

# Multi-Agent Optimization

## I. Variational Analysis Tools (2)

# Variational Analysis: Bivariate Functions

Roger J-B Wets  
University of California, Davis



# Collaborators & Contributors

- ◆ Alejandro Jofré, Universidad de Chile
- ◆ Teemu Pennanen, School of Economics, Helsinki
- ◆ Hedy Attouch, Université de Montpellier
- ◆ Dominique Azé, Université de Toulouse
- ◆ Contributors:
  - Jean-Pierre Aubin, Michael Casey, Ruben López
  - Adib Bagh, Michel Volle, Mohammed Soueycatt,
  - Steve Wright, Mohamed Ait Mansour, Chi Do, ....

# SADDLE-POINTS

# Lagrangians

◆  $\max f_0(x)$

so that  $f_i(x) \geq 0, i = 1, \dots, s$

$f_i(x) = 0, i = s + 1, \dots, m.$

◆ Lagrangian:

$$L(x, y) = \begin{cases} f_0(x) + \sum_{i=1}^m y_i f_i(x) & \text{if } y_i \geq 0, i = 1, \dots, s \\ -\infty & \text{otherwise} \end{cases}$$

◆ concave-convex, **maxinf framework**

# Hamiltonians

- ◆ Opt-control:  $\min_u \Phi(x_T) + \int_0^T l(t, x_t, u_t) dt,$   
 $\dot{x}_t = f(t, u_t, x_t), u_t \in U_t, t \in [0, T].$
- ◆ Hamiltonian:  $H(t, x, y, u) = l(t, u, x) + \langle y, f(t, x, u) \rangle$
- ◆ Optimality:  $\dot{x}^* = \frac{\partial H^*}{\partial y}, \dot{y}^* = -\frac{\partial H^*}{\partial x}, y^*(T) = \nabla \Phi(x_T^*)$   
 $u_t^* \in \arg \min_{v \in U_t} H(t, x_t^*, v, y_t^*), t \in [0, T]$
- ◆ Approximation: state  $x$  and co-state  $y$

# Hypo/Epi-Convergence



saddle point  $L^a \sim$  saddle point  $L$



Attouch-W. def. (90's):

$\forall (x, y)$

(a)  $\forall x^\nu \rightarrow x, \exists y^\nu \rightarrow y : \liminf L^\nu(x^\nu, y^\nu) \geq L(x, y)$

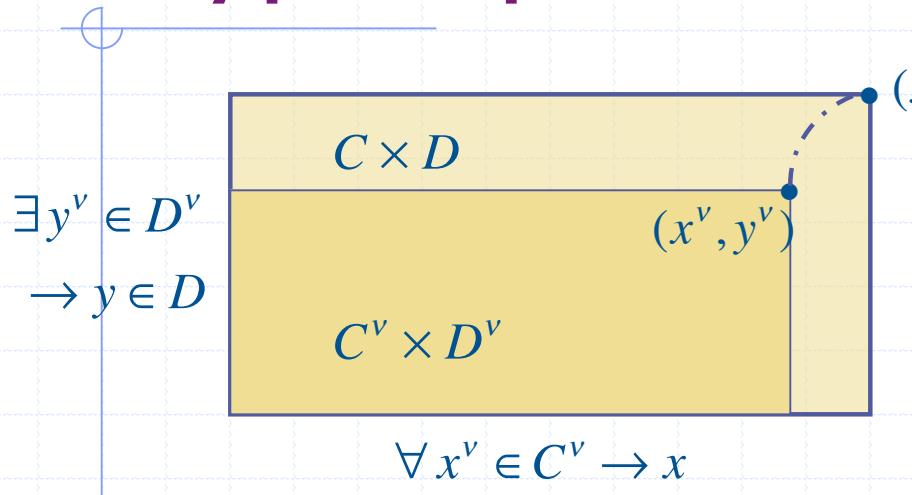
(b)  $\forall y^\nu \rightarrow y, \exists x^\nu \rightarrow x : \limsup L^\nu(x^\nu, y^\nu) \leq L(x, y)$



“problem”: hypo/epi-topology not Hausdorff  
.... the pitfall “equivalence classes”



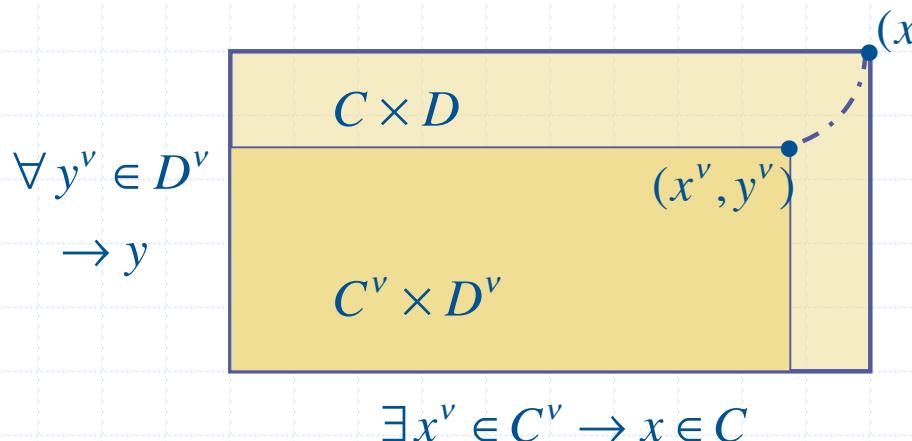
# Hypo/Epi convergence - 21st C.



