# Dynamic Programming Chapter 2

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#### Introduction

#### Summary of this lecture:

- Introduction to partial orders
- Pointwise orders
- Order-preserving maps
- Fixed points and order
- Conjugate maps
- Convergence rates and gradient-based methods

## Order

The next few slides give a quick introduction to order theory

One of the foundational subjects of maths, on par with

- algebra
- geometry
- topology
- number theory
- set theory

#### But not commonly taught in foundational math courses

#### Why?

Rarely used in

- physics
- chemistry
- biology, etc.

Math courses are biased toward these subjects!

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# But very important for econ and related fields Examples.

- Does consumer X prefer good A or good B?
- Is welfare greater under policy A or policy B?
- Does R & D increase profits?
- How can firm Y minimize costs?

For these lectures, we need order for

- studying optimality
- fixed point results

#### Partial orders

Let P be a nonempty set

A **partial order** on a P is a binary relation  $\leq$  on  $P \times P$  satisfying, for any p,q,r in P,

$$p \leq p$$
,

$$p \leq q$$
 and  $q \leq p$  implies  $p = q$  and

$$p \leq q$$
 and  $q \leq r$  implies  $p \leq r$ 

(Reflexivity, antisymmetry, transitivity)

We call  $(P, \preceq)$  (or just P) a partially ordered set

#### Ex.

- 1. Show that the usual order  $\leqslant$  on  $\mathbb R$  is a partial order on  $\mathbb R$
- 2. Given set M, show that  $\subset$  is a partial order on  $\wp(M)$

Proof for 2: Clearly, for all  $A, B, C \subset M$ ,

- $A \subset A$  holds
- $A \subset B$  and  $B \subset A$  implies A = B
- $A \subset B$  and  $B \subset C$  implies  $A \subset C$

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A partial order  $\leq$  on P is called a **total order** if

either 
$$p \leq q$$
 or  $q \leq p$  for all  $p, q \in P$ 

Example.  $\leqslant$  is a total order on  $\mathbb R$ 

**Ex.** Prove:  $\subset$  is not a total order on  $\wp(M)$  when |M|>1

<u>Proof</u>: If M has more than two elements, then we can take nonempty  $A,B\subset M$  with  $A\cup B=\emptyset$ 

But then  $A \subset B$  and  $B \subset A$  both fail

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#### Pointwise Partial Orders

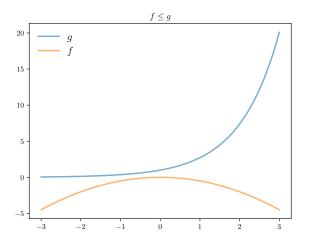
#### Let

- M be any set and
- let  $\mathbb{R}^M$  be all  $f \colon M \to \mathbb{R}$

The **pointwise partial order** over  $\mathbb{R}^M$  is writen as  $\leqslant$  and defined as follows:

• Given f,g in  $\mathbb{R}^M$ , we set

$$f \leqslant g \iff f(x) \leqslant g(x) \text{ for all } x \in M$$



## **Ex.** Show $\leqslant$ is a partial order on $\mathbb{R}^M$

#### Proof:

Let's just check antisymmetry

Fix  $f,g \in \mathbb{R}^M$  and suppose  $f \leqslant g$  and  $g \leqslant f$ 

Pick any  $x \in M$ 

By definition,  $f(x) \leqslant g(x)$  and  $g(x) \leqslant f(x)$ 

Therefore, f(x) = g(x)

Since x was arbitrary, we have f = g

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#### Let's define the pointwise partial order for matrices

Let  $\mathbb{M}^{n \times k} := \mathsf{all} \ n \times k \mathsf{matrices}$ 

For 
$$A=(a_{ij})$$
 and  $B=(b_{ij})$  in  $\mathbb{M}^{n\times k}$ , we set

$$A \leqslant B \iff a_{ij} \leqslant b_{ij} \text{ for all } i,j$$

Example.

$$\begin{pmatrix} 1 & 2 \\ -2 & 0 \end{pmatrix} \leqslant \begin{pmatrix} 10 & 20 \\ 0 & 10 \end{pmatrix}$$

