# Pareto Optimality and Planner's Problem

Econ 2100

Fall 2018

Lecture 15, October 22

#### Outline

- 4 Hyperplanes
- Minkowski's Separating Hyperplane Theorem
- Pareto optimality and social welfare maximization

## From Last Class

The utility possibility set is

$$\mathbb{U} = \left\{ (v_1, ..., v_I) \in \mathbb{R}^I : \text{ there exists a feasible } (x, y) \\ \text{such that } v_i \leq u(x_i) \text{ for } i = 1, ..., I \right\}$$

The utility possibility frontier is

$$\mathbb{UF} = \{ (\bar{v}_1, ..., \bar{v}_I) \in \mathbb{U} : \text{there is no } v \in \mathbb{U} \text{ such that } v > \bar{v} \}$$

• A (linear) social welfare function is a weighted sum of the individuals' utilities:

$$\sum_{i=1}^{l} \lambda_i v_i = \lambda \cdot v \qquad \text{with } \lambda_i \ge 0$$

#### **Theorem**

If the allocation  $(\hat{x}, \hat{y})$  is feasible for the economy  $\mathcal{E} = \left\{ \left\{ u_i, \omega_i \right\}_{i=1}^{I}, \left\{ Y_j \right\}_{j=1}^{J} \right\}$  and solves the problem

$$\max_{(x,y) \text{ is a feasible allocation}} \sum_{i=1}^{l} \lambda_i u_i(x_i) \qquad \text{where } \lambda_i > 0 \text{ for all } i$$

then  $(\hat{x}, \hat{y})$  is Pareto optimal.

- We are looking for a converse: any Pareto optimal allocation maximizes some social welfare function.
  - This means, find the  $\lambda$ s that work.

## From Last Class

- Consider the function  $W(x,y) = \sum_{i=1}^{l} \lambda_i u_i(x_i)$  where  $\lambda_i \geq 0$  for all i and (x,y) is an allocation.
  - Notice this is defined over allocations, not utility vectors.
- Think of it as the composition of

$$U(x, y) = (u_1(x_1), ..., u_l(x_l))$$
 and  $f(v) = \lambda \cdot v$ 

where  $U: \mathbb{A} \to \mathbb{R}^I$  (A is the set of allocations) and  $f: \mathbb{R}^I \to \mathbb{R}$ .

ullet The image of the set of feasible allocations under U is:

$$\mathbb{V} = \{U(x,y) \in \mathbb{R}^I : (x,y) \text{ is a feasible allocation}\}$$

- V is not the utility possibility set (it is smaller).
- The allocation  $(\hat{x}, \hat{y})$  solves the problem

$$\max_{(x,y) \text{ is a feasible allocation}} W(x,y)$$

if and only if the vector  $\hat{v} = U(\hat{x}, \hat{y})$  solves the problem  $\max_{v \in \mathbb{V}} \lambda \cdot v$ .

- Any Pareto optimal allocation maximizes  $\lambda \cdot v$  over the set  $\mathbb{V}$ .
- We want to think about the problem  $\max_{c \in \mathbb{R}} \lambda \cdot v$  geometrically.

## **Definition**

Given  $r \in \mathbb{R}$ , an hyperplane is defined as  $\{x \in \mathbb{R}^N : p \cdot x = r\}$ .

An hyperplane is orthogonal to p in the sense that

$$\left\{x \in \mathbb{R}^N : p \cdot x = r\right\} = \left\{x \in \mathbb{R}^N : p \cdot x = 0\right\} + \hat{x}$$

where  $\hat{x}$  is any vector such that  $p \cdot \hat{x} =$ 

- ullet For any  $t\in\mathbb{R}$ , the vectors tp form a line through the origin in direction p.
- The hyperplane is the boundary between two halves of  $\mathbb{R}^N$ : one in which  $p \cdot x < r$  and the other in which  $p \cdot x > r$ .

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# Hyperplanes and Maximization

- Suppose that  $B \subset \mathbb{R}^N$  and that  $x^*$  solves  $\max_{x \in B} p \cdot x$ .
- The hyperplane

$$\left\{x \in \mathbb{R}^N : p \cdot x = p \cdot x^*\right\}$$

is 'tangent' to B at the point  $x^*$ , and this is the hyperplane that is furthest in the direction from 0 to p.

