

# Convex Choice

Navin Kartik

Andreas Kleiner



# Introduction (1)

Agent with utility  $u(a, \theta)$ ,  $a \in A$  and  $\theta \in \Theta \subset \mathbb{R}^n$

Important result in 1-dim signaling & mech design

→ IC reduces to **local IC** under **single-crossing property** ( $\equiv$  “interval choice”)

How to extend to **multi-dim types**?

This paper: **convex choice**

→ from any choice set, any action is chosen by a convex set of types

Natural requirement; useful even beyond local IC

## Introduction (2)

Main results:

- ① Sense in which convex choice characterizes sufficiency of local IC
- ② Other applications: implementability; cheap talk
- ③ Convex choice  $\iff$  “directional single crossing”
- ④ For EU on lotteries, convex choice  $\implies$  “one-dim or affine” representation
$$u(a, \theta) = v(a) \cdot \theta + w(a)$$

This affine form has been salient in multi-dim studies

## Related Literature

Convex choice: [Grandmont 1978](#)

Interval choice and lotteries: [Kartik, Lee, Rappoport 2024](#)

Multi-dim single crossing: [McAfee & McMillan 1988](#); [Milgrom & Shannon 1994](#)

### Applications

- Sufficiency of local IC: [Carroll 2012](#)
- Implementability: [Saks & Yu 2005](#); [BCLMNS 2006](#)
- Cheap talk: [Levy & Razin 2004](#); [Sobel 2016](#)

# Convex Choice and Applications

# Convex Choice

Agent with utility  $u(a, \theta)$ ,  $a \in A$  and  $\theta \in \Theta \subset \mathbb{R}^n$ ,  $\Theta$  convex

## Definition

$u$  has **convex choice** if  $\forall B \subset A$  and  $\forall a \in B$ ,

$$\left\{ \theta : \{a\} = \operatorname{argmax}_{b \in B} u(b, \theta) \right\} \text{ is convex.}$$

(Enough to only consider all binary choice sets)

- Grandmont's 1978 "betweenness"
- In 1-dim, "interval choice" of Kartik, Lee, Rappoport 2024

# Convex Choice

Agent with utility  $u(a, \theta)$ ,  $a \in A$  and  $\theta \in \Theta \subset \mathbb{R}^n$ ,  $\Theta$  convex

## Definition

$u$  has **convex choice** if  $\forall B \subset A$  and  $\forall a \in B$ ,

$$\left\{ \theta : \{a\} = \operatorname{argmax}_{b \in B} u(b, \theta) \right\} \text{ is convex.}$$

(Enough to only consider all binary choice sets)

For talk, maintain “**regular**” **indifferences**:

$$[u(a', \theta') > u(a'', \theta') \text{ and } u(a', \theta'') = u(a'', \theta'')] \implies u(a', \theta) > u(a'', \theta) \quad \forall \theta \in (\theta', \theta'').$$

Satisfied, e.g., by **no indifferences** or by  $A \subset \mathbb{R}^n$  and  $u(a, \theta) = a \cdot \theta$

(Paper uses a weaker version, and selectively.)

# Incentive Compatibility

$N_\theta \subset \Theta$  denotes open neighborhood of  $\theta$  (in relative topology)

Direct mechanisms  $\Theta \rightarrow A$  (subsumes stochastic mechs)

## Definition

Mechanism  $m : \Theta \rightarrow A$  is

- **incentive compatible** (IC) if  $\forall \theta \in \Theta$ ,

$$\forall \theta' \in \Theta : u(m(\theta), \theta) \geq u(m(\theta'), \theta) \text{ and } u(m(\theta'), \theta') \geq u(m(\theta), \theta').$$

- **locally IC** if  $\forall \theta \in \Theta$ ,  $\exists N_\theta \subset \Theta$  s.t.

$$\forall \theta' \in N_\theta : u(m(\theta), \theta) \geq u(m(\theta'), \theta) \text{ and } u(m(\theta'), \theta') \geq u(m(\theta), \theta').$$

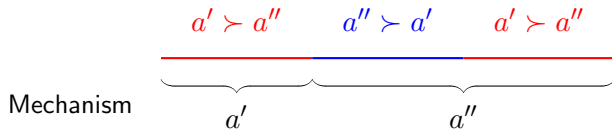
Analogously for a mechanism defined on  $\Theta' \subset \Theta$

(Elaborate)



## Sufficiency of Local IC (1)

Local IC does not generally imply IC:



Not convex choice!

## Sufficiency of Local IC (2)

### Proposition

$u$  has convex choice  $\implies$  if  $m : \Theta \rightarrow A$  is locally IC then it is IC.



for any **line segment**  $\ell \subset \Theta$  and any mechanism  $m : \ell \rightarrow A$ , if  $m$  is locally IC then it is IC.