**Ex.** Show that  $\leq$  is a partial order on  $\mathbb{M}^{n \times k}$ 

#### Special case: pointwise order for vectors

Recall 
$$[n] := \{1, ..., n\}$$

For 
$$x=(x_1,\ldots,x_n)$$
 and  $y=(y_1,\ldots,y_n)$  in  $\mathbb{R}^n$ , we write

$$x \leqslant y \iff x_i \leqslant y_i \text{ for all } i \in [n]$$

## Pointwise partial order $\leq$ on $\mathbb{R}^2$ :

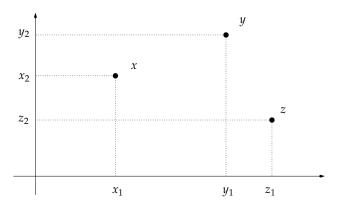


Figure:  $x \le y$  but neither x, z nor y, z are comparable

**Ex.** Prove: for  $a, b \in \mathbb{R}^n$  and sequence  $(x_k)$  in  $\mathbb{R}^n$ , we have

$$a \leqslant x_k \leqslant b$$
 for all  $k \in \mathbb{N}$  and  $x_k \to x$  implies  $a \leqslant x \leqslant b$ 

Proof: Fix  $i \in [n]$ 

Let  $a^i$  be the *i*-th element of a, etc.

It suffices to show that

$$a^i \leqslant x^i \leqslant b^i \tag{1}$$

Note  $x_k \to x$  implies  $x_k^i \to x^i$ 

Moreover,  $a^i \leqslant x^i_k \leqslant b^i$  for all k

Weak inequalities in  $\mathbb{R}$  are preserved under limits, so (1) holds

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Weak inequalities in  $\mathbb R$  are preserved under limits, so (1) holds

In other words, the pointwise partial order  $\leqslant$  is preserved under limits

As a result, these sets are closed

- $\bullet \ \mathbb{R}^n_+ := \{ x \in \mathbb{R}^n : 0 \leqslant x \}$
- $[a,b] := \{x \in \mathbb{R}^n : a \leqslant x \leqslant b\}$
- etc.

A key connection between order and topology!

#### **Ex.** Prove: If B is $m \times k$ and $B \geqslant 0$ , then

$$|Bx| \leq B|x|$$
 for all  $k \times 1$  column vectors  $x$ 

<u>Proof</u>: Fix  $B \in \mathbb{M}^{m \times k}$  with  $b_{ij} \geqslant 0$  for all i, j

Fix  $i \in [m]$  and  $x \in \mathbb{R}^k$ 

By the triangle inequality, we have  $|\sum_j b_{ij} x_j| \leqslant \sum_j b_{ij} |x_j|$ 

Stacking these inequalities yields

$$|Bx| \leqslant B|x|$$

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**Lemma.** Given a finite set M and f,g in  $\mathbb{R}^M$ , we have

$$|\max_{x \in M} f(x) - \max_{x \in M} g(x)| \leqslant \max_{x \in M} |f(x) - g(x)|$$

Proof: Fixing  $f,g \in \mathbb{R}^M$ , we have

$$f = f - g + g \le |f - g| + g$$
 (pointwise)

$$\therefore \max f \leqslant \max(|f - g| + g) \leqslant \max|f - g| + \max g$$

$$\therefore \max f - \max g \leqslant \max |f - g|$$

Reversing the roles of f and g proves the claim

# Order-preserving maps

#### Let

- $(P, \preceq)$  and  $(Q, \preceq)$  be partially ordered sets
- $T: P \to Q$

T is called **order-preserving** if, for all  $x, y \in P$ ,

$$x \leq y \implies Tx \leq Ty$$

- Meaning: If x goes up then Tx goes up
- Very important concept for dynamic programming

Example. Let  $(P, \preceq) = (\mathcal{C}, \leqslant)$  where

- $\mathcal C$  is all continuous functions from [a,b] to  $\mathbb R$
- ullet  $\leqslant$  is the pointwise partial order

If  $I \colon \mathfrak{C} \to \mathbb{R}$  is defined by

$$Ig := \int_{a}^{b} g(x)dx \qquad (g \in \mathcal{C})$$

then I is order-preserving on  ${\mathcal C}$ 

(Larger functions have larger integrals)

## Example. Let $\leqslant$ denote the pointwise partial order on $\mathbb{R}^n$

Let  $T: \mathbb{R}^n \to \mathbb{R}^n$  be defined by Tx = Ax + b

If  $A \geqslant 0$ , then T is order preserving on  $\mathbb{R}^n$ 

Proof: Fix  $x \leq y$ 

Then  $0 \leqslant y - x$ 

$$\therefore \quad 0 \leqslant A(y-x) \leqslant Ay - Ax$$

$$\therefore Ax \leqslant Ay$$

$$\therefore Tx \leqslant Ty$$

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# Special Case: Real-Valued Functions

Special case: maps from  $(P, \preceq)$  into  $(\mathbb{R}, \leqslant)$ 

Then "order-preserving" = "increasing"