- Draw a picture (highlight the half spaces below and above the tangent hyperplane).
- Think about pictures where the tangent is not unique.

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• Given this geomety, the content of the last proposition is that if the function  $\lambda \cdot v$  is maximized over  $\mathbb V$  at  $\hat v$  then the hyperplane

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- This hyperplane gives the  $\lambda$ s we need to assert that for any Pareto optimal allocation there exists a vector of individual weights such that this allocation maximizes W(x,y).
- To use these ideas, one needs to know what the set V looks like
- The utility possibility set U is equal to the set V plus all the points dominated by points in V.
- Therefore, if  $(\hat{x}, \hat{y})$  maximizes W(x, y) then  $\mathbb{U}$  is tangent to the hyperplane  $\{v \in \mathbb{R}^l : \lambda \cdot v = \lambda \cdot \hat{v}\}$  at  $\hat{v}$ .

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# Hyperplanes and Pareto Optimality: An Example

- Draw a picture for an Edgeworth box economy.
- $\hat{v} = (u_A(\hat{x}_A), u_B(\hat{x}_B))$ , and the Pareto optimal allocation  $\hat{x}$  maximizes the social welfare function

$$\lambda_A u_A(x_A) + \lambda_B u_B(x_B)$$

• The vector  $\hat{v}$  is a solution to

$$\max_{v \in \mathbb{U}} \lambda_A v_A + \lambda_B v_B$$

- We want the converse: if  $(\tilde{x}, \tilde{y})$  is Pareto optimal then there exists some vector  $\tilde{\lambda}$  such that the hyperplane  $\tilde{\lambda} \cdot \tilde{v}$  is tangent to  $\mathbb{U}$  at  $\tilde{v}$ .
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#### **Definition**

Given two sets  $X,Y\subset\mathbb{R}^N$ , a vector  $p\in\mathbb{R}^N$  is said to separate these sets if  $p\neq 0$  and  $p\cdot x\leq p\cdot y$  for all  $x\in X$  and  $y\in Y$ .

## Draw a picture

- If the vector p separates X from Y, then a straight line H perpendicular to p lies between X and Y.
- If the vector p separates X from Y, then there is an hyperplane  $\{x \in \mathbb{R}^N : p \cdot x = r\}$  that comes between X and Y, though it may touch one or both sets on their boundaries.

## Theorem (Separating Hyperplane (Minkowski)

Let  $X,Y\subset\mathbb{R}^N$  be convex sets; suppose that intX is not empty, and that  $Y\cap int X=\emptyset$ . Then, there exists a p
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## Pareto Efficient Allocations and Planner's Problem

## **Theorem**

Assume the economy  $\mathcal{E} = \left\{ \left\{ u_i, \omega_i \right\}_{i=1}^{J}, \left\{ Y_j \right\}_{j=1}^{J} \right\}$  is such that

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If  $(\hat{x}, \hat{y})$  is a Pareto optimal allocation, then there exists  $\lambda \in \mathbb{R}^{I}_{+}$  with  $\lambda \neq 0$  such that  $(\hat{x}, \hat{y})$  solves

$$\max_{(x,y) \text{ is a feasible allocation }} \sum_{i=1}^{r} \lambda_i u_i(x_i)$$

- For any Pareto efficient allocation, there is a vector of weights that makes that allocation the solution to the planner's problem.
- Notice that in this theorem some  $\lambda$ s can be zero, so this is not quite a converse of the theorem from last class.

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## **Proof Part**

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 where  $\hat{v} = u(\hat{x}, \hat{y})$ 

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  - This contradicts Pareto optimality of  $(\hat{x}, \hat{y})$ .

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$$\Gamma = \{ v \in \mathbb{R}^l : v \ge \hat{v} \}$$
 where  $\hat{v} = u(\hat{x}, \hat{y})$ 

- Clearly,  $\hat{v}$  belongs to  $\mathbb{U}$  and  $\Gamma$ .
- ullet Claim: no other point belongs to  ${\mathbb U}$  and  $\Gamma$ .
  - Suppose not. There exists  $w \neq \hat{v}$  that belongs to both sets. Thus:
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Geometric idea of proof: start from a Pareto optimal allocation, and show that there is an hyperplane through this allocation that separates the utility possibility set from the set of utility vectors that improve on that allocation.