$\exists y^\nu \in D^\nu \rightarrow y \in D$

$\limsup_\nu K^\nu(x^\nu, y^\nu) \leq K(x, y)$  when  $x \in C$

$K^\nu(x^\nu, y^\nu) \rightarrow -\infty$  when  $x \notin C$



$\forall y^\nu \in D^\nu \rightarrow y$

$\liminf_\nu K^\nu(x^\nu, y^\nu) \geq K(x, y)$  when  $y \in D$

$K^\nu(x^\nu, y^\nu) \rightarrow \infty$  when  $x \notin D$

# Hypo/epi: max-inf framework



$$L : C \times D \rightarrow \mathbb{R}, \quad \left\{ L^\nu : C^\nu \times D^\nu \rightarrow \mathbb{R} \right\}$$

(a)  $\forall x^\nu \in C^\nu \rightarrow x \in C, \forall y \in D, \exists y^\nu \in D^\nu \rightarrow y :$

$$\limsup_\nu L^\nu(x^\nu, y^\nu) \leq L(x, y)$$

(a<sub>∞</sub>)  $\forall x^\nu \in C^\nu \rightarrow x \notin C, \forall y \in D, \exists y^\nu \in D^\nu \rightarrow y :$

$$L^\nu(x^\nu, y^\nu) \rightarrow -\infty$$

(b)  $\forall y^\nu \in D^\nu \rightarrow y \in D, \forall x \in C, \exists x^\nu \in C^\nu \rightarrow x :$

$$\liminf_\nu L^\nu(x^\nu, y^\nu) \geq L(x, y)$$

(a<sub>∞</sub>)  $\forall y^\nu \in D^\nu \rightarrow y \notin D, \forall x \in C, \exists x^\nu \in C^\nu \rightarrow x :$

$$L^\nu(x^\nu, y^\nu) \rightarrow \infty$$

# Hypo/epi: elementary properties

saddle points  $L^a \sim$  saddle points  $L$

hypo/epi  $\Rightarrow$  epi-convergence if  $L(x, y) = g(y)$

hypo/epi  $\Rightarrow$  hypo-convergence if  $L(x, y) = f(x)$

but not  $y$ -epi-convergence or  $x$ -hypo-convergence

Remarks: not a  $\Gamma$ -convergence (De Giorgi)  
in  $\infty$ -dim. different topologies ( $\sigma$ -hypo/ $\tau$ -epi)

# Hypo/epi-convergence: Properties

$L^\nu \xrightarrow[h/e]{} L, \quad (x^\nu, y^\nu) \text{ saddle-pts } L^\nu,$

$(x^\nu, y^\nu) \xrightarrow[\nu \in N]{} (\bar{x}, \bar{y}) \Rightarrow (\bar{x}, \bar{y}) \text{ saddle-pt of } L$

Moreover,  $F(\bar{x}, \bar{y}) = \lim_{\nu \in N} F^\nu(x^\nu, y^\nu).$

$\{L \text{ concave-convex}\}$  closed under epi/hypo-convergence

hypo/epi-limits:  $L(\cdot, y)$  usc,  $L(x, \cdot)$  lsc

bijection  $\eta: fv\text{-biv}(\mathbb{R}^{n+m}) \leftrightarrow pr\text{-biv}(\mathbb{R}^{n+m})$

# Zero-sum games

◆ Strategies:  $x \in X, y \in Y$  - Players: Max-1 & Minie-2

Payoff:  $u_1(x, y) + u_2(x, y) = 0$ ; set  $u = u_1$

Nash:  $x^* \in \arg \max_{x \in X} u(x, y^*)$ ,  $y^* \in \arg \min_{y \in Y} u(x^*, y)$

◆ Existence:  $X, Y$  compact, convex

$u(\cdot, y)$  concave-usc,  $u(x, \cdot)$  convex-lsc

$\Rightarrow \exists$  optimal strategies (equilibrium)

◆ Convergence:  $u^\nu : X^\nu \times Y^\nu \rightarrow \mathbb{R}$  (as for existence),

$X^\nu \rightarrow X, Y^\nu \rightarrow Y$  and  $u^\nu \xrightarrow{h/e} u \Rightarrow$

$\exists$  soln's  $(x^\nu, y^\nu) \xrightarrow{\text{cluster}} (\bar{x}, \bar{y})$  equilibrium for  $u : X \times Y \rightarrow \mathbb{R}$

# MAX-INF POINTS (maxinf)

or MIN-SUP POINTS

# Fixed Points (Brouwer)

$C$  compact convex,  $G : C \rightarrow C$  continuous

$C^\nu$  compact convex,  $G^\nu : C^\nu \rightarrow C^\nu$  continuous

$\bar{x}^\nu \in C^\nu$  fixed points of  $G^\nu$  on  $C^\nu$

and  $\bar{x}^\nu \rightarrow \bar{x}$ . When is  $\bar{x}$  a fixed point of  $G$  on  $C$ ?

