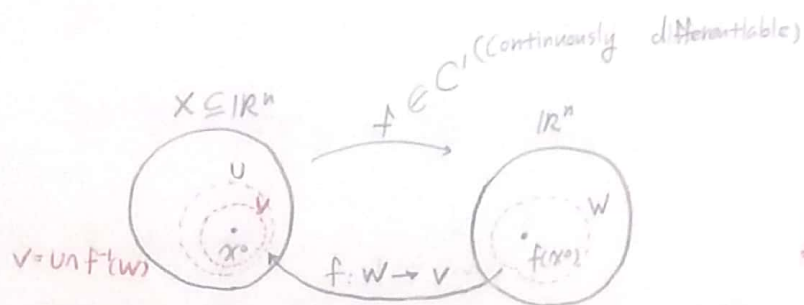


• Inverse function thm review



$$\underline{Df^{-1}(y) = (Df(f^{-1}(y)))^{-1}}$$

$Df(x_0)$  is invertible in  $\mathbb{R}^n$

$f^{-1}$  locally well defined / locally  $C^1$

$f(x) = y : f^{-1}(y)$  changes locally in  $y$ .

**Thm 47** < Implicit Function Theorem >

$$X \times \Omega \subseteq \mathbb{R}^{n+p} \rightarrow \begin{pmatrix} x_1 \\ \vdots \\ x_n \\ \alpha_1 \\ \vdots \\ \alpha_p \end{pmatrix}$$

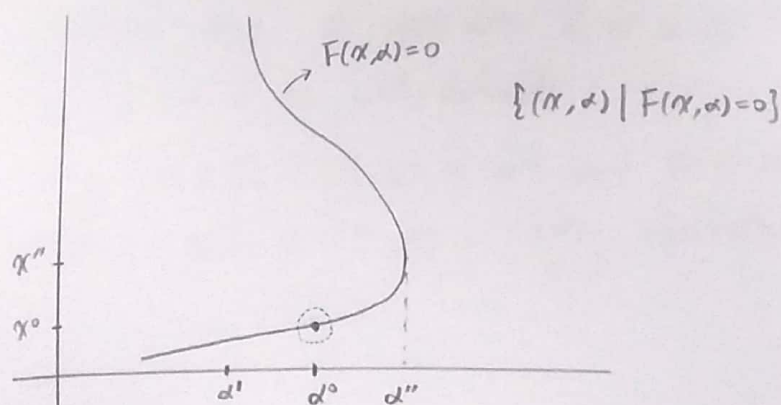
Let  $X \times \Omega \subseteq \mathbb{R}^{n+p}$  be open and  $F: X \times \Omega \rightarrow \mathbb{R}^p$  be continuously differentiable. Consider the system of equations  $F(x, \alpha) = 0$  and assume that given some  $\alpha^0 \in \Omega$  it has a solution  $x^0 \in X$ .

If  $D_x F(x^0, \alpha^0)$  is invertible,

then (i) there exist open sets  $V \subseteq \mathbb{R}^{n+p}$  and  $V_\alpha \subseteq \Omega$  with  $(x^0, \alpha^0) \in V$  and  $\alpha^0 \in V_\alpha$  such that  $\forall \alpha \in V_\alpha, \exists$  a unique  $x_\alpha \in \mathbb{R}^n$  s.t.  $(x_\alpha, \alpha) \in V$  and  $F(x_\alpha, \alpha) = 0$ .

(ii) the solution function  $x: V_\alpha \rightarrow \mathbb{R}^n : \alpha \mapsto x_\alpha$  is continuously differentiable and its derivative is given by  $D_{x_\alpha} = -[D_x F(x_\alpha, \alpha)]^{-1} \circ D_\alpha F(x_\alpha, \alpha)$

**Def**  $D_x F(x, \alpha) = \left[ \frac{\partial F^i(x, \alpha)}{\partial x_j} \right]$



\* The Inverse function thm as a special case of the Implicit Function theorem.

If  $F(x, y) := f(x) - y = 0$ , with Implicit Function theorem,

$$Df^{-1}(y) = D\alpha(y) = [D_Y(F(x_y, y))]^{-1} = (Df(f^{-1}(y)))^{-1}$$

proof of thm 4.9)

(i) Consider the function  $G: X \times \Omega \rightarrow \mathbb{R}^{n+p}$  defined by

$$G^i(x, \alpha) = F^i(x, \alpha) \quad \text{for } i=1, \dots, n$$

$$G^i(x, \alpha) = \alpha_i \quad \text{for } i=n+1, \dots, n+p$$

Note that  $G(x^0, \alpha^0) = \begin{pmatrix} 0 \\ \alpha^0 \end{pmatrix}$  and

$$DG(x^0, \alpha^0) = \begin{pmatrix} D_x F(x^0, \alpha^0) & D_\alpha F(x^0, \alpha^0) \\ 0 & I \end{pmatrix}$$

Claim: The associated linear operator is invertible.