In particular, we also call  $h \in \mathbb{R}^P$ 

- increasing if  $x \leq y$  implies  $h(x) \leqslant h(y)$  and
- **decreasing** if  $x \leq y$  implies  $h(x) \geqslant h(y)$

Let P be partially ordered by  $\leq$ 

We write  $i\mathbb{R}^P$  for the increasing functions in  $\mathbb{R}^P$ 

Thus,

$$h \in i\mathbb{R}^P \quad \iff \quad x,y \in P \text{ and } x \leq y \text{ implies } h(x) \leqslant h(y)$$

Example. Let  $P = \{1, ..., n\}$  and let  $\leq$  be the usual order  $\leq$  on  $\mathbb R$ 

Then

- $x \mapsto 2x$  and  $x \mapsto \mathbb{1}\{2 \leqslant x\}$  are in  $i\mathbb{R}^P$
- $x \mapsto -x$  and  $x \mapsto \mathbb{1}\{x \leqslant 2\}$  are not

#### **Ex.** Prove the following:

If  $f,g \in i\mathbb{R}^P$ , then

- $\alpha f + \beta g \in i\mathbb{R}^P$  when  $\alpha, \beta \geqslant 0$
- $f \lor g \in i\mathbb{R}^P$
- $f \wedge g \in i\mathbb{R}^P$

**Ex.** Given finite P, show that  $i\mathbb{R}^P$  is closed in  $\mathbb{R}^P$ 

<u>Proof</u>: Take  $(f_k)_{k\geqslant 1}$  in  $i\mathbb{R}^P$  and  $f\in\mathbb{R}^P$  with  $f_k\to f$ 

Since  $f_k \to f$  we have  $f_k(z) \to f(z)$  for all  $z \in P$ 

norm convergence implies pointwise convergence

Fix  $x, y \in P$  with  $x \leq y$ 

From  $(f_k) \subset i\mathbb{R}^P$  we have  $f_k(x) \leqslant f_k(y)$  for all k

Since weak inequalities are preserved under limits,  $f(x) \leqslant f(y)$ 

Hence  $f \in i\mathbb{R}^P$ 

## **Strict** inequalities

#### We write

- $f \ll g$  if f(x) < g(x) for all  $x \in$  some given set M
- $x \ll y$  if  $x_i < y_i$  for all  $i \in [n]$
- $A \ll B$  if  $a_{ij} < b_{ij}$  for all i, j

These are not partial orders

**Ex.** Why is  $f \ll g$  not a partial order on  $\mathbb{R}^M$ ?

## Parametric Monotonicity

Let  $(P, \preceq)$  be a partially ordered set

Given two self-maps S and T on P, we set

$$S \leq T \iff Sx \leq Tx \text{ for every } x \in P$$

We say that T dominates S on P

**Ex.** Show that  $\leq$  is a partial order on

$$S_P := P^P := \text{ set of all self-maps on } P$$

Proof of antisymmetry of  $\leq$  on  $S_P$ :

Let  $(P, \preceq)$  and  $S, T \in S_P$  be as defined above

Suppose  $S \leq T$  and  $T \leq S$ 

Fix any  $x \in P$ 

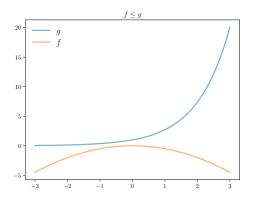
We have  $Sx \leq Tx$  and  $Tx \leq Sx$ 

Since  $\leq$  is antisymmetric on P, we have Sx = Tx

Since p was arbitrary, S = T

Hence  $\leq$  is antisymmetric on  $S_P$ 

Example. If  $(\preceq, P) = (\leqslant, \mathbb{R})$ , then  $\leqslant$  is the pointwise partial order over functions



# Example. Consider $\mathbb{R}^n_+$ with the pointwise partial order $\leqslant$

• Called the **positive cone** in  $\mathbb{R}^n$ 

#### Let

- Sx = Ax + b
- Tx = Bx + b

**Ex.** Show that  $0 \le A \le B \implies T$  dominates S on  $\mathbb{R}^n_+$ 

<u>Proof</u>: Fixing  $x \in \mathbb{R}^n_+$ , suffices to show that  $Sx \leqslant Tx$ 

Since  $A \leq B$  and  $x \geq 0$ , we have  $Ax \leq Bx$ 

Hence  $Sx \leq Tx$ 

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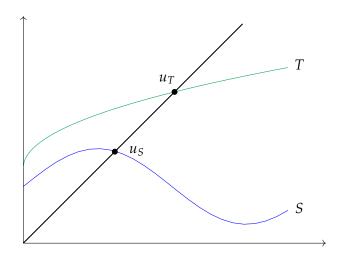
Conjecture: If  $S \leqslant T$ , then the fixed points of T will be larger

This is not true in general...