## **Proof Part I**

Let  $\mathbb U$  be the utility possibility set for  $\mathcal E$  and given a Pareto optimal allocation  $(\hat x,\hat y)$  let  $\Gamma$  be defined as

$$\Gamma = \left\{ v \in \mathbb{R}^I : v \ge \hat{v} \right\} \quad \text{where } \hat{v} = u\left(\hat{x}, \hat{y}\right)$$

- Clearly,  $\hat{v}$  belongs to  $\mathbb{U}$  and  $\Gamma$ .
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  - This contradicts Pareto optimality of  $(\hat{x}, \hat{y})$ .
- Draw a picture.

We have shown that  $\mathbb U$  and  $\Gamma$  'touch' only at  $\hat v=u\,(\hat x,\hat y)$ .

Proof Part I

We have shown that  $\mathbb{U}$  and  $\Gamma$  'touch' only at  $\hat{v} = u(\hat{x}, \hat{y})$ .

## **Proof Part II**

- U is convex (homework), and Γ is also convex (obvious); the interior of Γ does not intersect U and it is obviously not empty.
- Therefore, by Minkowksi: there exists  $\lambda \in \mathbb{R}^{I}$  with  $\lambda \neq 0$  such that

$$\lambda \cdot w \ge \lambda \cdot v$$
 for each  $w \in \Gamma$  and for each  $v \in \Gamma$ 

ullet I claim that  $\lambda>0$  (showing  $\geq$  is enough since it cannot be 0)

- To any leasible anotation (x,y) let v=u(x,y).
- Since  $\hat{v} \in \Gamma$  and  $v \in U$ , separation says  $\lambda \cdot \hat{v} \geq \lambda \cdot v$ .
- Therefore

$$\sum_{i=1}^{I} \lambda_{i} u_{i}\left(\hat{x}_{i}\right) \geq \sum_{i=1}^{I} \lambda_{i} u_{i}\left(x_{i}\right) \quad \text{for each feasible } \left(x,y\right)$$

Hence  $(\hat{x}, \hat{y})$  maximizes W(x, y).

Why does the proof use  $\mathbb U$  instead of  $\mathbb V$ ?

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For any feasible allocation 
$$(x, y)$$
 let  $v = u(x, y)$ .

- Since  $\hat{v} \in \Gamma$  and  $v \in \mathbb{U}$ , separation says  $\lambda \cdot \hat{v} \geq \lambda \cdot v$ .
- **T**

$$\sum_{i=1}^{l} \lambda_{i} u_{i}\left(\hat{x}_{i}\right) \geq \sum_{i=1}^{l} \lambda_{i} u_{i}\left(x_{i}\right) \quad \text{for each feasible } (x, y)$$

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$$\lambda\cdot(\hat{v}+(1,0,...,0))\geq\lambda\cdot\hat{v}=v \Rightarrow \qquad \lambda_1\geq0$$
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• Why does the proof use  $\mathbb U$  instead of  $\mathbb V$ ?

- Another planner's problem aims to maximize the utility of one consumer subject to everyone else achieving a prespecified utility level.
- Consider, for simplicity, an exchange economy.

## Proposition

In an exchange economy where each i=1,...,l,  $\succsim_i$  can be represented by a strictly increasing and continuous utility function  $u_i\left(x_i\right)$ , an allocation  $\hat{x}$  is Pareto optimal if and only if it solves the following

$$u_i(x_i) \geq \bar{u}_i \text{ for all } i \neq j,$$

$$\max_{\mathbf{x}} u_j(\mathbf{x}_j)$$
 subject to:  $\sum_{i=1}^{r} \mathbf{x}_i = \sum_{i=1}^{r} \omega_i$ 

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- An allocation is Pareto optimal if and only if it maximizes one consumer's utility subject to all others obtaining some given utility level.
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## Finding Pareto Optimal Allocations: An Example

