

# Equilibrium Welfare Properties

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# Introduction

- ▶ In this section we will prove:
  - ▶ The *First Welfare Theorem*: The allocation from any competitive equilibrium with transfers is Pareto optimal.
  - ▶ The *Second Welfare Theorem*: For any Pareto optimal allocation, there is a price vector that can support it as an equilibrium with transfers.
- ▶ Both theorems require complete markets, rational and locally nonsatiated preferences, and nonempty and closed production sets.
- ▶ However, the second welfare theorem requires a number of additional assumptions.

# Price Equilibrium with Transfers

## Definition

Given an economy specified by  $\left(\{(X_i, \geq_i)\}_{i=1}^I, \{Y_j\}_{j=1}^J, \bar{\omega}\right)$ , an allocation  $(\mathbf{x}^\star, \mathbf{y}^\star)$  and a price vector  $\mathbf{p}$  constitute a *price equilibrium with transfers* if there is an assignment of wealth levels  $(w_1, \dots, w_I)$  with  $\sum_{i=1}^I w_i = \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}_j^\star$  such that

- (i) For every  $j$ ,  $\mathbf{y}_j^\star$  maximizes profits in  $Y_j$ ; that is,

$$\mathbf{p} \cdot \mathbf{y}_j \leq \mathbf{p} \cdot \mathbf{y}_j^\star \text{ for all } \mathbf{y}_j \in Y_j$$

- (ii) For every  $i$ ,  $\mathbf{x}_i^\star$  is maximal for  $\geq_i$  in the budget set:

$$\{\mathbf{x}_i \in X_i : \mathbf{p} \cdot \mathbf{x}_i \leq w_i\}$$

- (iii)  $\sum_{i=1}^I \mathbf{x}_i^\star = \bar{\omega} + \sum_{j=1}^J \mathbf{y}_j^\star$ .

If  $w_i = \mathbf{p} \cdot \omega_i + \sum_{j=1}^J \theta_{ij} \mathbf{p} \cdot \mathbf{y}_j \forall i$ , then there are no transfers.

# The First Fundamental Theorem of Welfare Economics

## Theorem

*If preferences are locally nonsatiated, and if  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  is a price equilibrium with transfers, then the allocation  $(\mathbf{x}^*, \mathbf{y}^*)$  is Pareto optimal.*

## Proof:

1. Because  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  is an equilibrium, if  $\mathbf{x}_i \succ_i \mathbf{x}_i^*$ , then  $\mathbf{p} \cdot \mathbf{x}_i > w_i$ .
2. Furthermore, if  $\mathbf{x}_i \geq_i \mathbf{x}_i^*$ , then  $\mathbf{p} \cdot \mathbf{x}_i \geq w_i$ .
  - ▶ Suppose there is an  $\mathbf{x}'_i$  satisfying  $\mathbf{x}'_i \geq_i \mathbf{x}_i^*$  but  $\mathbf{p} \cdot \mathbf{x}'_i < w_i$ .
  - ▶ By LNS,  $\exists \mathbf{x}''_i$  arbitrarily close to  $\mathbf{x}'_i$  where  $\mathbf{x}''_i \succ_i \mathbf{x}'_i$  and  $\mathbf{p} \cdot \mathbf{x}''_i < w_i$ .
  - ▶ But this contradicts that  $\mathbf{x}_i^*$  was maximal in  $i$ 's budget set, because by transitivity  $\mathbf{x}''_i \succ_i \mathbf{x}_i^*$ .

# First Welfare Theorem Proof

3. Suppose  $\exists (\mathbf{x}', \mathbf{y}')$  that Pareto dominates  $(\mathbf{x}^*, \mathbf{y}^*)$ .
  - ▶ By (1) & (2),  $\mathbf{p} \cdot \mathbf{x}'_i \geq w_i \forall i$  and  $\mathbf{p} \cdot \mathbf{x}'_i > w_i$  for at least one  $i$ .
  - ▶ So  $\sum_{i=1}^I \mathbf{p} \cdot \mathbf{x}'_i > \sum_{i=1}^I w_i = \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}^*_j$ .
4. Because  $\mathbf{y}^*_j$  is profit-maximizing at  $\mathbf{p}$ , for all  $j$  we have  $\mathbf{p} \cdot \mathbf{y}^*_j \geq \mathbf{p} \cdot \mathbf{y}_j \forall \mathbf{y}_j \in Y_j$ .
  - ▶ Therefore  $\mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}^*_j \geq \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}'_j$ .
5. Because  $(\mathbf{x}', \mathbf{y}')$  is Pareto improving:  $\sum_{i=1}^I \mathbf{x}'_i = \bar{\omega} + \sum_{j=1}^J \mathbf{y}'_j$ .
  - ▶ This implies  $\sum_{i=1}^I \mathbf{p} \cdot \mathbf{x}'_i = \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}'_j$
6. But (3) & (4) imply  $\sum_{i=1}^I \mathbf{p} \cdot \mathbf{x}'_i > \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}'_j$ .
  - ▶ But this contradicts (5).

