Supporting Prices and Convexity

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Overview of Chapter 1

- Theory of Constrained Maximization

 Why should we care about this?
- What is Economics?
- Economics is the study of how society manages its scarce resources (Mankiw, Ch.1)

 "Economics is the science which studies human behavior as a relationship between given ends and scarce means which have alternative uses." (Lionel Robbins, 1932)

Overview of Chapter 1

- Other Historical Accounts:
 - Economics is the "study of how societies use scarce resources to produce valuable commodities and distribute them among different people." (Paul A. Samuelson, 1948)
- I think Economics is the study of institutions
 & human behavior (reaction to institutions)
- Either way, constrained maximization is key!

Tools Introduced in Chapter 1

- 1. Supporting Hyperplanes (and Convexity)
- 2. First Order Conditions (Kuhn-Tucker)
- 3. Envelope Theorem
- But why do I need to know the math?
- When does Coase conjecture work?
 - It depends—Math makes these predictions precise
- What happens if you ignore the conditions required for theory to work? (Recall 2008/09!)

Publication Reward Problem

- Example: How should NTU reward its professors to publish journal articles?
 - Should NTU pay, say, NT\$300,000 per article published in Science or Nature?
- Well, it depends...
- Peek the answer ahead:
 - -Yes, if the production set is convex.
 - No, if, for example, there is initial increasing returns to scale.

Supporting Prices

- More generally,
- can prices and profit maximization provide appropriate incentives to induce all possible efficient production plans?
 - Is there a price vector that supports each efficient production plan?
- (Yes, but when?)
- Need some definitions first...

Production Plan and Production Set

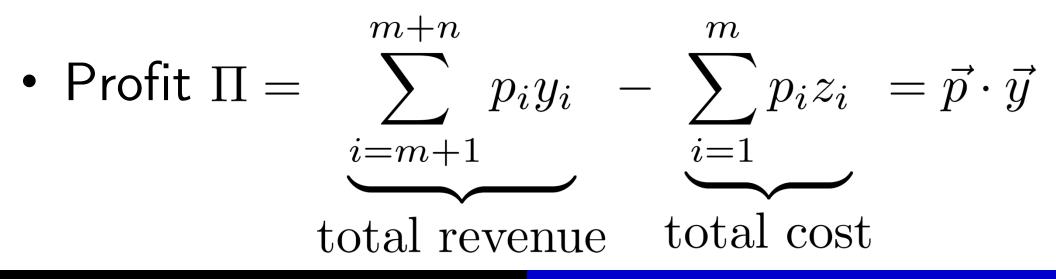
- A plant can:
- produce n outputs $\vec{q} = (q_1, \cdots, q_n)$
- using up to m inputs $\vec{z} = (z_1, \cdots, z_m)$
- Production Plan (\vec{z}, \vec{q})
- Production Set $\mathcal{Y} \subset \mathbb{R}^{m+n}_+$ = Set of all Feasible Production Plan
- Production Vector (treat inputs as negative)

$$\vec{y} = (-\vec{z}, \vec{q}) = (-z_1, \cdots, -z_m, q_1, \cdots, q_n)$$

Motivation Baseline Games Robustness Games Behavioral Theory

Production Set and Profits

- Production vector
- $\vec{y} = (y_1, \cdots, y_{m+n}) = (-z_1, \cdots, -z_m, q_1, \cdots, q_n)$
- Production Set $\mathcal{Y} \subset \mathbb{R}^{m+n}$ = Set of Feasible Production Plan
- Price vector $\vec{p} = (p_1, \cdots, p_{m+n})$



Motivation Baseline Games Robustness Games Behavioral Theory

EX: Production Function & Production Set

- A professor has 25 units of "brain-power"
- Allocates z_1 units to produce TSSCI papers
- Produce $q_1 = 4\sqrt{z_1}$ (Production Function)
- Production Set

$$\mathcal{Y}_1 = \left\{ (z_1, q_1) \middle| z_1 \ge 0, q_1 \le 4\sqrt{z_1} \right\}$$