- So IC between  $\theta$  and  $\theta'$  requires only checking local IC along line segment  $(\theta, \theta')$
- Such “integration up” is a common strategy
- Corollary: sufficiency of local IC on full type space  $\Theta$

## Sufficiency of Local IC (2)

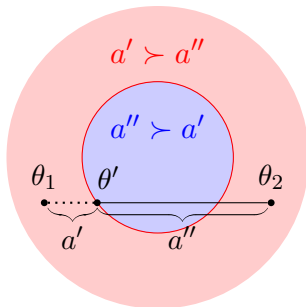
### Proposition

$u$  has convex choice  $\implies$  if  $m : \Theta \rightarrow A$  is locally IC then it is IC.



for any **line segment**  $\ell \subset \Theta$  and any mechanism  $m : \ell \rightarrow A$ , if  $m$  is locally IC then it is IC.

- Sufficiency of local IC on  $\Theta \not\Rightarrow$  CC
- But sufficiency on all line segments does
- All line segments of interest because:
  - 1 Reduces checking IC between any two types to a 1-d task, which is 'tractable'
  - 2 A tractable problem must remain tractable when restricted to lower dimensions



## Sufficiency of Local IC (2)

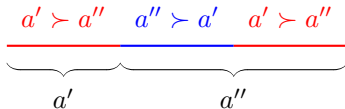
### Proposition

$u$  has convex choice  $\implies$  if  $m : \Theta \rightarrow A$  is locally IC then it is IC.



for any **line segment**  $\ell \subset \Theta$  and any mechanism  $m : \ell \rightarrow A$ , if  $m$  is locally IC then it is IC.

Proof idea: **Necessity** of CC captured by earlier 1-dim example



## Sufficiency of Local IC (2)

### Proposition

$u$  has convex choice  $\implies$  if  $m : \Theta \rightarrow A$  is locally IC then it is IC.



for any **line segment**  $\ell \subset \Theta$  and any mechanism  $m : \ell \rightarrow A$ , if  $m$  is locally IC then it is IC.

Proof idea: Heuristic for **sufficiency**

Assume no indiff, take **any**  $\theta, \theta'$  and a fine grid on their line segment,  $\theta = \theta_1, \dots, \theta_n = \theta'$

■ local IC  $\implies u(m(\theta_i), \theta_i) > u(m(\theta_{i+1}), \theta_i) \quad \forall i = 1, 2$   
 $u(m(\theta_3), \theta_3) > u(m(\theta_2), \theta_3)$

■ convex choice  $\implies u(m(\theta_1), \theta_1) > u(m(\theta_3), \theta_1)$

■ iterate logic, using local IC and CC each time, to get  $u(m(\theta_1), \theta_1) > u(m(\theta_1), \theta_n)$ .

## Sufficiency of Local IC (2)

### Proposition

$u$  has convex choice  $\implies$  if  $m : \Theta \rightarrow A$  is locally IC then it is IC.



for any **line segment**  $\ell \subset \Theta$  and any mechanism  $m : \ell \rightarrow A$ , if  $m$  is locally IC then it is IC.

Carroll 2012 establishes sufficiency of local IC using “**domain representation**” of prefs

Our **parameter representation** approach is complementary

Formally, his result is subsumed by  $A \subset \mathbb{R}^n$  and  $u(a, \theta) = a \cdot \theta$

► Implementability

# Cheap Talk

In cheap talk or costly signaling,

sender's utility having convex choice  $\implies$  every eqm is “convex partitional”  
(modulo details about indifferences)

Has been interest in extending Crawford & Sobel 1982 to multiple dims

- Levy & Razin 2004, 2007; Chakraborty & Harbaugh 2007

Also common-interest cheap talk with finite message space

- Jager, Metzger, Riedel 2011; Saint-Paul 2017; Sobel 2016; Bauch 2024

## Remark

Assume  $A \subset \mathbb{R}^n$  and  $u(a, \theta) = -l(\|a - \theta\|)$ , with  $l(\cdot)$  strictly  $\uparrow$ .

(and  $A \cap \Theta$  has nonempty interior)

Convex choice  $\iff$  norm is weighted Euclidean

(i.e.,  $\|x\| = \sqrt{xWx^T}$ , with  $W$  sym pos def)

## Directional Single Crossing

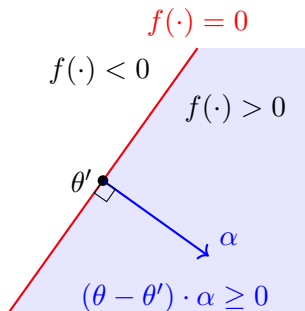


# Directional Single Crossing (1)

Convex choice can be viewed as single crossing

## Definition

$f : \Theta \rightarrow \mathbb{R}$  is **directionally single crossing** if  $\exists \alpha \in \mathbb{R}^n \setminus \{0\}$  s.t.  $\forall \theta, \theta' \in \Theta$ ,  
 $(\theta - \theta') \cdot \alpha \geq 0 \implies \text{sign}(f(\theta)) \geq \text{sign}(f(\theta'))$ .