# Separating and Supporting Hyperplane Theorems

We will use these two theorems to prove certain propositions:

## Theorem (Separating Hyperplane Theorem)

*Suppose that the convex sets  $\mathcal{A} \subset \mathbb{R}^N$  and  $\mathcal{B} \subset \mathbb{R}^N$  are disjoint. Then there is  $\mathbf{p} \in \mathbb{R}^N$  with  $\mathbf{p} \neq \mathbf{0}$  and a value  $c \in \mathbb{R}$  such that  $\mathbf{p} \cdot \mathbf{x} \geq c$  for every  $\mathbf{x}$  in  $\mathcal{A}$  and  $\mathbf{p} \cdot \mathbf{y} \leq c$  for every  $\mathbf{y} \in \mathcal{B}$ .*

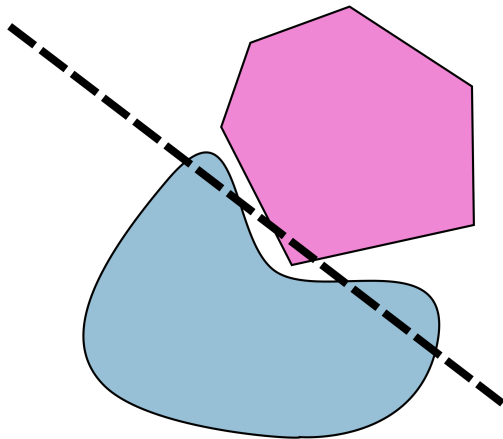
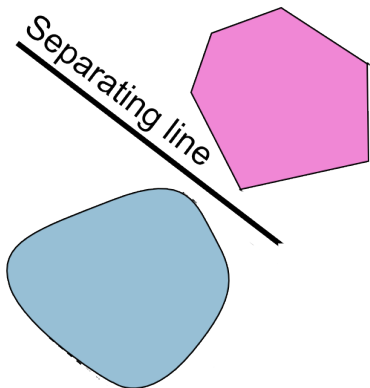
- There is a hyperplane that separates  $A$  and  $B$ , with  $A$  and  $B$  on different sides of it.

## Theorem (Supporting Hyperplane Theorem)

*Suppose that  $\mathcal{B} \subset \mathbb{R}^N$  is convex and that  $\mathbf{x}$  is not an element of the interior of the set  $\mathcal{B}$ . Then there is a  $\mathbf{p} \in \mathbb{R}^N$  with  $\mathbf{p} \neq \mathbf{0}$  such that  $\mathbf{p} \cdot \mathbf{x} \geq \mathbf{p} \cdot \mathbf{y}$  for every  $\mathbf{y} \in \mathcal{B}$ .*

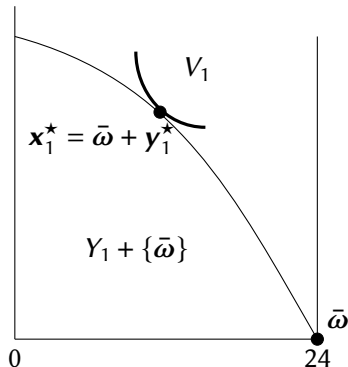
## Examples

- ▶ Example 1: 2 convex, disjoint sets. SHT can be applied.
- ▶ Example 2: 2 nonconvex, disjoint sets. SHT can't be applied.



## SHT Example in the Robinson Crusoe Economy

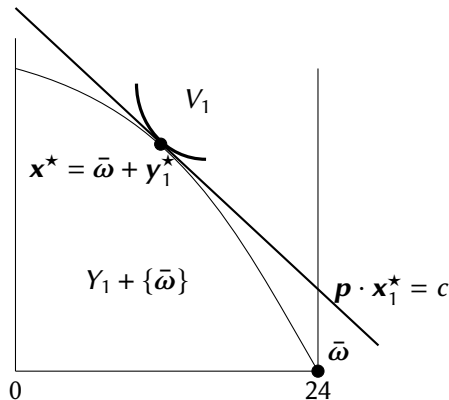
- ▶ Suppose  $(\mathbf{x}_1^*, \mathbf{y}_1^*)$  is Pareto optimal.
- ▶ Crusoe's "better than set" is  $V_1 = \{\mathbf{x}_1 \in X_1 : \mathbf{x}_1 \succ_1 \mathbf{x}_1^*\}$ .
- ▶ The two sets  $V_1$  and  $Y_1 + \{\bar{\omega}\}$  are:
  - ▶ disjoint (by Pareto the optimality of  $(\mathbf{x}_1^*, \mathbf{y}_1^*)$ ), and
  - ▶ convex (if  $\succeq_1$  and  $Y_1$  are convex).
- ▶ The separating hyperplane theorem can be applied.