- Treating inputs as negatives, $\vec{y} = (-\vec{z}, \vec{q})$
- Production Set is $\mathcal{V}_{i} = \int (a_{i}, a_{i}) = 16a_{i} = 16a_{i}$

$$\mathcal{V}_1 = \left\{ (y_1, y_2) \middle| -16y_1 - y_2^2 \ge 0 \right\}$$

Production Efficiency

- A production plan \vec{y} is wasteful if another plan in \mathcal{Y} achieves larger output with smaller input
- $\vec{\overline{y}}$ is production efficient (=non-wasteful) if

There is no
$$ec{y} \in \mathcal{Y}$$
 such that $ec{y} > ec{ec{y}}$

– Note:
$$\vec{y} \ge \vec{\overline{y}}$$
 if $y_j \ge \overline{y}_j$ for all j

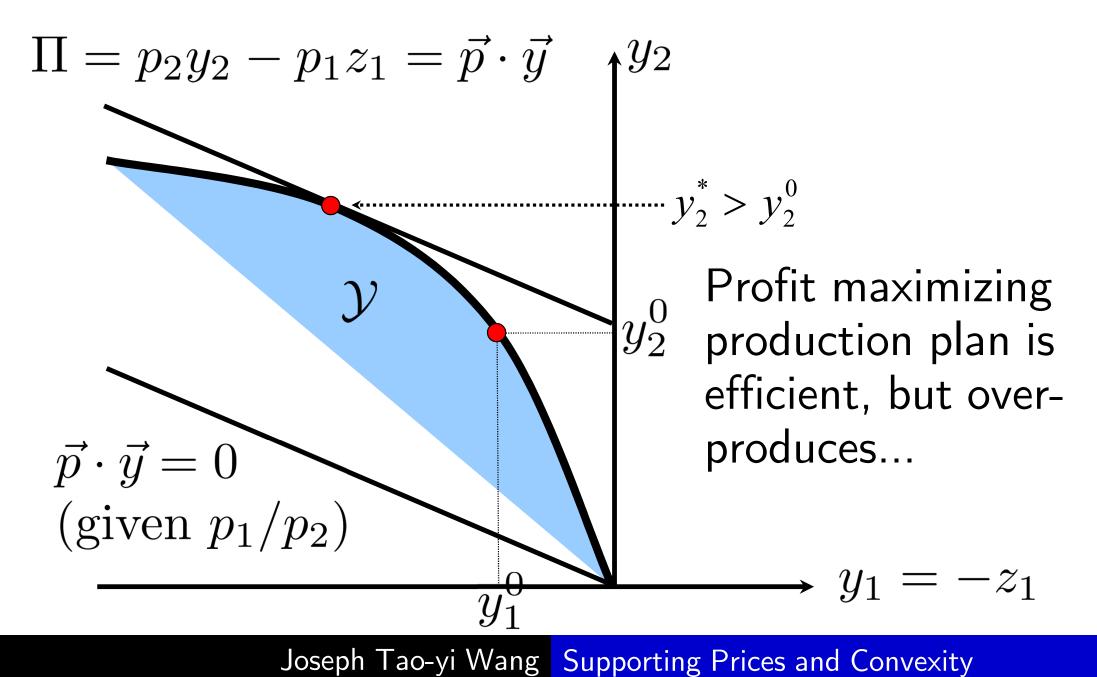
 $\vec{y} > \vec{\overline{y}}$ if inequality is strict for <u>some</u> *j*