**The approach**  $K : C \times C \rightarrow \mathbb{R}$ ,  $K^\nu : C^\nu \times C^\nu \rightarrow \mathbb{R}$ ,

set  $K(x, y) = \langle G(x) - x, y \rangle$ ,  $K^\nu(x, y) = \langle G^\nu(x) - x, y \rangle$

$x$  fixed point  $\Leftrightarrow x \in \arg \max_C (\inf_{y \in C} K(\bullet, y))$

So,  $K^\nu \xrightarrow{?} K$  yields convergence of max-inf points?

# Variational Inequalities

- ◆  $C \subset \mathbb{R}^n$  non-empty, convex
- ◆  $G: C \rightarrow \mathbb{R}^n$  continuous
- ◆ find  $\bar{u} \in C$  such that  $-G(\bar{u}) \in N_C(\bar{u})$   
where  $v \in N_C(\bar{u}) \Leftrightarrow \langle v, u - \bar{u} \rangle \leq 0, \forall u \in C$
- ◆  $C^\nu \rightarrow C, G^\nu: C^\nu \rightarrow \mathbb{R}^n$  continuous
- ◆ approx.  $u^\nu \in C^\nu$  such that  $-G^\nu(u^\nu) \in N_{C^\nu}(u^\nu)$
- ◆ Question:  $u^\nu \rightarrow \bar{u}$  ? when, how.

# V.I.: The approach

- ◆ Let  $K(u, v) = \langle G(u), v - u \rangle$  on  $\text{dom } K = C \times C$
- ◆ then  $-G(\bar{u}) \in N_C(\bar{u})$  if and only if
  - ◆  $\exists \hat{u} \in \arg \max \inf K$  with  $K(\hat{u}, \cdot) \geq 0$ 
    - $\langle G(\bar{u}), v - \bar{u} \rangle = K(\bar{u}, v) \geq 0, \forall v \in C$   
 $\Rightarrow 0 \leq K(\bar{u}, \cdot) \leq \sup_{u \in C} (\inf_{v \in C} K(u, v))$
    - $\hat{u} \in \arg \max \inf K$  &  $K(\hat{u}, \cdot) \geq 0 \Rightarrow$   
 $\langle -G(\hat{u}), v - \hat{u} \rangle \leq 0, \forall v \in C$  or  $-G(\hat{u}) \in N_C(\hat{u})$

# V.I.:The approach

◆  $K^v(u, v) := \langle G^v(u), v - u \rangle$ ,  $\text{dom } K^v = C^v \times C^v$

◆  $u^v \in \arg \max -\inf K^v$  with  $K^v(u^v, \cdot) \leq 0$



$K^v \rightarrow K$  and ...  
"..."

◆  $\bar{u} \in \text{cluster points } \{u^v\} \Rightarrow ? \bar{u} \in \arg \min -\sup K$

# Non-Cooperative Games

- ◆ player:  $a \in \mathcal{A}$ , payoff:  $u_a(x_a, x_{-a}) : \mathbb{R}^N \rightarrow \bar{\mathbb{R}}$
- ◆ Nash equilibrium:  $(\bar{x}_a, a \in \mathcal{A})$  such that
$$\bar{x}_a \in \arg \max_{x_a} u_a(x_a, \bar{x}_{-a}), \forall a \in \mathcal{A}$$
- ◆ Nikaido-Isoda function:
$$N(x, y) = \sum_{a \in \mathcal{A}} u_a(x_a, x_{-a}) - \sum_{a \in \mathcal{A}} u_a(y_a, x_{-a})$$
- ◆  $\bar{x} = (\bar{x}_a, a \in \mathcal{A})$  is a Nash equilibrium
$$\Leftrightarrow \bar{x} \in \arg \max - \inf N$$

# The approach:

- ◆ Nikaido-Isoda functions of approximating games

$$N^v(x, y) = \sum_{a \in \mathcal{A}} u_a^v(x_a, x_{-a}) - \sum_{a \in \mathcal{A}} u_a^v(y_a, x_{-a})$$

- ◆  $x^v \in \arg \max-\inf N^v$ ,  $\bar{x} \in$  cluster points  $\{x^v\}$

$N^v \rightarrow N$  and ...  
"..."