## SHT Example in the Robinson Crusoe Economy

- ▶ The SHT says  $\exists \mathbf{p} \neq \mathbf{0}$  and a  $c$  such that  $\mathbf{p} \cdot \mathbf{x}_1 \geq c \ \forall \mathbf{x}_1 \in V_1$  and  $\mathbf{p} \cdot (\mathbf{y}_1 + \bar{\omega}) \leq c \ \forall \mathbf{y}_1 + \bar{\omega} \in Y_1 + \{\bar{\omega}\}$ .



- ▶ What we will show: if we transfer wealth  $w_1 = c = \mathbf{p} \cdot \mathbf{x}_1^*$  to Crusoe,  $(\mathbf{x}_1^*, \mathbf{y}_1^*, \mathbf{p})$  is an equilibrium.

# The Second Fundamental Theorem of Welfare Economics

## Theorem

*Consider an economy specified by  $\left(\{(X_i, \succeq_i)\}_{i=1}^I, \{Y_j\}_{j=1}^J, \bar{\omega}\right)$ , and suppose that*

- ▶ *Every  $X_i$  is convex with  $\mathbf{0} \in X_i$ .*
- ▶ *Every preference relation  $\succeq_i$  is convex, continuous and locally nonsatiated.*
- ▶ *Every  $Y_j$  is convex and exhibits free disposal.*

*If  $(\mathbf{x}^*, \mathbf{y}^*)$  is a Pareto optimal allocation, where  $\mathbf{x}_i^* \gg \mathbf{0}$  for all  $i$ , there exists a price vector  $\mathbf{p} \geq \mathbf{0}$ ,  $\mathbf{p} \neq \mathbf{0}$  such that  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  is a price equilibrium with transfers.*

Thus, there is a price vector and an assignment of wealth levels  $(w_1, \dots, w_I)$  satisfying  $\sum_{i=1}^I w_i = \mathbf{p} \cdot \bar{\omega} + \sum_{j=1}^J \mathbf{p} \cdot \mathbf{y}_j^*$  such that  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  is a Walrasian equilibrium.

## Second Welfare Theorem Proof: Preliminaries

The goal is to show that the wealth levels  $w_i = \mathbf{p} \cdot \mathbf{x}_i^*$  for all  $i$  support  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  as a price equilibrium with transfers.

Define the sets:

- ▶  $V_i = \{\mathbf{x}_i \in X_i : \mathbf{x}_i \succ_i \mathbf{x}_i^*\} \subset \mathbb{R}^L$
- ▶  $V = \sum_{i=1}^I V_i = \{\sum_{i=1}^I \mathbf{x}_i \in \mathbb{R}^L : \mathbf{x}_1 \in V_1, \dots, \mathbf{x}_I \in V_I\}$
- ▶  $Y = \sum_{j=1}^J Y_j = \{\sum_{j=1}^J \mathbf{y}_j \in \mathbb{R}^L : \mathbf{y}_1 \in Y_1, \dots, \mathbf{y}_J \in Y_J\}$
- ▶  $V$  is the set of aggregate consumption bundles that *could* be split across the  $I$  individuals with each  $i$  preferring it to  $\mathbf{x}_i^*$ .
- ▶  $Y + \{\bar{\omega}\}$  is the set of aggregate bundles producible with the given technology and endowments.

With this, we split the proof into multiple steps.

## Second Welfare Theorem Proof Outline

Step 1 Every set  $V_i$  is convex.

Step 2 The sets  $V$  and  $Y + \{\bar{\omega}\}$  are convex.

Step 3  $V$  and  $Y + \{\bar{\omega}\}$  are disjoint.

Step 4 There is a vector  $\mathbf{p} \geq \mathbf{0}$ ,  $\mathbf{p} \neq \mathbf{0}$  and a number  $c$  such that  $\mathbf{p} \cdot \mathbf{z} \geq c$  for every  $\mathbf{z} \in V$  and  $\mathbf{p} \cdot \mathbf{z} \leq c$  for every  $\mathbf{z} \in Y + \{\bar{\omega}\}$ .

Step 5 If  $\mathbf{x}_i \geq_i \mathbf{x}_i^*$  for every  $i$ , then  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i \right) \geq c$ .

Step 6  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^* \right) = \mathbf{p} \cdot \left( \bar{\omega} + \sum_{j=1}^J \mathbf{y}_j^* \right) = c$ .

Step 7 For every  $j$ , we have  $\mathbf{p} \cdot \mathbf{y}_j \leq \mathbf{p} \cdot \mathbf{y}_j^*$  for all  $\mathbf{y}_j \in Y_j$ .