 $\vec{y} \gg \vec{\overline{y}}$ if inequality is strict for <u>all</u> *j*

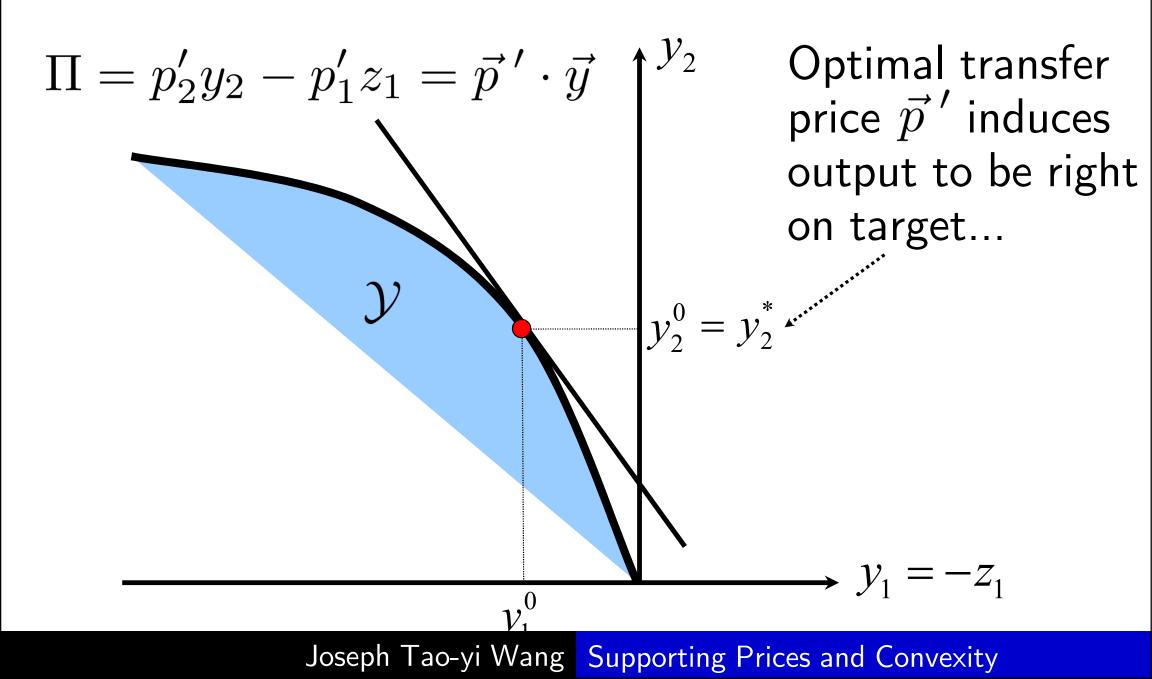
Can Prices Support Efficient Production?

- A professor has 25 units of "brain-power"
- Allocates y_1 units to produce TSSCI papers
- Price of brain-power is p_1
- Production Set \mathcal{Y}_1
- Can we induce production target y_2^0 ?
- With piece-rate prize p_2 ?

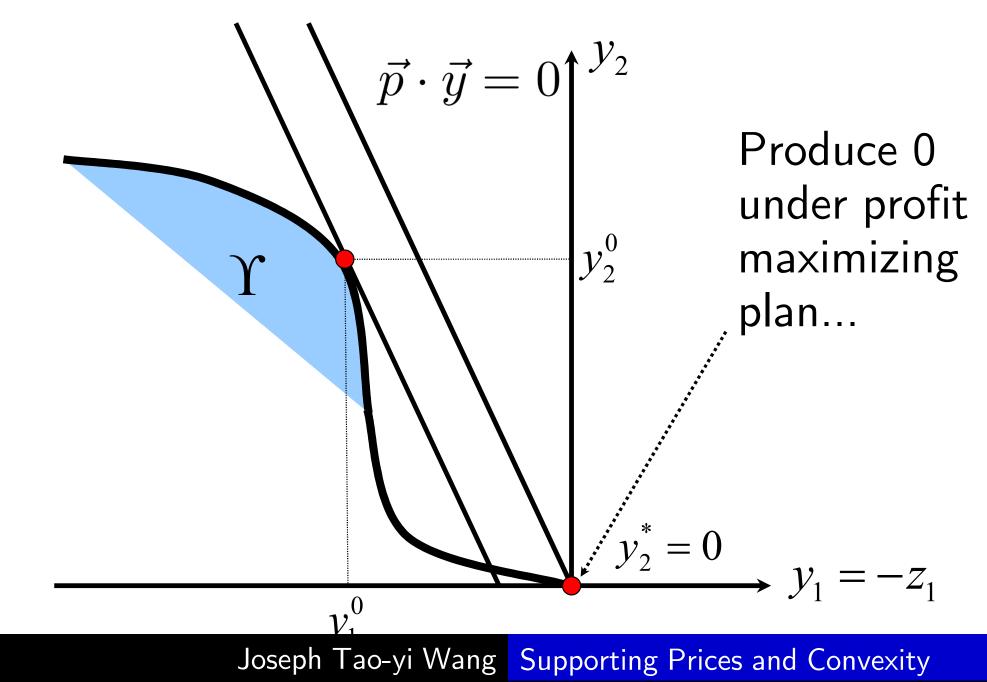
Can Prices Support Efficient Production?