- ◆  $\Rightarrow ? \bar{x} \in \arg \max-\inf N \sim$  equilibrium point

# Applications: Convergence and stability

- ◆ Saddle-points: Lagrangians, Hamiltonians
- ◆ Fixed points
- ◆ Solutions of cooperative and  
non-cooperative games
- ◆ Economic-Equilibrium points (Walras)
- ◆ Generalized Nash Equilibrium Problems
- ◆ Solutions of set-valued inclusions
- ◆ Stability of mountain-pass paths

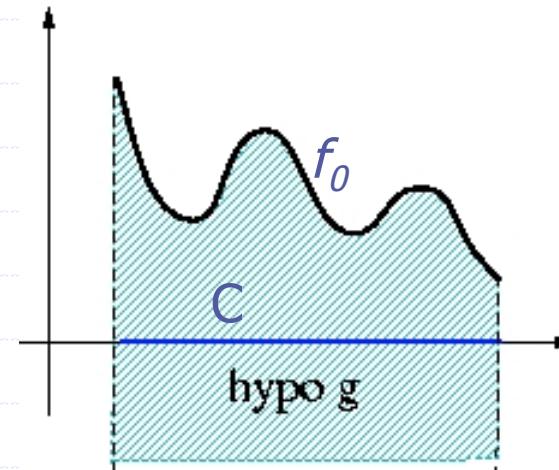
# Optimization: Max-Framework

◆  $\max f(x), x \in \mathbb{R}^n$  with

$$f : \mathbb{R}^n \rightarrow \bar{\mathbb{R}} = [-\infty, \infty]$$

◆  $f(x) = \begin{cases} f_0(x) & \text{if } x \in C \subset \mathbb{R}^n \\ -\infty & \text{if } x \notin C \end{cases}$

$$C = \left\{ x \in X \mid f_i(x) \leq 0, i \in I_1, f_i(x) = 0, i \in I_2 \right\}$$



# Hypo-Convergence (max-framework)



$$\operatorname{argmax} f^a \sim \operatorname{argmax} f$$



$$\operatorname{argmax} (f^\nu + g) \rightarrow \operatorname{argmax} (f+g),$$

$\forall$  cont.  $g \Rightarrow$  hypo-convergence



'new' definition  $f : D \rightarrow \mathbb{R}$ ,  $\left\{ f^\nu : D^\nu \rightarrow \mathbb{R} \right\}_{\nu \in \mathbb{N}}$

(a)  $\forall x^\nu \in D^\nu \rightarrow x \in D, \limsup f^\nu(x^\nu) \leq f(x)$

(a <sub>$\infty$</sub> )  $\forall x^\nu \in D^\nu \rightarrow x \notin D, f^\nu(x^\nu) \rightarrow -\infty$

(b)  $\forall x \in D, \exists x^\nu \rightarrow x, \liminf f^\nu(x^\nu) \geq f(x)$

# Hypo-Convergence



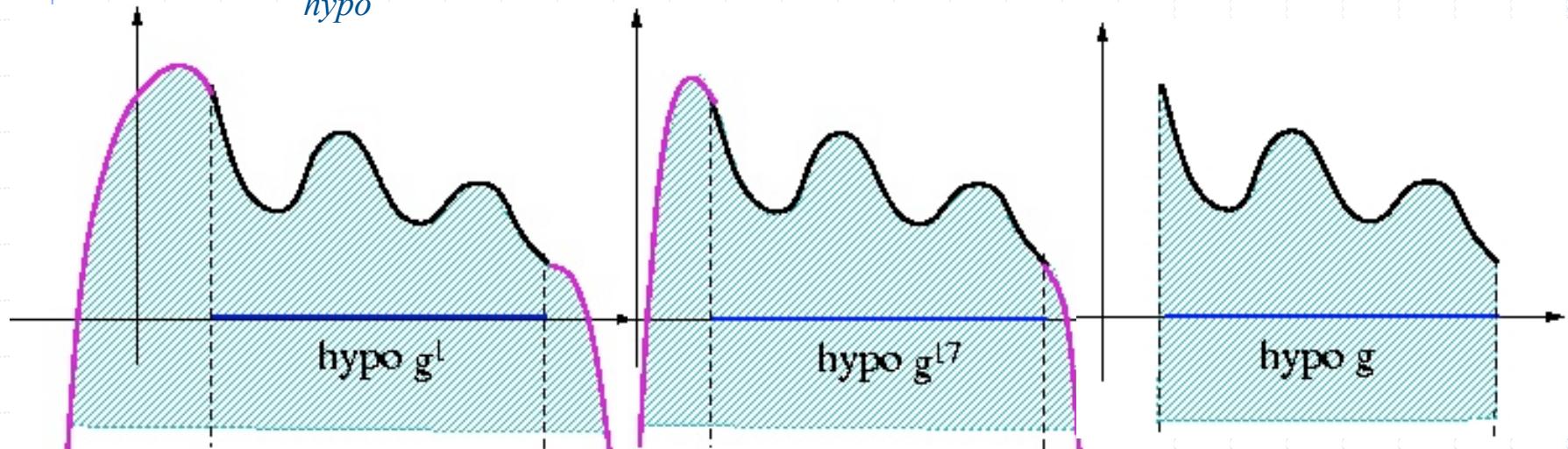
$$\operatorname{argmax} g^a \sim \operatorname{argmax} g$$



$$g^v \xrightarrow{h} g \Leftrightarrow -g^v \xrightarrow{\text{epi}} -g$$

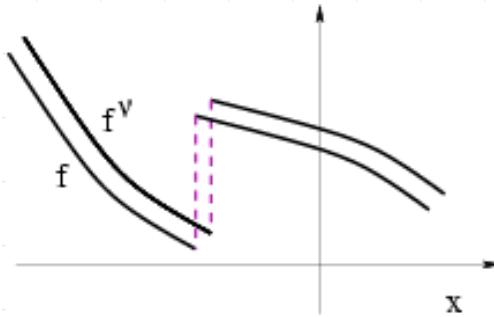


$$g^v \xrightarrow{\text{hypo}} g \Leftrightarrow \text{hypo } g^v \rightarrow \text{hypo } g$$