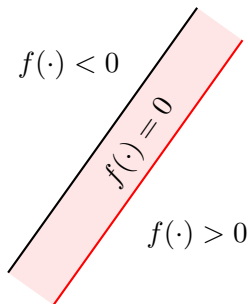


# Directional Single Crossing (1)

Convex choice can be viewed as single crossing

## Definition

$f : \Theta \rightarrow \mathbb{R}$  is **directionally single crossing** if  $\exists \alpha \in \mathbb{R}^n \setminus \{0\}$  s.t.  $\forall \theta, \theta' \in \Theta$ ,  
 $(\theta - \theta') \cdot \alpha \geq 0 \implies \text{sign}(f(\theta)) \geq \text{sign}(f(\theta'))$ .

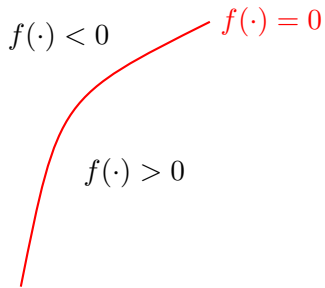


# Directional Single Crossing (1)

Convex choice can be viewed as single crossing

## Definition

$f : \Theta \rightarrow \mathbb{R}$  is **directionally single crossing** if  $\exists \alpha \in \mathbb{R}^n \setminus \{0\}$  s.t.  $\forall \theta, \theta' \in \Theta$ ,  
 $(\theta - \theta') \cdot \alpha \geq 0 \implies \text{sign}(f(\theta)) \geq \text{sign}(f(\theta'))$ .



Not DSC

## Directional Single Crossing (2)

Convex choice can be viewed as single crossing

### Definition

$u : A \times \Theta \rightarrow \mathbb{R}$  has **directionally single-crossing differences** if  $\forall a, a' \in A$ ,

$u(a, \theta) - u(a', \theta)$  is directionally single crossing.

- $\forall a, a'$ , strict preference sets are parallel half-spaces, either open or closed  
(intersected with the type space)
- Direction of defining hyperplanes can vary across action pairs

## Directional Single Crossing (2)

Convex choice can be viewed as single crossing

### Definition

$u : A \times \Theta \rightarrow \mathbb{R}$  has **directionally single-crossing differences** if  $\forall a, a' \in A$ ,

$u(a, \theta) - u(a', \theta)$  is directionally single crossing.

Leading example families, when  $A \subset \mathbb{R}^n$ :

① **weighted Euclidean**: any  $\downarrow$  fn of  $(a - \theta)W(a - \theta)^T$ , with  $W$  sym pos def

② **CES**:  $A, \Theta \subset \mathbb{R}_+^n$  and  $u(a, \theta) = (\sum_{i=1}^n (a_i)^r \theta_i)^s$  with  $r \in \mathbb{R}$  and  $s > 0$

For these families, adding a type-independent function preserves DSCD, so, e.g.,

$u(a, \theta) = a \cdot \theta + w(a)$  has DSCD

## Directional Single Crossing (3)

Convex choice can be viewed as single crossing

### Proposition

If  $u$  has DSCD, then  $u$  has convex choice.

If  $u$  “strictly violates” DSCD, then  $u$  does not have convex choice.

- 1st statement straightforward from geometry
- 2nd follows from a sep hyp thm
- Closely related to Grandmont 1978; his form is more restrictive (e.g., continuity)

# Convex Environments

# Convex Environments (1)

Choice among lotteries with EU:  $A \equiv \Delta X$  and  $u(a, \theta) \equiv \sum_x a(x) \bar{u}(x, \theta)$

- stochastic or multiple-agent mechanisms
- cheap talk where sender is uncertain about receiver prefs

More generally, **convex environment**:  $\{u(a, \cdot) : \Theta \rightarrow \mathbb{R}\}_{a \in A}$  is convex

- rank-dependent EU / prob distortion, where distortion function has convex image
- choice over  $T$ -period consumption streams:

$$A \equiv [\underline{a}, \bar{a}]^T \text{ and } u(a, \theta) \equiv \sum_t v(a_t) \rho(t; \theta), \text{ with } v(\cdot) \text{ continuous}$$



## Convex Environments (2)

### Proposition

Assume  $\Theta = \mathbb{R}^n$ ,  $u(a, \theta)$  is differentiable in  $\theta$ , and no type is totally indifferent.