Too High? Let's Lower the Transfer Price...



Will this Always Work?



What Made It Fail?

- The last production set was NOT convex.
- Production Set \mathcal{Y}_1 is convex if for any \vec{y}^0, \vec{y}^1
- Its convex combination (for $0 < \lambda < 1$)

 $\vec{y}^{\lambda} = (1 - \lambda)\vec{y}^0 + \lambda\vec{y}^1 \in \mathcal{Y}_1$

- (is also in the production set)

 Is it true that we can use prices to guide production decisions as long as production sets are convex?

Supporting Hyperplane Theorem

Proposition 1.1-1 (Supporting Hyperplane)

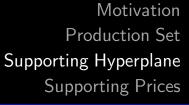
- Suppose $\mathcal{Y} \subset \mathbb{R}^n$ is non-empty and convex,
- And \vec{y}^0 lies on the boundary of γ
- Then, there exists $\vec{p} \neq 0$ such that
- i. For all $\vec{y} \in \mathcal{Y}$, $\vec{p} \cdot \vec{y} \leq \vec{p} \cdot \vec{y}^0$
- ii. For all $\vec{y} \in \operatorname{int} \mathcal{Y}, \ \vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$

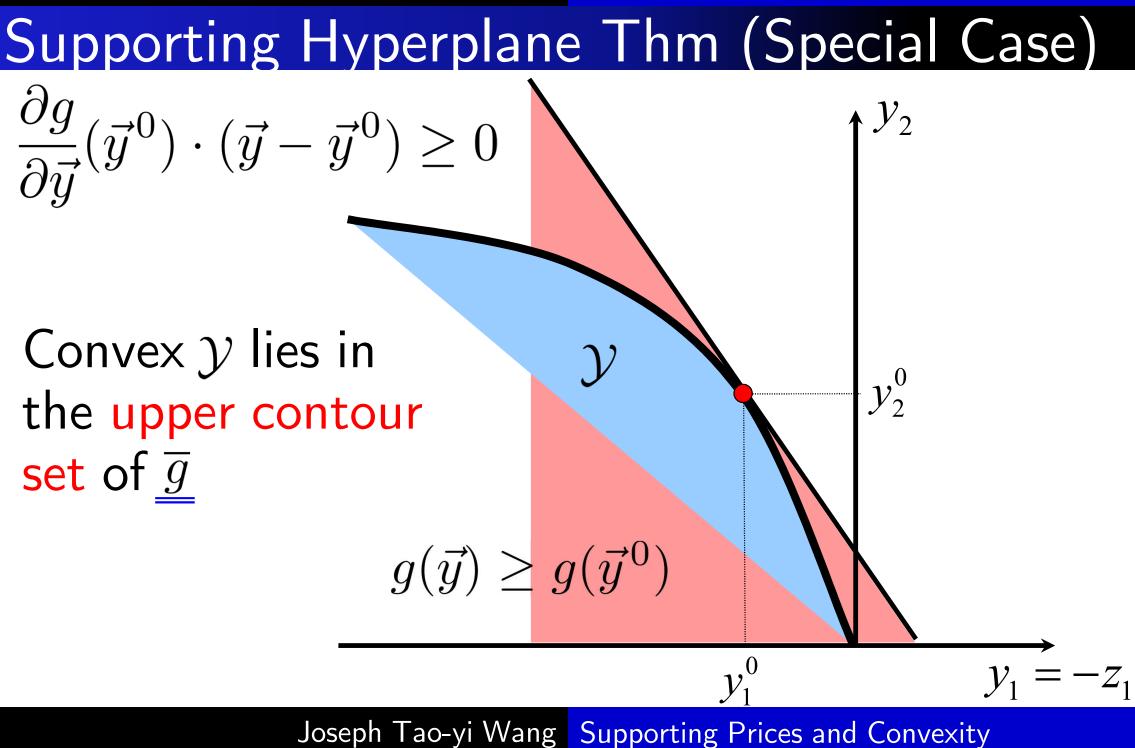
- Can we obtain part (ii)???