# Hypo-convergence: Properties

- ◊  $f^v \xrightarrow{h} f \neq f^v \xrightarrow{p} f$
- ◊  $f^v \xrightarrow{u} f \Rightarrow f^v \xrightarrow{h} f$



- ◊  $f^v \xrightarrow{hypo} f, x^v \in \arg \max_{D^v} f^v, x^{v_k} \rightarrow \bar{x} \in D \Rightarrow \bar{x} \in \arg \max_D f$
- ◊  $\bar{x} \in \arg \max_D f \Rightarrow \exists \varepsilon^v \searrow 0, x^v \in \varepsilon^v - \arg \max_{D^v} f^v : x^v \rightarrow \bar{x}$
- ◊  $f^v \xrightarrow{h} f \Leftrightarrow dl(\text{hypo } f^v, \text{hypo } f) \rightarrow 0$   
( $\Gamma$ -convergence)

# Tight Hypo-Convergence

◆  $f_{D^\nu} \xrightarrow{h-tightly} f_D : f_{D^\nu} \xrightarrow{h} f_D$  &  $\forall \varepsilon > 0, \exists B \text{ compact} :$

$$\forall \nu > \nu_\varepsilon : \sup_{B \cap D^\nu} f^\nu \leq \sup_{D^\nu} f^\nu - \varepsilon$$

◆ THM:  $f_{D^\nu} \xrightarrow{h-tightly} f_D, \sup_D f \in \mathbb{R} \Rightarrow \sup_{D^\nu} f^\nu \rightarrow \sup_D f$

also:  $x^\nu \in \arg \max f_{D^\nu}, x^{\nu_k} \rightarrow \bar{x} \in D \Rightarrow \bar{x} \in \arg \max f_D$

# Lopsided convergence

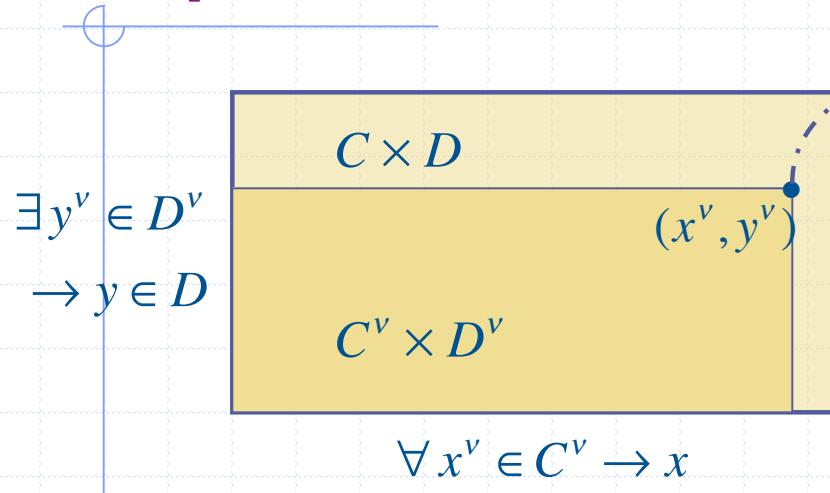
## max-inf framework

$$K : C \times D \rightarrow \mathbb{R}, \left\{ K^\nu : C^\nu \times D^\nu \rightarrow \mathbb{R} \right\}_{\nu \in \mathbb{N}}$$

$$\arg \max\text{-inf } K^a \simeq \arg \max\text{-inf } K$$

$$\bar{x} \in \arg \max\text{-inf } K \Leftrightarrow \bar{x} \in \arg \max(\inf_y K(x, y))$$