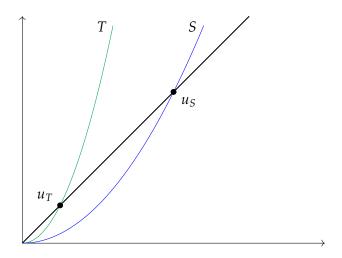
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This is <u>not</u> true in general...

#### Sometimes true:



## And sometimes false:



One difference: in the first case, T is globally stable

This leads us to our next result

### **Proposition.** Let

- ullet S and T be self-maps on  $M\subset \mathbb{R}^n$
- ullet  $\leqslant$  be the pointwise partial order on M

lf

- 1. T dominates S on M and
- 2. T is order-preserving and globally stable on M,

then the unique fixed point of T dominates any fixed point of S

#### Proof: Assume the conditions

#### Let

- $u_T$  be the unique fixed point of T and
- $u_S$  be any fixed point of S

Since  $S \leqslant T$ , we have  $u_S = Su_S \leqslant Tu_S$ 

Applying T to both sides of  $u_S \leqslant Tu_S$  gives

$$u_S \leqslant Tu_S \leqslant T^2u_S$$

Continuing in this fashion yields  $u_S \leqslant T^k u_S$  for all  $k \in \mathbb{N}$ Since  $\leqslant$  is preserved under limits and T is globally stable,

$$u_S \leqslant \lim_k T^k u_S = u_T$$

Example. Recall that, in the job search model,

$$h^* = c + \beta \sum_{w'} \max \left\{ \frac{w'}{1 - \beta'}, h^* \right\} \varphi(w')$$

We found  $h^*$  as the fixed point of  $g\colon \mathbb{R}_+ \to \mathbb{R}_+$  defined by

$$g(h) = c + \beta \sum_{w'} \max \left\{ \frac{w'}{1 - \beta'}, h \right\} \varphi(w')$$

In the exercise, you showed that g is a contraction map on  $\mathbb{R}_+$ 

# **Ex.** Prove that the optimal continuation value $h^*$ is increasing in $\beta$

Proof: Fix  $\beta_1 \leqslant \beta_2$  and let

- $h_i^* :=$  fixed point corresponding to  $\beta_i$
- $g_i := \text{fixed point map corresponding to } \beta_i$

Since  $\beta_1 \leqslant \beta_2$ , we have  $g_1(h) \leqslant g_2(h)$  for all  $h \in \mathbb{R}_+$ 

In addition, g2 is

- 1. a contraction (so globally stable) and
- 2. increasing (order-preserving)

Hence  $h_1^* \leqslant h_2^*$ 

**Ex.** Prove that the optimal continuation value  $h^*$  is increasing in  $\beta$ 

<u>Proof</u>: Fix  $\beta_1 \leqslant \beta_2$  and let

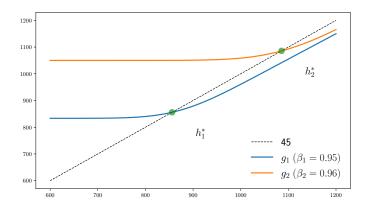
- $h_i^* := \mathsf{fixed} \; \mathsf{point} \; \mathsf{corresponding} \; \mathsf{to} \; \beta_i$
- $g_i := \text{fixed point map corresponding to } \beta_i$

Since  $\beta_1 \leqslant \beta_2$ , we have  $g_1(h) \leqslant g_2(h)$  for all  $h \in \mathbb{R}_+$ 

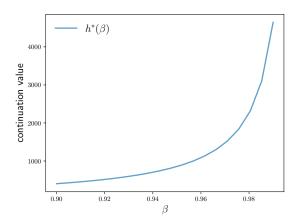
In addition,  $g_2$  is

- 1. a contraction (so globally stable) and
- 2. increasing (order-preserving)

Hence  $h_1^* \leqslant h_2^*$ 



# Ex. Replicate this figure



# (First Order) Stochastic Dominance

In the discussion above we obtained some results from order theory

• Example. parametric monotonicity

To use these results in a stochastic setting, we need to order distributions!