Consider an economy: where

$$\begin{array}{ll} \text{consumer 1} & \text{consumer 2} \\ \omega_1 = (\omega_{11}, \omega_{22}) & \omega_2 = (\omega_{12}, \omega_{22}) \\ u_1 \left( x_1 \right) = u_1 \left( x_{11}, x_{21} \right) & u_2 \left( x_2 \right) = u_2 \left( x_{12}, x_{22} \right) \end{array}$$

 To find the set of Pareto optimal allocations for an Edgeworth box economy in which consumers' utility functions are

$$u_1(x_{11}, x_{21}) = (x_{11})^{\alpha} (x_{21})^{1-\alpha}$$
 and  $u_2(x_{12}, x_{22}) = (x_{12})^{\beta} (x_{22})^{1-\beta}$ 

• We must solve the following planner's problem

$$\max_{\substack{x_{11},x_{21},x_{12},x_{22}\\x_{11},x_{21},x_{12},x_{22}\\x_{21},x_{22},x_{22}\\x_{21},x_{22},x_{22}\\x_{21},x_{22},x_{22}\\x_{22}} (x_{11})^{\alpha} (x_{21})^{1-\alpha} \quad \text{subject to} \quad \begin{aligned} &(x_{12})^{\beta} (x_{22})^{1-\beta} \geq \bar{u} \\ &x_{11}+x_{12}=\omega_{11}+\omega_{12} \\ &x_{21}+x_{22}=\omega_{21}+\omega_{22} \\ &x_{11},x_{21},x_{12},x_{22}\geq 0 \end{aligned}$$

for some  $ar{u}$  .

 In this case, the utility functions are differentiable, so we can write the Lagrangean, the first order conditions, and then solve.

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$$\max_{x_{11}, x_{21}, x_{12}, x_{22}} (x_{11})^{\alpha} (x_{21})^{1-\alpha} \quad \text{subject to} \quad \begin{array}{c} (x_{21})^{\alpha} \\ x_{21} \\ x_{21} \\ x_{22} \\ x_{23} \\ x_{24} \\ x_{24} \\ x_{24} \\ x_{24} \\ x_{25} \\ x_{25} \\ x_{26} \\$$

$$(x_{12})^{\beta} (x_{22})^{1-\beta} \ge \bar{u}$$

$$x_{11} + x_{12} = \omega_{11} + \omega_{12}$$

$$x_{21} + x_{22} = \omega_{21} + \omega_{22}$$

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for some  $\bar{u}$ .

 In this case, the utility functions are differentiable, so we can write the Lagrangean, the first order conditions, and then solve.

#### Finding Pareto Optimal Allocations: An Example

Consider an economy: where

$$\begin{array}{ll} \text{consumer 1} & \text{consumer 2} \\ \omega_1 = (\omega_{11}, \omega_{22}) & \omega_2 = (\omega_{12}, \omega_{22}) \\ u_1 \left( x_1 \right) = u_1 \left( x_{11}, x_{21} \right) & u_2 \left( x_2 \right) = u_2 \left( x_{12}, x_{22} \right) \end{array}$$

 To find the set of Pareto optimal allocations for an Edgeworth box economy in which consumers' utility functions are

$$u_1(x_{11}, x_{21}) = (x_{11})^{\alpha} (x_{21})^{1-\alpha}$$
 and  $u_2(x_{12}, x_{22}) = (x_{12})^{\beta} (x_{22})^{1-\beta}$ 

• We must solve the following planner's problem:

$$\max_{\substack{x_{11},x_{21},x_{12},x_{22}\\ x_{11},x_{21},x_{12},x_{22}}} (x_{11})^{\alpha} (x_{21})^{1-\alpha} \quad \text{subject to} \quad \begin{array}{l} (x_{12})^{\beta} (x_{22})^{1-\beta} \geq \bar{u} \\ x_{11}+x_{12}=\omega_{11}+\omega_{12} \\ x_{21}+x_{22}=\omega_{21}+\omega_{22} \\ x_{11},x_{21},x_{12},x_{22} \geq 0 \end{array}$$

for some  $\bar{u}$ .

• In this case, the utility functions are differentiable, so we can write the Lagrangean, the first order conditions, and then solve.

# **Next Class**

• Competitive equilibrium