# Lopsided convergence: definition

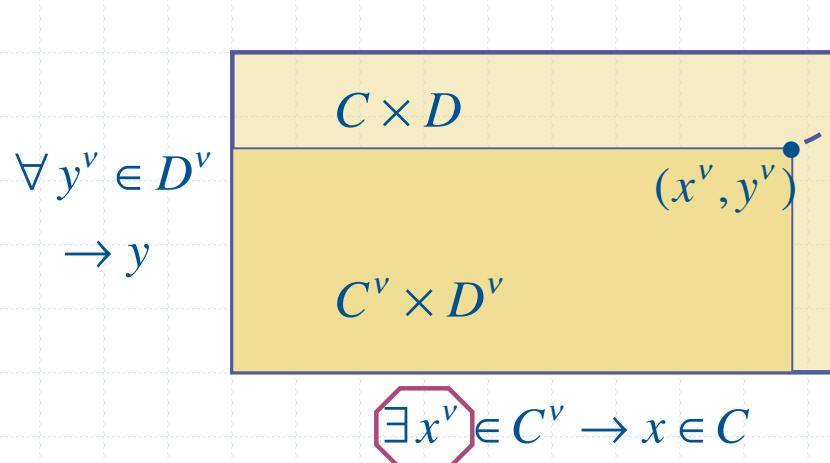


$(x, y)$

$\limsup_\nu K^\nu(x^\nu, y^\nu) \leq K(x, y)$  when  $x \in C$

$K^\nu(x^\nu, y^\nu) \rightarrow -\infty$  when  $x \notin C$

like hypo/epi convergence



$(x, y)$

$\liminf_\nu K^\nu(x^\nu, y^\nu) \geq K(x, y)$  when  $y \in D$

$K^\nu(x^\nu, y^\nu) \rightarrow \infty$  when  $x \notin D$

unlike hypo/epi convergence

# Lopsided convergence

$$\text{argmax-inf } K^a \sim \text{argmax-inf } K$$

$$\bar{x} \in \arg \max\text{-inf } K \Leftrightarrow \bar{x} \in \arg \max(\inf_y K(x, y))$$

**definition:**  $K : C \times D \rightarrow \mathbb{R}$ ,  $\{K^\nu : C^\nu \times D^\nu \rightarrow \mathbb{R}\}_{\nu \in \mathbb{N}}$

(a)  $\forall x^\nu \in C^\nu \rightarrow x \in C, \forall y \in D \exists y^\nu \in D^\nu \rightarrow y :$

$$\limsup K^\nu(x^\nu, y^\nu) \leq K(x, y)$$

(a<sub>∞</sub>)  $\forall x^\nu \in C^\nu \rightarrow x \notin C, \forall y \exists y^\nu \in D^\nu \rightarrow y : K^\nu(x^\nu, y^\nu) \rightarrow -\infty$

(b)  $\forall x \in C, \exists x^\nu \rightarrow x, \forall y^\nu \in D^\nu \text{ and } y^\nu \rightarrow y :$

$$\limsup K^\nu(x^\nu, y^\nu) \leq K(x, y) \text{ if } y \in D$$

$$K^\nu(x^\nu, y^\nu) \rightarrow \infty \text{ if } y \notin D$$

# Lopsided: elementary properties

$$\operatorname{argmax}\inf K^a \sim \operatorname{argmax}\inf K$$

lopsided  $\Rightarrow$  epi-convergence if  $K(x, y) = g(y)$

lopsided  $\Rightarrow$  hypo-convergence if  $K(x, y) = f(x)$

Remarks: not a  $\Gamma$ -convergence,  
in  $\infty$ -dim. different topologies

# Ancillary tight lop-convergence



$K_{C^\nu \times D^\nu}^\nu \xrightarrow{\text{ancil.-tight}} K_{C \times D}$  if  $K_{C^\nu \times D^\nu}^\nu \xrightarrow{\text{lop}} K_{C \times D}$  &

(b)  $\forall x \in C, \exists x^\nu \rightarrow x, \forall y^\nu \in D^\nu$  and  $y^\nu \rightarrow y$ :

$$\liminf K^\nu(x^\nu, y^\nu) \geq K(x, y) \text{ if } y \in D$$

$$K^\nu(x^\nu, y^\nu) \rightarrow \infty \text{ if } y \notin D$$

but also  $\forall \varepsilon > 0$ ,  $\exists B_\varepsilon$  compact (depends on  $x^\nu \rightarrow x$ ):

$$\inf_{B_\varepsilon \cap D^\nu} K^\nu(x^\nu, \cdot) \leq \inf_{D^\nu} K^\nu(x^\nu, \cdot) + \varepsilon, \quad \forall \nu \geq \nu_\varepsilon$$



THM.  $K_{C^\nu \times D^\nu}^\nu \xrightarrow[\text{lop}]{} K_{C \times D}$  ancillary tight,  $\bar{x}$  cluster point of

$$\{x^\nu \in \arg \max - \inf K_{C^\nu \times D^\nu}^\nu\}_{\nu \in \mathbb{N}} \Rightarrow \bar{x} \in \arg \max - \inf K_{C \times D}$$

# Proof ....



$K_{C^\nu \times D^\nu}^\nu \xrightarrow{lop} K_{C \times D}$  ancillary tight

Let  $g^\nu = \inf_{y \in D^\nu} K^\nu(\bullet, y)$ ,  $g = \inf_{y \in D} K(\bullet, y)$ .