Step 8 For every  $i$ , if  $\mathbf{x}_i \succ_i \mathbf{x}_i^*$ , then  $\mathbf{p} \cdot \mathbf{x}_i > \mathbf{p} \cdot \mathbf{x}_i^*$ .

Step 9 Steps 7 & 8 with feasibility from the Pareto optimal allocation implies that the wealth levels  $w_i = \mathbf{p} \cdot \mathbf{x}_i^*$  for all  $i$  support  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  as a price equilibrium with transfers.

## Second Welfare Theorem Proof

### Step 1

Every set  $V_i = \{\mathbf{x}_i \in X_i : \mathbf{x}_i \succ_i \mathbf{x}_i^\star\}$  is convex.

- ▶ We need to show that if  $\mathbf{x}_i \in V_i$  and  $\mathbf{x}'_i \in V_i$ , then  $\mathbf{x}_i^\alpha = \alpha\mathbf{x}_i + (1 - \alpha)\mathbf{x}'_i \in V_i$  for all  $\alpha \in [0, 1]$ .
- ▶ First, by the convexity of  $X_i$ ,  $\mathbf{x}_i^\alpha \in X_i$ .
- ▶  $\mathbf{x}_i, \mathbf{x}'_i \in V_i$  means  $\mathbf{x}_i \succ_i \mathbf{x}_i^\star$  and  $\mathbf{x}'_i \succ_i \mathbf{x}_i^\star$ .
- ▶ Suppose wlog that  $\mathbf{x}_i \geq_i \mathbf{x}'_i$ .
- ▶ Because preferences are convex:  $\mathbf{x}_i^\alpha \geq_i \mathbf{x}'_i \forall \alpha \in [0, 1]$
- ▶ Then by transitivity  $\mathbf{x}_i^\alpha \succ_i \mathbf{x}_i^\star$ .
- ▶ Hence  $\mathbf{x}_i^\alpha \in V_i$ .

# Second Welfare Theorem Proof

## Step 2

The sets  $V$  and  $Y + \{\bar{\omega}\}$  are convex.

- ▶ The sum of convex sets is convex.
  - ▶ See note at end of slide deck for  $I = 2$  case.

## Step 3

$V$  and  $Y + \{\bar{\omega}\}$  are disjoint.

- ▶  $V$  contains all bundles that can be distributed such that everyone is strictly better off than with  $\mathbf{x}_i^*$ .
- ▶  $Y + \{\bar{\omega}\}$  is the set of all feasible bundles.
- ▶ If they were not disjoint, then  $(\mathbf{x}^*, \mathbf{y}^*)$  would not be Pareto optimal.

## Second Welfare Theorem Proof

### Step 4

There is a vector  $\mathbf{p} \geq \mathbf{0}$ ,  $\mathbf{p} \neq \mathbf{0}$  and a number  $c$  such that  $\mathbf{p} \cdot \mathbf{z} \geq c$  for every  $\mathbf{z} \in V$  and  $\mathbf{p} \cdot \mathbf{z} \leq c$  for every  $\mathbf{z} \in Y + \{\bar{\omega}\}$ .

- ▶ That such a  $\mathbf{p} \in \mathbb{R}^L$ ,  $\mathbf{p} \neq \mathbf{0}$  exists follows directly from the separating hyperplane theorem (two disjoint convex sets).
- ▶ We only need to rule out the possibility of  $p_\ell < 0$  for any  $\ell$ .
- ▶ Because firms have free disposal, if  $p_\ell < 0$  then  $\mathbf{p} \cdot \mathbf{y}_j$  could become unboundedly large, violating  $\mathbf{p} \cdot \mathbf{z} \leq c$  for all  $\mathbf{z} \in Y + \{\bar{\omega}\}$ .

## Second Welfare Theorem Proof

### Step 5

If  $\mathbf{x}_i \succeq_i \mathbf{x}_i^*$  for every  $i$ , then  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i \right) \geq c$ .

- ▶ Take  $\mathbf{x}_i \succeq_i \mathbf{x}_i^*$ . By LNS we have,  $\forall \varepsilon > 0$ ,  $\exists \hat{\mathbf{x}}_i$  satisfying  $\|\hat{\mathbf{x}}_i - \mathbf{x}_i\| \leq \varepsilon$  such that  $\hat{\mathbf{x}}_i \succ_i \mathbf{x}_i$ .
- ▶ By transitivity  $\hat{\mathbf{x}}_i \succ_i \mathbf{x}_i^*$  so  $\hat{\mathbf{x}}_i \in V_i$ .
- ▶ Such a  $\hat{\mathbf{x}}_i$  exists for every consumer, so  $\sum_{i=1}^I \hat{\mathbf{x}}_i \in V$ .
- ▶ By Step 4:  $\mathbf{p} \cdot \left( \sum_{i=1}^I \hat{\mathbf{x}}_i \right) \geq c$ .
- ▶ As  $\varepsilon \rightarrow 0$  (so  $\hat{\mathbf{x}}_i \rightarrow \mathbf{x}_i \forall i$ ), we have  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i \right) \geq c$ .
  - ▶ Limits preserve inequalities.