• Proof: For the general case, see Appendix C.

Supporting Hyperplane Thm (Special Case)

- Consider special case where
- Production set \mathcal{Y} is the upper contour set $\mathcal{Y} = \{\vec{y}|g(\vec{y}) \ge g(\vec{y}^0)\}, g \text{ is differentiable}$
- Suppose the gradient vector is non-zero at \vec{y}^0
- The linear approximation of g at \vec{y}^0 is: $\underline{\overline{g}(\vec{y})} = g(\vec{y}^0) + \frac{\partial g}{\partial \vec{y}}(\vec{y}^0) \cdot (\vec{y} - \vec{y}^0)$
- If $\mathcal Y$ is convex, it lies in upper contour set of $\underline{\overline{g}}$





Special Case of Supporting Hyperplane Thm

- Lemma 1.1-2
- If g is differentiable and $\mathcal{Y}=\left\{\vec{y}|g(\vec{y})\geq g(\vec{y}^0)\right\}$ is convex, then

$$\vec{y} \in \mathcal{Y} \Rightarrow \frac{\partial g}{\partial \vec{y}}(\vec{y}^0) \cdot (\vec{y} - \vec{y}^0) \ge 0$$

- This tells us how to calculate the supporting prices (under this special case):
- For boundary point \vec{y}^0 , choose $\vec{p} = -\frac{\partial g}{\partial \vec{x}}(\vec{y}^0)$

From Lemma to Supporting Hyperplane Thm

- If g is differentiable and $\mathcal{Y} = \{\vec{y}|g(\vec{y}) \ge g(\vec{y}^0)\}$ is convex, then set $\vec{p} = -\frac{\partial g}{\partial \vec{y}}(\vec{y}^0)$
- By lemma: $\vec{y} \in \mathcal{Y} \implies -\vec{p} \cdot (\vec{y} \vec{y}^0) \ge 0$ $\implies \vec{p} \cdot \vec{y} \le \vec{p} \cdot \vec{y}^0$
- This gives us part (i) of S. H. T.
 What about part (ii)? See Prop. 1-1.3...

Supporting Hyperplane Theorem

Proposition 1.1-1 (Supporting Hyperplane)

- Suppose $\mathcal{Y} \subset \mathbb{R}^n$ is non-empty and convex,
- And \vec{y}^0 lies on the boundary of γ
- Then, there exists $\vec{p} \neq 0$ such that
- i. For all $\vec{y} \in \mathcal{Y}$, $\vec{p} \cdot \vec{y} \leq \vec{p} \cdot \vec{y}^0$
- Proof: For the general case, see Appendix C.

From Lemma to Supporting Hyperplane Thm

- If g is differentiable and $\mathcal{Y} = \left\{ \vec{y} | g(\vec{y}) \ge g(\vec{y}^0) \right\}$ is convex, then $\vec{y} \in \mathcal{Y} \implies -\vec{p} \cdot (\vec{y} - \vec{y}^0) \ge 0$ $\implies \vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$
- Attempt part (ii) of S. H. T.
- Note: $\vec{y} \in \operatorname{int} \mathcal{Y} \Rightarrow \exists \vec{y}' = \vec{y} + \vec{\epsilon} \in \mathcal{Y}, \vec{\epsilon} \gg 0$
- And $\vec{p} \cdot \vec{y}' = \vec{p} \cdot \vec{y} + \vec{p} \cdot \vec{\epsilon} \le \vec{p} \cdot \vec{y}^0$
- If $\vec{p} \cdot \vec{\epsilon} > 0 \Rightarrow \vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$
 - Why is this the case? See Prop. 1-1.3...