$\Rightarrow g^\nu \xrightarrow{hypo} g$  when  $\begin{cases} C_g^\nu = \{x \in C^\nu \mid g^\nu(x) > -\infty\} \\ C_g = \{x \in C \mid g(x) > -\infty\} \end{cases} \neq \emptyset$



then apply

$g_{C^\nu}^\nu \xrightarrow{hypo} g_C$ ,  $x^\nu \in \arg \max_{C^\nu} g^\nu$ ,  $x^{\nu_k} \rightarrow \bar{x} \in C \Rightarrow \bar{x} \in \arg \max_C g$

# Ky Fan functions & Inequality

◆  $K : C \times C \rightarrow \mathbb{R}$  Ky Fan function if

(a)  $\forall y: x \mapsto K(x, y)$  usc

(b)  $\forall x: y \mapsto K(x, y)$  convex

◆  $K$  Ky Fan fcn,  $\text{dom } K = C \times C$ ,  $C$  compact

$\Rightarrow \arg \max -\inf K \neq \emptyset$

if  $K(x, x) \geq 0$  on  $\text{dom } K$ ,  $\bar{x} \in \arg \max -\inf K$

$\Rightarrow \inf_y K(\bar{x}, y) \geq 0.$

# Extending Ky Fan's inequality



$K^v \xrightarrow[\text{lop}]{} K$  ancillary tight with  $C^v \rightarrow C$ ,



$K^v$  Ky Fan  $\Rightarrow K$  Ky Fan

&  $\forall v : \arg \max\text{-}\inf K^v \neq \emptyset$

if  $\bar{x} \in \text{cluster-pts } \{\arg \max\text{-}\inf K^v\}$

$\Rightarrow \bar{x} \in \arg \max\text{-}\inf K$  &  $K(\bar{x}, \cdot) \geq 0$



Ky Fan fcns closed under lopsided

saddle fcns closed under h/e-convergence

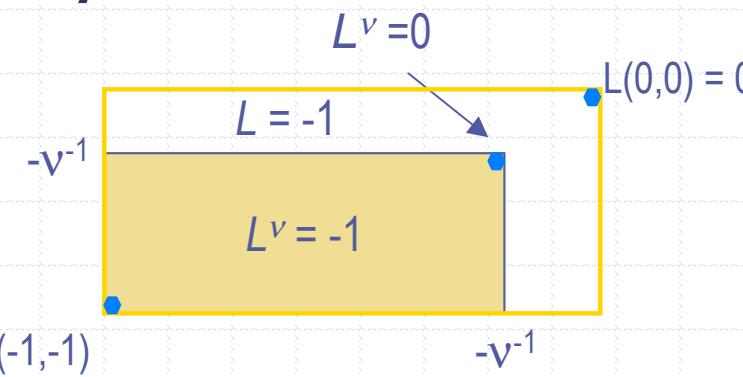
usc fcns closed under hypo-convergence

# Lopsided & hypo/epi-convergence

- lopsided  $\rightarrow$  hypo/epi-convergence
- not conversely

$\forall x^\nu \rightarrow 0,$

$\exists y^\nu \rightarrow 0 \dots$   
hypo/epi



$\exists x^\nu \rightarrow 0,$

$\forall y^\nu \rightarrow 0 \dots$   
lopsided

$$(x, y) = (0, 0), y^\nu = -v^{-1} \Rightarrow x^\nu = -v^{-1} : \liminf K^\nu(x^\nu, y^\nu) \geq 0$$

but with  $\bar{y}^\nu = -v^{-1} / 2 : \liminf K^\nu(x^\nu, \bar{y}^\nu) = -1 < 0 !$

# Lopsided & hypo/epi-convergence



$$L^v \xrightarrow[h/e]{} L, \text{ convex-concave} \implies L^v \xrightarrow[\text{lop}]{} L$$

first condition ... identical

for the second condition: given  $(x, y)$ , let

$$u \in \partial_x L(x, y), -v \in \partial_y (-L(x, y))$$

$$L^v \xrightarrow[h/e]{} L \implies L^v - \langle u, \cdot \rangle + \langle v, \cdot \rangle \xrightarrow[h/e]{} L - \langle u, \cdot \rangle + \langle v, \cdot \rangle$$

$(x, y)$  saddle point of  $L - \langle u, \cdot \rangle + \langle v, \cdot \rangle$

$\exists (x^v, y^v)$  saddle-pts  $\rightarrow (x, y) \& \{x^v\}_{v \in \mathbb{N}}$  is 'lop-sequence'  $\rightarrow x$

# APPLICATIONS

# Fixed Points (Brouwer)

$C$  compact convex,  $G : C \rightarrow C$  continuous

find  $\bar{x} \in C$ :  $G(\bar{x}) = \bar{x}$ , fixed point

set  $K(x, y) = \langle G(x) - x, y \rangle$

Ky Fan Inequality applies (usc, convex,  $K(x, x) \geq 0$ ):

$\exists \bar{x}$  such that  $\langle G(\bar{x}) - \bar{x}, y \rangle \leq 0, \forall y \in C$