Suppose  $\begin{pmatrix} \lambda \\ \mu \end{pmatrix} \in \ker DG(x^0, \alpha^0)$

$$\text{Then } \begin{pmatrix} D_x F(x^0, \alpha^0) & D_\alpha F(x^0, \alpha^0) \\ 0 & I \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_n \\ \mu_1 \\ \vdots \\ \mu_p \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \text{ which implies } \mu_i = 0 \text{ for } i=1, \dots, p$$

Since  $D_x F(x^0, \alpha^0)$  is invertible (by assumption), this implies  $\lambda_i = 0$  for  $i=1, \dots, n$

Hence, we can apply the inverse function theorem.

and conclude that  $\exists$  open sets  $V, W \subseteq \mathbb{R}^{n+p}$  such that  $(x^0, \alpha^0) \in V$ ,  $(0, \alpha^0) \in W$  and  $G^{-1}: W \rightarrow V$  is well-defined.

Let  $V_\alpha = \{x \in \Omega \mid (0, \alpha) \in W\}$  and  $\forall \alpha \in V_\alpha$ ,  $\exists$  a unique  $x(\alpha) \in X$  such that  $(x(\alpha), \alpha) \in V$  and  $G^{-1}(0, \alpha) = \begin{pmatrix} x(\alpha) \\ \alpha \end{pmatrix}$  equivalent to  $G(x(\alpha), \alpha) = \begin{pmatrix} 0 \\ \alpha \end{pmatrix}$

(ii) The Inverse Function Theorem also implies that  $G^{-1}$  is continuously differentiable. Since  $G^{-1}(\underline{0}, \alpha) = \begin{pmatrix} x(\alpha) \\ \alpha \end{pmatrix}$ , this implies that  $x(\alpha)$  is continuously differentiable as well.

Now, viewing  $G$  as a function of  $\alpha$ , the chain rule and the definition of  $x(\alpha)$  implies

$$DG(x(\alpha), \alpha) = \begin{pmatrix} D_x F(x(\alpha), \alpha) & D_\alpha F(x(\alpha), \alpha) \\ \underline{0} & I_p \end{pmatrix} \circ \begin{pmatrix} Dx(\alpha) \\ I_p \end{pmatrix} = \begin{pmatrix} \underline{0} \\ \underline{1} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}^n \begin{pmatrix} \vdots \\ 1 \end{pmatrix}^p$$

Hence,  $D_x F(x(\alpha), \alpha) \cdot Dx(\alpha) + D_\alpha F(x(\alpha), \alpha) \cdot I = \underline{0}$

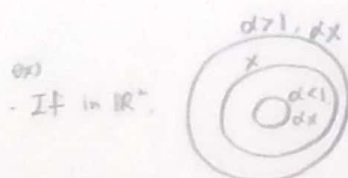
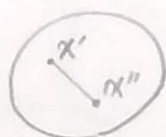
Therefore,  $Dx(\alpha) = -(D_x F(x(\alpha), \alpha))^{-1} \cdot D_\alpha F(x(\alpha), \alpha) \quad \parallel$

## Convex Analysis

**Def.** Let  $V$  be a vector space  $X \subseteq V$

$X$  is convex if,  $\forall x, x'' \in X$  and  $\lambda \in [0, 1]$ ,  $\lambda x' + (1-\lambda)x'' \in X$

- A point  $y \in V$  is a convex combination of the vectors  $x_1, \dots, x_m \in V$  if  $\exists \lambda_1, \dots, \lambda_m \in [0, 1]$  such that  $\sum_{i=1}^m \lambda_i = 1$  and  $y = \sum_{i=1}^m \lambda_i x_i$
- Let  $X \subseteq V$ . The convex hull of  $X$ ,  $\text{conv } X$ , is the smallest convex set that contains  $X$ .