Supporting Hyperplane Theorem

Proposition 1.1-1 (Supporting Hyperplane)

- Suppose $\mathcal{Y} \subset \mathbb{R}^n$ is non-empty and convex,
- And \vec{y}^0 lies on the boundary of γ
- Then, there exists $\vec{p} \neq 0$ such that
- i. For all $\vec{y} \in \mathcal{Y}$, $\vec{p} \cdot \vec{y} \leq \vec{p} \cdot \vec{y}^0$
- ii. For all $\vec{y} \in \operatorname{int} \mathcal{Y}, \ \vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$
- Proof: For the general case, see Appendix C.

Proof of Lemma 1.1-2

• If g is differentiable and $\mathcal{Y}=\left\{\vec{y}|g(\vec{y})\geq g(\vec{y}^0)\right\}$ is convex, then

$$\vec{y} \in \mathcal{Y} \Rightarrow \frac{\partial g}{\partial \vec{y}}(\vec{y}^0) \cdot (\vec{y} - \vec{y}^0) \ge 0$$

– Proof:

- For $\vec{y} \in \mathcal{Y}$, consider $\vec{y}^{\lambda} = (1 \lambda)\vec{y}^0 + \lambda\vec{y} \in \mathcal{Y}$
- So, $g(\vec{y}^{\lambda}) g(\vec{y}^{0}) \ge 0$
- Define $h(\lambda) \equiv g(\vec{y}^{\lambda}) = g(\vec{y}^0 + \lambda(\vec{y} \vec{y}^0))$

Proof of Lemma 1.1-2

$$\begin{split} \vec{y} \in \mathcal{Y} &\Rightarrow \vec{y}^{\lambda} = (1 - \lambda)\vec{y}^{0} + \lambda\vec{y} \in \mathcal{Y} \\ h(\lambda) \equiv g(\vec{y}^{\lambda}) = g(\vec{y}^{0} + \lambda(\vec{y} - \vec{y}^{0})) & \text{dh} \\ \underline{h(\lambda) - h(0)} &= \underbrace{g((\vec{y}^{0} + \lambda(\vec{y} - \vec{y}^{0})) - g(\vec{y}^{0})}_{> 0} \end{split}$$

• By Lemma. Therefore, by chain rule:

$$\frac{dh}{d\lambda}(\lambda) \Big|_{\lambda=0} = \frac{\partial g}{\partial \vec{y}} (\vec{y}^0 + \lambda(\vec{y} - \vec{y}^0)) \cdot (\vec{y} - \vec{y}^0) \Big|_{\lambda=0}$$
$$= \frac{\partial g}{\partial \vec{y}} (\vec{y}^0) \cdot (\vec{y} - \vec{y}^0) \ge 0. \square$$

Example

- A professor has z=25 units of "brain-power"
- Allocates z_2 units to produce TSSCI papers
- Produce $y_2 = 2\sqrt{z_2}$ number of TSSCI papers
- Allocates z_3 units to produce SSCI papers
- Produce $y_3 = \sqrt{z_3}$ number of SSCI papers
- Set of feasible plans is $(y_1 = -z)$

$$\mathcal{Y} = \left\{ \vec{y} \middle| g(\vec{y}) = -y_1 - \frac{1}{4}y_2^2 - y_3^2 \ge 0 \right\}$$

Example

– Professor W is working at full capacity

- Professor W's output is $\vec{y}^0 = (-25, 8, 3)$ - 8 TSSCI papers and 3 SSCI papers!
- What reward scheme can support this? $\vec{p} = -\frac{\partial g}{\partial \vec{y}}(\vec{y}^0) = (1, \frac{1}{2}y_2^0, 2y_3^0) = (1, \underline{4}, \underline{6})$
- To instead induce $(y_2^1, y_3^1) = (2, 2\sqrt{6}) \approx (2, 5)$ - Approx. 2 TSSCI papers and 5 SSCI papers $\vec{p} = (1, \frac{1}{2}y_2^1, 2y_3^1) = (1, 1, 4\sqrt{6}) \approx (1, \underline{1, 10})$

Positive Prices (Free Disposal)