## Second Welfare Theorem Proof

- ▶ As a consequence of Step 5, because  $\mathbf{x}_i^\star \geq_i \mathbf{x}_i^\star$ , we have  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^\star \right) \geq c$

### Step 6

$$\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^\star \right) = \mathbf{p} \cdot \left( \bar{\omega} + \sum_{j=1}^J \mathbf{y}_j^\star \right) = c.$$

- ▶ By feasibility,  $\sum_{i=1}^I \mathbf{x}_i^\star = \sum_{j=1}^J \mathbf{y}_j^\star + \bar{\omega} \in Y + \{\bar{\omega}\}$ .
- ▶ Therefore  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^\star \right) \leq c$  because  $\mathbf{p} \cdot \mathbf{z} \leq c$  for every  $\mathbf{z} \in Y + \{\bar{\omega}\}$ .
- ▶ But Step 5 implies that  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^\star \right) \geq c$
- ▶ Therefore  $\mathbf{p} \cdot \left( \sum_{i=1}^I \mathbf{x}_i^\star \right) = c$ .

## Second Welfare Theorem Proof

### Step 7

For every  $j$ , we have  $\mathbf{p} \cdot \mathbf{y}_j \leq \mathbf{p} \cdot \mathbf{y}_j^\star$  for all  $\mathbf{y}_j \in Y_j$ .

- ▶ For all firms,  $\forall \mathbf{y}_j \in Y_j$  we have  $\mathbf{y}_j + \sum_{h \neq j} \mathbf{y}_h^\star \in Y$ .
- ▶ From Steps 4 and 6,  $\forall \mathbf{y}_j \in Y_j$ :

$$\mathbf{p} \cdot \left( \bar{\omega} + \mathbf{y}_j + \sum_{h \neq j} \mathbf{y}_h^\star \right) \leq c = \mathbf{p} \cdot \left( \bar{\omega} + \mathbf{y}_j^\star + \sum_{h \neq j} \mathbf{y}_h^\star \right)$$

- ▶ Cancelling terms yields  $\mathbf{p} \cdot \mathbf{y}_j \leq \mathbf{p} \cdot \mathbf{y}_j^\star$  for all  $\mathbf{y}_j \in Y_j$ , for all  $j$ .

## Second Welfare Theorem Proof

### Step 8

For every  $i$ , if  $\mathbf{x}_i \succ_i \mathbf{x}_i^\star$ , then  $\mathbf{p} \cdot \mathbf{x}_i > \mathbf{p} \cdot \mathbf{x}_i^\star$ .

- ▶ If  $\mathbf{x}_i \succ_i \mathbf{x}_i^\star$ , then  $\mathbf{x}_i \in V_i$ . From Steps 5 and 6 above we have:

$$\mathbf{p} \cdot \left( \mathbf{x}_i + \sum_{k \neq i} \mathbf{x}_k^\star \right) \geq c = \mathbf{p} \cdot \left( \mathbf{x}_i^\star + \sum_{k \neq i} \mathbf{x}_k^\star \right)$$

- ▶ Cancelling terms yields  $\mathbf{p} \cdot \mathbf{x}_i \geq \mathbf{p} \cdot \mathbf{x}_i^\star$ .
- ▶ Now we just need to rule out the  $\mathbf{p} \cdot \mathbf{x}_i = \mathbf{p} \cdot \mathbf{x}_i^\star$  case.

## Second Welfare Theorem Proof

- ▶ Suppose toward a contradiction there is a  $\mathbf{x}'_i \in \mathbb{R}^L_+$  satisfying  $\mathbf{x}'_i \succ_i \mathbf{x}^\star_i$  such that  $\mathbf{p} \cdot \mathbf{x}'_i = \mathbf{p} \cdot \mathbf{x}^\star_i$ .
- ▶ Because  $\mathbf{0} \in X_i$  and  $X_i$  is convex,  $\alpha \mathbf{x}'_i + (1 - \alpha) \mathbf{0} \in X_i$  for all  $\alpha \in [0, 1]$ .
- ▶ Because  $\mathbf{p} \geq \mathbf{0}$ ,  $\mathbf{p} \neq \mathbf{0}$  and  $\mathbf{x}^\star_i \gg \mathbf{0}$ , we know that  $\mathbf{p} \cdot \mathbf{x}^\star_i > 0$ .
- ▶  $\forall \alpha \in [0, 1)$ ,  $\alpha \mathbf{p} \cdot \mathbf{x}'_i + (1 - \alpha) \mathbf{p} \cdot \mathbf{0} < \mathbf{p} \cdot \mathbf{x}^\star_i$ .
- ▶ By continuity, for  $\alpha$  close enough to 1,  $\alpha \mathbf{x}'_i \succ_i \mathbf{x}^\star_i$ .
- ▶ As we have found a bundle that is preferred to  $\mathbf{x}^\star_i$  and is strictly cheaper, we have found a contradiction to what we found above.