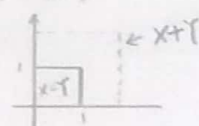
**Thm 48** Let  $V$  be a vector space and  $\alpha \in \mathbb{R}$

If  $X, Y \subseteq V$  are convex, then  $\alpha X := \{z \in V \mid \exists x \in X : z = \alpha x\}$  the Minkowski Sum

and  $X+Y := \{z \in V \mid \exists x \in X, \exists y \in Y : z = x+y\}$  are convex

$\Rightarrow$  If  $X=Y$  in  $\mathbb{R}^2$ ,

Note: We write  $C-D$  as shorthand for  $C+(-D)$



**Thm 49** Let  $V$  be a vector space

A set  $X \subseteq V$  is convex if and only if every convex combination of elements in  $X$  lies in  $X$

**Thm 50** Let  $V$  be a vector space

The convex hull of a set  $X \subseteq V$  is the set of all convex combinations of elements of  $X$

$\rightarrow$  Idea: Let  $\gamma := \{y \in V : \exists x_1, \dots, x_m \in X, \lambda_1, \dots, \lambda_m \in [0, 1], \sum_{i=1}^m \lambda_i = 1 \text{ and } \sum_{i=1}^m \lambda_i x_i = y\}$

One can show  $\gamma$  is convex, which implies  $\gamma \supseteq \text{conv } X$  (Increasing case)

On the other hand, for every convex set  $Z \supseteq X$ ,  $Z$  contains all convex combination of elements of  $X$ . (decreasing case)

Therefore,  $Z \supseteq \gamma$ , which implies  $\gamma = \text{conv } X$ .  $\parallel$



## Separating Hyperplane Theorem

(Thm 51)  $X \subseteq V$  convex  
 $\Rightarrow \text{int } X$  and  $\bar{X}$  are convex

(Def) Let  $h \in \mathbb{R}^n \setminus \{0\}$  and  $\beta \in \mathbb{R}$ .

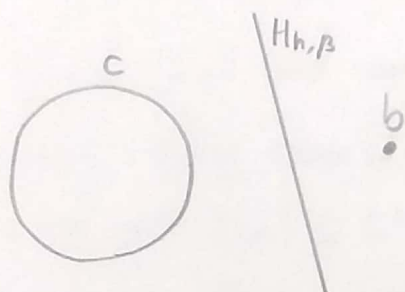
- The set  $H_{h,\beta} = \{x \in \mathbb{R}^n \mid h \cdot x = \beta\}$  is called a hyperplane
- The set  $\{x \in \mathbb{R}^n \mid h \cdot x \leq \beta\}$  is called a halfspace.

(Thm 52) < Strict Separating Hyperplane theorem >

Let  $C \subseteq \mathbb{R}^n$  be closed and convex and  $b \in \mathbb{R}^n$ ,  $b \notin C$ .

Then,  $\exists$  a hyperplane that strictly separates  $b$  and  $C$ ,

that is,  $\exists h \in \mathbb{R}^n$  and  $\bar{\beta} \in \mathbb{R}$  such that  $h \cdot b < \bar{\beta}$  and  $h \cdot x > \bar{\beta}$ ,  $\forall x \in C$



proof) We first prove the result for the translated sets  $\underline{0} = b - b$  and  $D = C - \{b\}$

Pick arbitrary  $y \in C$ , and define  $D' := \{x \in D \mid d(x, \underline{0}) \leq d(y, b)\}$

Note that  $D'$  is closed and bounded and therefore compact (by Heine-Borel thm)

By the extreme value theorem, there exists  $x^0 \in D'$  s.t.  $d(x^0, \underline{0}) = \min_{x \in D'} d(x, \underline{0}) > 0$

Define  $m = \frac{x^0}{\|x^0\|}$ ,  $h = x^0$ ,  $\beta = m \cdot h = \frac{\|x^0\|^2}{\|x^0\|} > 0$

Consider arbitrary  $x \in D$ ,  $x \neq x^0$  and define  $\forall \lambda \in [0, 1]$ ,  $x^\lambda := x^0 + \lambda(x - x^0)$

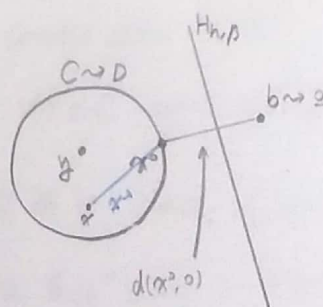
By definition of  $x^0$ ,

$$\|x^0\|^2 \leq \|x^\lambda\|^2 = (x^0 + \lambda(x - x^0)) \cdot (x^0 + \lambda(x - x^0)) = \|x^0\|^2 + 2\lambda x^0 \cdot (x - x^0) + \lambda^2 \|x - x^0\|^2$$

This implies for all  $\lambda \in (0, 1)$   $0 \leq 2x^0 \cdot (x - x^0) + \lambda \|x - x^0\|^2$

Since this holds for all  $\lambda > 0$ , it must also hold for  $\lambda = 0$ .

