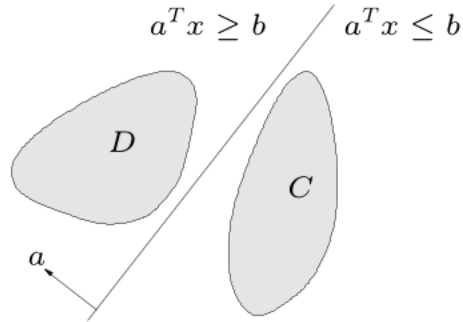
Hence proved!

## Separating hyperplane theorem

Thus

if  $C$  and  $D$  are disjoint convex sets, then there exists  $a \neq 0, b$  such that

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**supporting hyperplane theorem:** if  $C$  is convex, then there exists a supporting hyperplane at every boundary point of  $C$

Convex sets

2-20