## Second Welfare Theorem Proof

### Step 9

If we assign wealth levels  $w_i = \mathbf{p} \cdot \mathbf{x}_i^*$  to each consumer,  $(\mathbf{x}^*, \mathbf{y}^*, \mathbf{p})$  is a price equilibrium with transfers.

This satisfies all the conditions for equilibrium:

- ▶ By Step 8: If  $\mathbf{x}_i \succ_i \mathbf{x}_i^*$ , then  $\mathbf{p} \cdot \mathbf{x}_i > w_i, \forall i$ .
  - ▶  $\mathbf{x}_i^*$  is maximal for  $\succeq_i$  in the budget set.
- ▶ By Step 7:  $\mathbf{p} \cdot \mathbf{y}_j \leq \mathbf{p} \cdot \mathbf{y}_j^*$  for all  $\mathbf{y}_j \in Y_j, \forall j$ 
  - ▶  $\mathbf{y}_j^*$  maximizes profits in  $Y_j$ .
- ▶ Because  $(\mathbf{x}^*, \mathbf{y}^*)$  is Pareto optimal, we have feasibility and hence market clearing in each good:

$$\sum_{i=1}^I \mathbf{x}_i^* = \bar{\omega} + \sum_{j=1}^J \mathbf{y}_j^*$$

# Utility Possibilities Set and Pareto Frontier

- Recall the utility possibility set:

$$\mathcal{U} = \{ (u_1, \dots, u_I) \in \mathbb{R}^I : \exists \text{ feasible } (\mathbf{x}, \mathbf{y}) \text{ s.t. } u_i \leq u_i(\mathbf{x}_i) \ \forall i \}$$

- The Pareto frontier is:

$$\mathcal{UP} = \left\{ (u_1, \dots, u_I) \in \mathcal{U} : \text{there is no } (u'_1, \dots, u'_I) \in \mathcal{U} \right. \\ \left. \text{such that } u'_i \geq u_i \ \forall i \text{ and } u'_i > u_i \text{ for some } i \right\}$$

## Theorem

*A feasible allocation  $(\mathbf{x}, \mathbf{y})$  is a Pareto optimum if and only if  $(u_1(\mathbf{x}_1), \dots, u_I(\mathbf{x}_I)) \in \mathcal{UP}$*

# Social Welfare

- ▶ Suppose we have the linear social welfare function:

$$W(u_1, \dots, u_I) = \sum_{i=1}^I \lambda_i u_i$$

where  $\lambda_i \geq 0 \forall i$ .

- ▶ The planner's problem is then:

$$\max_{\mathbf{u} \in \mathcal{U}} \boldsymbol{\lambda} \cdot \mathbf{u}$$

- ▶ The optimum of every linear social welfare function with  $\boldsymbol{\lambda} \gg \mathbf{0}$  is Pareto optimal.
- ▶ If  $\mathcal{U}$  is convex, every Pareto optimal allocation is the solution to the planner's problem for *some* welfare weights.

# All Social Welfare Optima are Pareto Optimal

## Theorem

*If  $\mathbf{u}^\star$  is a solution to the social welfare maximization problem*

$$\max_{\mathbf{u} \in \mathcal{U}} \boldsymbol{\lambda} \cdot \mathbf{u}$$

*with  $\boldsymbol{\lambda} \gg \mathbf{0}$ , then  $\mathbf{u}^\star \in \mathcal{UP}$ .*

**Proof:** If not, there is another  $\mathbf{u}' \in \mathcal{U}$  where  $\mathbf{u}' \geq \mathbf{u}^\star$  and  $\mathbf{u}' \neq \mathbf{u}^\star$ . Then, since  $\boldsymbol{\lambda} \gg \mathbf{0}$ , we have  $\boldsymbol{\lambda} \cdot \mathbf{u}' > \boldsymbol{\lambda} \cdot \mathbf{u}^\star$ , contradicting that  $\mathbf{u}^\star$  solved the planner's problem.