Hence, for all  $x \in D$ ,  $0 < \beta < \|x^0\|^2 \leq h \cdot x$



(Translate back)

Adding  $h \cdot b$  to all inequalities and setting  $\bar{\beta} := \beta + hb$ ,

We get  $hb < \beta + hb < h(x+tb)$ ,  $\forall x \in D$

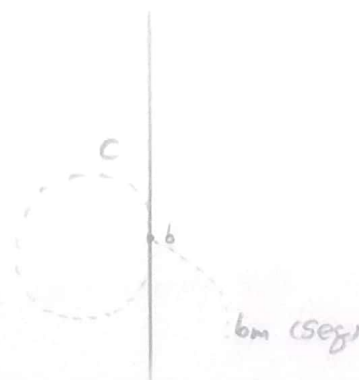
or  $hb < \bar{\beta} < hy$ ,  $\forall y \in C$ . ||

**Thm 53** <Weak Separating Hyperplane Theorem>

Let  $C \subseteq \mathbb{R}^n$  be convex and  $b \in \mathbb{R}^n$ ,  $b \notin C$

Then  $\exists$  a hyperplane that separates  $b$  and  $C$ ;

that is,  $\exists h \in \mathbb{R}^n$ ,  $\beta \in \mathbb{R}$  such that  $hb \leq \beta \leq hx$ ,  $\forall x \in C$ .



Idea)  $\bar{C}$  is closed and convex, and either  $b \notin \bar{C}$  or  $b \in \partial C$

For  $b \notin \bar{C}$  apply (thm 52) if  $b \in \partial C$

$\exists$  a sequence  $\{b_m\} \subseteq C^\circ$  that converges to  $b$

For each  $b_m$ , thm 52 applies that  $\exists h_m \in \mathbb{R}^n$ ,  $\beta_m \in \mathbb{R}$  s.t.  $h_m b_m < \beta_m < h_m x$ ,  $\forall x \in C$

Assume WLOG,  $\|h_m\| = 1$  since  $\{z \in \mathbb{R}^n \mid \|z\| = 1\}$  is compact.

Then, there exists a convergent subsequence  $h_{m_k}$

$hb = \lim_{k \rightarrow \infty} h_{m_k} \cdot b_{m_k} \leq \lim_{k \rightarrow \infty} h_{m_k} x = hx$ ,  $\forall x \in C$ , setting  $\beta = hb$  complete the proof.

**Thm 54** Let  $C$  and  $D$  be non-empty, disjoint and convex sets in  $\mathbb{R}^n$ .

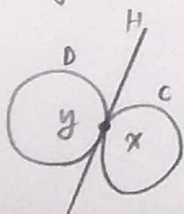
Then there is a hyperplane  $(h, \beta)$  such that  $\forall x \in C$  and  $y \in D$ ,  $hx \geq \beta \geq hy$ .

proof: Note that  $K := C - D$  is convex and  $0 \notin K$  since  $C$  and  $D$  are disjoint.

By thm 53, there exists  $(h, \beta')$  such that  $h \cdot 0 \leq \beta' \leq hz$ ,  $\forall z \in K$ .

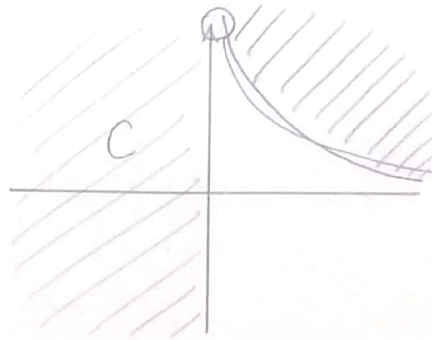
Given arbitrary  $x \in C$  and  $y \in D$ ,  $z := x - y \in K$ . Hence  $0 \leq hx - hy$ .

This yields  $hx \geq \beta := \inf_{x' \in C} hx' \geq hy$ ,  $\forall x \in C$  and  $y \in D$ . ||



Remark: Even if sets are closed, strict separation is not always possible.

ex)



finally, this part will converge so that  $\neq$  strict hyperplanes.

**Thm 5.5** Let  $C$  and  $D$  be two non-empty, closed, disjoint and convex sets. Suppose  $C$  is bounded.

Then  $\exists$  a hyperplane  $(h, \beta)$  such that  $hx > \beta > hy$ ,  $\forall x \in C$  and  $y \in D$  (strictly separated!)

Idea: 1) Show that  $K := C - D$  is closed (Bolzano-Weierstrass)  
2) Apply strict separating hyper plane theorem.