Convex environment and DSCD  $\implies u$  is 1-dimensional or has affine representation.

- 1-dimensional if  $\exists \alpha \in \mathbb{R}^n \setminus \{0\}$  and  $\tilde{u} : A \times \mathbb{R} \rightarrow \mathbb{R}$  s.t.  
 $\tilde{u}(a, \alpha \cdot \theta)$  represents the same prefs for every  $\theta$
- Affine representation if  $\exists v : A \rightarrow \mathbb{R}^n$  and  $w : A \rightarrow \mathbb{R}$  s.t.  
 $v(a) \cdot \theta + w(a)$  represents the same prefs for every  $\theta$

## Convex Environments (2)

### Proposition

Assume  $\Theta = \mathbb{R}^n$ ,  $u(a, \theta)$  is differentiable in  $\theta$ , and no type is totally indifferent.

Convex environment and DSCD  $\implies u$  is 1-dimensional or has affine representation.

- Consider CES prefs:  $X, \Theta \subset \mathbb{R}_+^n$  (with nonempty interiors) and

$$\bar{u}(x, \theta) = \left( \sum_{i=1}^n (x_i)^r \theta_i \right)^s + w(x) \text{ with } r \in \mathbb{R} \text{ and } s > 0.$$

- Although  $\bar{u}$  satisfies DSCD, does the induced EU over  $A = \Delta X$ ?
- If  $n = 1$ , yes. But when  $n > 1$ , if and only if  $s = 1$ .

## Convex Environments (2)

### Proposition

Assume  $\Theta = \mathbb{R}^n$ ,  $u(a, \theta)$  is differentiable in  $\theta$ , and no type is totally indifferent.

Convex environment and DSCD  $\implies u$  is 1-dimensional or has affine representation.

Conclusion also holds under alternate assumptions

- Prop 5: quasi-linear, differentiable in type, and minimally rich (drop  $\Theta = \mathbb{R}^n$ )

Interpretation:

- In rich environments, genuinely multi-dim prefs are unwieldy unless affine
- New perspective on why multi-dim mech design has emphasized affine form

Our exercise only allows changing representation; not redefining types

# Conclusion

Convex choice is a valuable property

- characterizes sufficiency of local IC (on all line segments)
- other applications: implementability; cheap talk
- essentially equiv to a form of single crossing with simple geometric interpretation
- in convex envs with some regularity, “one-dimensional or affine representation”

(Others: preference aggregation; social learning)

Another interesting notion: connected choice

- also relevant for sufficiency of local IC (on full type space)
- we view convex choice as more appealing

Thank you!

# On Local IC Definition

(Back)

## Definition

Mechanism  $m : \Theta \rightarrow A$  is locally IC if  $\forall \theta \in \Theta, \exists N_\theta \subset \Theta$  s.t.

$$\forall \theta' \in N_\theta : u(m(\theta), \theta) \geq u(m(\theta'), \theta) \quad \text{and} \quad u(m(\theta'), \theta') \geq u(m(\theta), \theta').$$

## Example

$$\text{Mechanism } \frac{a'' \succ a'}{[ \text{-----} a' \text{-----} ] [ \text{-----} a'' \text{-----} ]}$$

Our defn is weaker than:  $\exists \varepsilon > 0$  s.t.  $\forall \theta \in \Theta, \exists B_\theta^\varepsilon$  s.t.

$$\forall \theta' \in B_\theta^\varepsilon \cap \Theta : u(m(\theta), \theta) \geq u(m(\theta'), \theta).$$

# Implementability

$A \equiv Y \times \mathbb{R}$ ; assume  $Y$  finite. Quasilinear prefs:  $u((y, t), \theta) \equiv \tilde{u}(y, \theta) - t$

(Back)

Allocation rule  $v : \Theta \rightarrow Y$  is implementable if  $\exists \tau : \Theta \rightarrow \mathbb{R}$  s.t.  $(v, \tau)$  is IC

Which allocation rules are implementable?

Necessary condition is weak (or 2-cycle) monotonicity:

$$\forall \theta, \theta' : \tilde{u}(v(\theta), \theta) - \tilde{u}(v(\theta'), \theta) \geq \tilde{u}(v(\theta), \theta') - \tilde{u}(v(\theta'), \theta')$$

(Rochet 1987: “cyclical monotonicity” is nec & suff)

Saks & Yu 2005: weak mon is suff if  $\Theta$  convex,  $Y \subset \mathbb{R}^n$ , and  $\tilde{u}(y, \theta) = y \cdot \theta$

## Proposition

Assume  $u$  has convex choice and is continuous in  $\theta$ .

Every weakly monotone allocation rule is implementable.

Proof uses result from

Berger, Müller, Naeemi 2017