# All Pareto Optimal Allocations are a Social Welfare Optimum

## Theorem

*If the set  $\mathcal{U}$  is convex, then for any  $\tilde{\mathbf{u}} \in \mathcal{UP}$ , there is a vector of welfare weights  $\lambda \geq \mathbf{0}$ ,  $\lambda \neq \mathbf{0}$ , such that  $\lambda \cdot \tilde{\mathbf{u}} \geq \lambda \cdot \mathbf{u}$  for all  $\mathbf{u} \in \mathcal{U}$ .*

**Proof:** If  $\tilde{\mathbf{u}} \in \mathcal{UP}$ , then  $\tilde{\mathbf{u}} \in bd(\mathcal{U})$ . Using the convexity of  $\mathcal{U}$ , by the supporting hyperplane theorem,  $\exists \lambda \neq \mathbf{0}$  such that  $\lambda \cdot \tilde{\mathbf{u}} \geq \lambda \cdot \mathbf{u} \forall \mathbf{u} \in \mathcal{U}$ . Moreover  $\lambda \geq \mathbf{0}$  since otherwise you could choose a  $u_i < 0$  large enough in absolute value to get  $\lambda \cdot \mathbf{u} > \lambda \cdot \tilde{\mathbf{u}}$ .

## When is $\mathcal{U}$ convex?

- If each  $X_i$  and  $Y_i$  is convex and each  $u_i(\mathbf{x}_i)$  is concave, then  $\mathcal{U}$  is convex (part of tutorial 3).

# First-Order Conditions for Pareto Optimality

- ▶ Assume now  $X_i = \mathbb{R}_+^L$  for all  $i$ .
- ▶  $\succeq_i$  is represented by  $u_i(\mathbf{x}_i)$  which is twice continuously differentiable and satisfies  $\nabla u_i(\mathbf{x}_i) \gg \mathbf{0}$  and  $u_i(\mathbf{0}) = 0$ .
- ▶ Firm  $j$ 's production set is  $Y_j = \{\mathbf{y} \in \mathbb{R}^L : F_j(\mathbf{y}) \leq 0\}$ , where  $F_j : \mathbb{R}^L \rightarrow \mathbb{R}$  is twice continuously differentiable,  $F_j(\mathbf{0}) \leq 0$  and  $\nabla F_j(\mathbf{y}_j) \gg \mathbf{0}$ .
- ▶  $(\mathbf{x}, \mathbf{y})$  is Pareto optimal if it solves:

$$\max_{(\mathbf{x} \in \mathbb{R}_+^L, \mathbf{y} \in \mathbb{R}^L)} u_1(\mathbf{x}_1)$$

subject to:

- ▶  $u_i(\mathbf{x}_i) \geq \bar{u}_i$  for all  $i = 2, \dots, I$ .
- ▶  $F_j(\mathbf{y}_j) \leq 0$  for all  $j = 1, \dots, J$
- ▶  $\sum_{i=1}^I x_{\ell i} \leq \bar{\omega}_\ell + \sum_{j=1}^J y_{\ell j}$  for all  $\ell = 1, \dots, L$ .

# First-Order Conditions for Pareto Optimality

The Lagrangian is:

$$\mathcal{L}(\cdot) = u_1(\mathbf{x}_1) + \sum_{i=2}^I \delta_i (u_i(\mathbf{x}_i) - \bar{u}_i) + \sum_{i=1}^I \sum_{\ell=1}^L \xi_{\ell i} x_{\ell i} - \sum_{j=1}^J \gamma_j F_j(\mathbf{y}_j) + \sum_{\ell=1}^L \mu_{\ell} \left( \bar{\omega}_{\ell} + \sum_{j=1}^J y_{\ell j} - \sum_{i=1}^I x_{\ell i} \right)$$

- ▶ All constraints except for nonnegativity (with multipliers  $\xi_{\ell i}$ ) will necessarily bind at the optimum.
- ▶ The first-order conditions are (where  $\delta_1 = 1$ ):

$$x_{\ell i} : \delta_i \frac{\partial u_i}{\partial x_{\ell i}} + \xi_{\ell i} - \mu_{\ell} = 0 \text{ for all } i, \ell \text{ where } \xi_{\ell i} = 0 \text{ if } x_{\ell i} > 0$$

$$y_{\ell j} : \mu_{\ell} - \gamma_j \frac{\partial F_j}{\partial y_{\ell}} = 0 \text{ for all } j, \ell$$