$\Rightarrow \bar{x}$  is a fixed point of  $G$

# Convergence of fixed points

$G^\nu : C^\nu \rightarrow C^\nu$  continuous,  $C^\nu$  convex

$C^\nu \rightarrow C \Rightarrow C^\nu$  compact,  $G^\nu \xrightarrow{cont} G$

i.e.,  $\forall x^\nu \rightarrow x : G^\nu(x^\nu) \rightarrow G(x)$

$\Rightarrow K^\nu \xrightarrow{lop} K, K(x, y) = \langle G(x) - x, y \rangle$

Hence  $\exists x^\nu$  fixed points of  $G^\nu$  on  $C^\nu$

$\rightarrow \bar{x}$  fixed point of  $G$  on  $C$

Extension:  $C$  just convex

# Set-valued Inclusions

- ◆  $S^\nu, S : \mathbb{R}^d \rightarrow cl/cvx\text{-sets}(\mathbb{R}^n)$ , non-empty
- ◆ Inclusions:  $S^\nu(u) \ni a^\nu, S(u) \ni 0, a^\nu \rightarrow 0$
- ◆  $\sigma^\nu(\cdot, u) = \text{supp. fcn } S^\nu(u), \sigma(\cdot, u)$  of  $S(u)$
- ◆  $\forall u : \sigma^\nu(\cdot, u), \sigma(\cdot, u)$  convex
- ◆  $S^\nu, S$  continuous, compact-valued (sufficient)
- ◆  $\Leftrightarrow \sigma^\nu(\cdot, u^k) \rightarrow_{epi} \sigma^\nu(\cdot, \bar{u})$  as  $u^k \rightarrow \bar{u}$ ; also  $\sigma$
- ◆  $\Leftrightarrow \sigma^\nu(\cdot, u^k) \rightarrow_p \sigma^\nu(\cdot, \bar{u}) \Rightarrow \sigma^\nu(x, \cdot), \sigma(x, \cdot)$  usc
- ◆  $K^\nu = \sigma^\nu - \langle a^\nu, \cdot \rangle, K = \sigma \because$  Ky Fan fcns

# Variational Inequalities

- ◆  $C \subset \mathbb{R}^n$  nonempty, convex, compact
- ◆  $G : C \rightarrow \mathbb{R}^m$  continuous, ( $m=n$ )
- ◆ find  $\bar{u} \in C$  such that  $-G(\bar{u}) \in N_C(\bar{u})$   
where  $v \in N_C(\bar{u}) \Leftrightarrow \langle v, u - \bar{u} \rangle \leq 0, \forall u \in C$
- ◆ with  $K(u, v) = \langle G(u), v - u \rangle$  on  $\text{dom } K = C \times C$   
 $\Rightarrow K$  is a Ky Fan function,  $K(u, u) \geq 0$ .

Find

$$\bar{u} \in \arg \max -\inf K(\cdot, \cdot) \text{ so that } K(\bar{u}, \cdot) \geq 0$$

# Convergence: V.I. + extension

$C^\nu \rightarrow C \Rightarrow C^\nu$  compact  $\nu \geq \bar{\nu}$ ,  $G^\nu$  continuous

$G^\nu \xrightarrow{cont} G: G^\nu(x^\nu) \rightarrow G(x), \forall x^\nu \in C^\nu \rightarrow x$

$K^\nu(u, v) = \langle G^\nu(u), v - u \rangle$  on  $\text{dom } K^\nu = C^\nu \times C^\nu$

i.e., Ky Fan functions

- ◆ lop-converge ancillary tight to  $K$ , i.e.  
any cluster point of the solutions of the  
approximating V.I. is sol'n of limit V.I.

(sufficient conditions)

# Nash Equilibrium points

◆  $\bar{x}_a \in \arg \max u_a(x_a, \bar{x}_{-a}), \forall a \in \mathcal{A}$

◆ Nash function:

$$N(x, y) = \sum_{a \in \mathcal{A}} u_a(x_a, x_{-a}) - \sum_{a \in \mathcal{A}} u_a(y_a, x_{-a})$$

◆  $u_a$  usc &  $u_a(x_a, \cdot)$  lsc;  $u_a(\cdot, x_{-a})$  concave

$\Rightarrow N$  a Ky Fan function &  $N(x, x) \geq 0$

◆  $u_a^v \rightarrow_{cont} u_a$  &  $\text{dom } u_a^v \rightarrow \text{dom } u_a$  compact

$\Rightarrow N^v \rightarrow N$  ancillary tight

$\Rightarrow$  Nash equilibr.<sup>v</sup>  $\rightarrow$  Nash equilibr. (cluster)

# Walras Equilibrium points

- ◆  $\forall a \in \mathcal{A}: d_a(p) = \arg \max \left\{ u_a(x_a) \mid \langle p, x_a \rangle \leq \langle p, e_a \rangle \right\}$
- ◆  $s(p) = \sum_a (e_a - d_a(p))$  excess supply
- ◆ find  $\bar{p} \in \Delta$  (unit simplex) so that  $s(\bar{p}) \geq 0$
- ◆ **Walrasian:**  $W(p, q) = \langle q, s(p) \rangle$  Ky Fan fcn
- ◆  $\bar{p} \in \arg \max\inf W \Leftrightarrow s(\bar{p}) \geq 0$
- ◆ conditions:  $e_a \in \text{int dom } u_a$ , "globally compact"
- ◆ **Convergence:**  $u_a^\nu \xrightarrow{\text{hypo}} u_a, e_a^\nu \rightarrow e_a \Rightarrow W^\nu$  lop-converge ancillary tight to  $W$