That is, we need a partial order over distributions

The most important of these partial orders is called "first order stochastic dominance"

In this section we define it

To start, let's consider ordering distributions in a special case

Example. The binomial distribution is defined as follows:

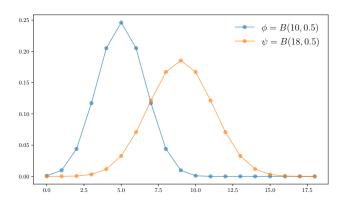
- $X \sim B(n, 0.5)$
- X counts the # of heads in n flips of a fair coin

Suppose  $\varphi \stackrel{d}{=} X \sim B(10,0.5)$  and  $\psi \stackrel{d}{=} Y \sim B(18,0.5)$ 

 Y counts over more flips, so it should be "larger" in some sense

Hence we expect that  $\varphi$  is " $\preceq$ "  $\psi$  in some sense

# Distribution $\psi$ seems "larger than" $\phi$ — more mass on higher draws



But how can we make this idea precise?

Let X be a finite set partially ordered by  $\leq$ 

Fix 
$$\varphi, \psi \in \mathfrak{D}(X)$$

Write  $\langle u, \varphi \rangle$  for  $\sum_{x} u(x) \varphi(x)$ , etc.

We say that  $\psi$  stochastically dominates  $\varphi$  and write  $\varphi \preceq_F \psi$  if

$$u \in i\mathbb{R}^{\mathsf{X}} \implies \langle u, \varphi \rangle \leqslant \langle u, \psi \rangle$$

# Example. If

- $\varphi \stackrel{d}{=} X \sim B(10, 0.5)$  and
- $\psi \stackrel{d}{=} Y \sim B(18, 0.5),$

then  $\varphi \preceq_{\mathrm{F}} \psi$ 

Proof: Fix  $u \in i\mathbb{R}^X$  and let

- $X = \{0, ..., 18\}$  and
- $W_1,\ldots,W_{18}$  be IID Bernoulli with  $\mathbb{P}\{W_i=1\}=0.5$  for all i

Then 
$$X:=\sum_{i=1}^{10}W_i\stackrel{d}{=}\varphi$$
 and  $Y:=\sum_{i=1}^{18}W_i\stackrel{d}{=}\psi$ 

Clearly  $X \leqslant Y$  with probability one (i.e., for any draw of  $\{W_i\}_{i=1}^{18}$ )

Hence 
$$u(X) \leqslant u(Y)$$

Hence 
$$\mathbb{E}u(X) \leqslant \mathbb{E}u(Y)$$

In other words,

$$\langle u, \varphi \rangle \leq \langle u, \psi \rangle$$

Example. An agent has preferences over outcomes in X

Preferences are determined by a utility function  $u \in \mathbb{R}^{X}$ 

The agent prefers more to less, so  $u \in i\mathbb{R}^X$ 

Suppose that the agent ranks lotteries over X according to expected utility

• evaluates  $\varphi \in \mathcal{D}(\mathsf{X})$  according to  $\sum_{x} u(x) \varphi(x)$ 

Then the agent (weakly) prefers  $\psi$  to  $\varphi$  whenever  $\varphi \preceq_F \psi$ 

# Another Perspective

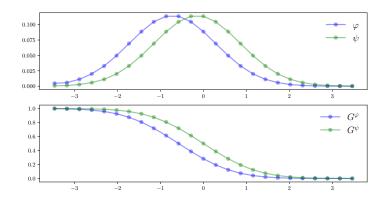
Given  $\varphi \in \mathfrak{D}(X)$ , let

$$G^{\varphi}(y) := \sum_{x \in X} \mathbb{1}\{y \le x\} \varphi(x) \qquad (y \in X)$$

This is the counter CDF of  $\phi$ 

**Lemma**. For each  $\varphi, \psi \in \mathcal{D}(X)$ , the following statements hold:

- 1.  $\varphi \leq_{\mathrm{F}} \psi \implies G^{\varphi} \leqslant G^{\psi}$
- 2. If X is totally ordered by  $\leq$ , then  $G^{\varphi} \leqslant G^{\psi} \implies \varphi \leq_F \psi$



**Lemma.**  $\leq_F$  is a partial order on  $\mathfrak{D}(X)$ 

# Proof:

Let's just prove transitivity

Suppose  $f, g, h \in \mathcal{D}(X)$  with  $f \leq_F g$  and  $g \leq_F h$ 

Fixing  $u \in i\mathbb{R}^X$ , we have

$$\langle u, f \rangle \leqslant \langle u, g \rangle$$
 and  $\langle u, g \rangle \leqslant \langle u, h \rangle$ 

Hence  $\langle u, f \rangle \leqslant \langle u, h \rangle$ 

Since u was arbitrary in  $i\mathbb{R}^X$ , we are done