# First-Order Conditions for Pareto Optimality

At an interior solution  $\mathbf{x}_i \gg \mathbf{0}$  for all  $i$ :

$$\text{Equal } MRS_{i\ell\ell'} \text{ across } i: \quad \frac{\frac{\partial u_i}{\partial x_{\ell i}}}{\frac{\partial u_i}{\partial x_{\ell' i}}} = \frac{\frac{\partial u_{i'}}{\partial x_{\ell i'}}}{\frac{\partial u_{i'}}{\partial x_{\ell' i'}}} \quad \text{for all } i, i', \ell, \ell'$$

$$\text{Equal } MRTS_{j\ell\ell'} \text{ across } j: \quad \frac{\frac{\partial F_j}{\partial y_{\ell j}}}{\frac{\partial F_j}{\partial y_{\ell' j}}} = \frac{\frac{\partial F_{j'}}{\partial y_{\ell j'}}}{\frac{\partial F_{j'}}{\partial y_{\ell' j'}}} \quad \text{for all } j, j', \ell, \ell'$$

$$MRS_{i\ell\ell'} = MRTS_{j\ell\ell'} \text{ for each } i, j: \quad \frac{\frac{\partial u_i}{\partial x_{\ell i}}}{\frac{\partial u_i}{\partial x_{\ell' i}}} = \frac{\frac{\partial F_j}{\partial y_{\ell j}}}{\frac{\partial F_j}{\partial y_{\ell' j}}} \quad \text{for all } i, j, \ell, \ell'$$

Note: If  $V_1$  and  $V_2$  are convex,  $V = V_1 + V_2$  is convex

- ▶ Take  $\mathbf{x}' = \mathbf{x}'_1 + \mathbf{x}'_2 \in V$  and  $\mathbf{x}'' = \mathbf{x}''_1 + \mathbf{x}''_2 \in V$ .
- ▶ WTS:  $\forall \alpha \in [0, 1]$  that  $\alpha \mathbf{x}' + (1 - \alpha) \mathbf{x}'' \in V$ .
- ▶  $\mathbf{x}'_1 \in V_1$  and  $\mathbf{x}''_1 \in V_1$  and similarly for  $\mathbf{x}'_2$  and  $\mathbf{x}''_2$ .
- ▶ Because  $V_1$  and  $V_2$  are convex,  $\forall \alpha \in [0, 1]$ ,  $\mathbf{x}_1^\alpha = \alpha \mathbf{x}'_1 + (1 - \alpha) \mathbf{x}''_1 \in V_1$  and similarly  $\mathbf{x}_2^\alpha \in V_2$ .
- ▶ So, by the definition of  $V$ :

$$\begin{aligned}\alpha \mathbf{x}' + (1 - \alpha) \mathbf{x}'' &= \alpha (\mathbf{x}'_1 + \mathbf{x}'_2) + (1 - \alpha) (\mathbf{x}''_1 + \mathbf{x}''_2) \\ &= \alpha \mathbf{x}'_1 + (1 - \alpha) \mathbf{x}''_1 + \alpha \mathbf{x}'_2 + (1 - \alpha) \mathbf{x}''_2 \\ &= \mathbf{x}_1^\alpha + \mathbf{x}_2^\alpha\end{aligned}$$

- ▶ This is an element of  $V$  since it is the sum of two vectors which are each elements of  $V_1$  and  $V_2$ .

## Note: Limits Preserve Inequalities

- ▶ Consider the sequence  $\sum_{i=1}^l \widehat{\mathbf{x}}_i \rightarrow \sum_{i=1}^l \mathbf{x}_i$  where  $\mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) \geq c$ .
- ▶ We want to show that this inequality is preserved at the limit:  $\mathbf{p} \cdot \left( \sum_{i=1}^l \mathbf{x}_i \right) \geq c$ .
- ▶ Suppose toward a contradiction that instead  $\mathbf{p} \cdot \left( \sum_{i=1}^l \mathbf{x}_i \right) = d < c$ .
- ▶ From the definition of the limit of a function:

$$\lim_{\sum_{i=1}^l \widehat{\mathbf{x}}_i \rightarrow \sum_{i=1}^l \mathbf{x}_i} \mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) = d$$

implies that  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $\forall \sum_{i=1}^l \widehat{\mathbf{x}}_i, 0 < \left| \sum_{i=1}^l \widehat{\mathbf{x}}_i - \sum_{i=1}^l \mathbf{x}_i \right| < \delta$  implies that  $\left| \mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) - d \right| < \varepsilon$ .

- ▶ This holds for all  $\varepsilon > 0$ . Choose  $\varepsilon = c - d$ .  $\exists \delta > 0$  s.t.  $\forall \sum_{i=1}^l \widehat{\mathbf{x}}_i,$   
 $0 < \left| \sum_{i=1}^l \widehat{\mathbf{x}}_i - \sum_{i=1}^l \mathbf{x}_i \right| < \delta \implies \left| \mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) - d \right| < \varepsilon = c - d$ .
- ▶ But then:

$$-\varepsilon < \mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) - d < \varepsilon = c - d \implies \mathbf{p} \cdot \left( \sum_{i=1}^l \widehat{\mathbf{x}}_i \right) < c \implies \text{Contradiction!}$$