- Supporting Hyperplane theorem has economic meaning if <u>prices are positive</u>
 - Need another assumption
- Free Disposal
- For any feasible production $\mathsf{plan}\,ec y\in\mathcal{Y}$ and any
- $\vec{\delta} > 0$, the production plan $\vec{y} \vec{\delta}$ is also feasible

Supporting Prices

• With free disposal, we can prove:

Proposition 1.1-3 (Supporting Prices)

- If \vec{y}^0 is a boundary point of a convex set $\mathcal Y$
- And the free disposal assumption holds,
- Then, there exists a price vector $\vec{p} > \vec{0}$ such
- that $\vec{p} \cdot \vec{y} \leq \vec{p} \cdot \vec{y}^0$ for all $\vec{y} \in \mathcal{Y}$
- Moreover, if $\vec{0} \in \mathcal{Y}$, then $\vec{p} \cdot \vec{y}^0 \geq 0$
- Finally, for all $\vec{y} \in \operatorname{int} \mathcal{Y}$, $\vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$ part (ii)

Motivation Baseline Games Robustness Games Behavioral Theory

Supporting Prices

- Proof: Supporting Hyperplane Theorem says:
- There is some $\vec{p} \neq \vec{0}$ such that, for all $\vec{y} \in \mathcal{Y}$,
- $\vec{p} \cdot (\vec{y}^0 \vec{y}) \ge 0$. Now need to show $\underline{p_i} \ge 0$
- By free disposal, $\vec{y}' = \vec{y}^0 \vec{\delta} \in \mathcal{Y}, \forall \vec{\delta} > \vec{0}$
- Set $\vec{\delta} = (1, 0, \cdots, 0), \vec{p} \cdot (\vec{y}^0 \vec{y}') = p_1 \ge 0$
- Set $\vec{\delta} = (0, 1, 0, \cdots), \ \vec{p} \cdot (\vec{y}^0 \vec{y}') = p_2 \ge 0$
- ... • Set $\vec{\delta} = (0, \cdots, 0, 1), \vec{p} \cdot (\vec{y}^0 - \vec{y}') = p_n \ge 0$

Supporting Prices

- Since $\vec{p} \cdot \vec{y} \leq \vec{p} \cdot \vec{y}^0$ for all $\vec{y} \in \mathcal{Y}$, if $\vec{0} \in \mathcal{Y}$
- Set $\vec{y} = \vec{0}$ and we have $\vec{p} \cdot \vec{y}^0 \ge 0$
- Claim: For all $\vec{y} \in \operatorname{int} \mathcal{Y}$, $\vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$ part (ii)
- For $\vec{y} \in \operatorname{int} \mathcal{Y} \implies \exists \vec{y}' = \vec{y} + \vec{\epsilon} \in \mathcal{Y}, \vec{\epsilon} \gg 0$
- And $\vec{p} \cdot \vec{y}' = \vec{p} \cdot \vec{y} + \vec{p} \cdot \vec{\epsilon} \le \vec{p} \cdot \vec{y}^0$
- Since $\underline{\vec{p}} > 0$, we have $\vec{p} \cdot \vec{\epsilon} > 0 \Rightarrow \vec{p} \cdot \vec{y} < \vec{p} \cdot \vec{y}^0$

Back to Publication Rewards

- Should NTU really pay NT\$300,000 per article published in Science or Nature?
 - Is the production set for Science/Nature convex?
- What would be a better incentive scheme to encourage publications in Science/Nature?
 - Efficient Wages (High Fixed Wages)?
 - Tenure?
 - Endowed Chair Professorships?

Back to Publication Rewards

- What are some tasks do you expect piecerate incentives to work?
 - Sales
 - Real estate agents
- What about a fixed payment?
 - Secretaries and Office Staff
 - Store Clerk
- What about other incentives schemes?
 That's for you to answer (in contract theory)!

Summary of 1.1

- Input = Negative Output
- Vector space of \vec{y}
- Convexity (quasi-concavity) is the key for supporting prices (=linearization)
- What is a good incentive scheme to induce professor to publish in Science and Nature?
- Consumer = Producer
- Homework: Exercise 1.1-4 (Optional: 1.1-6)

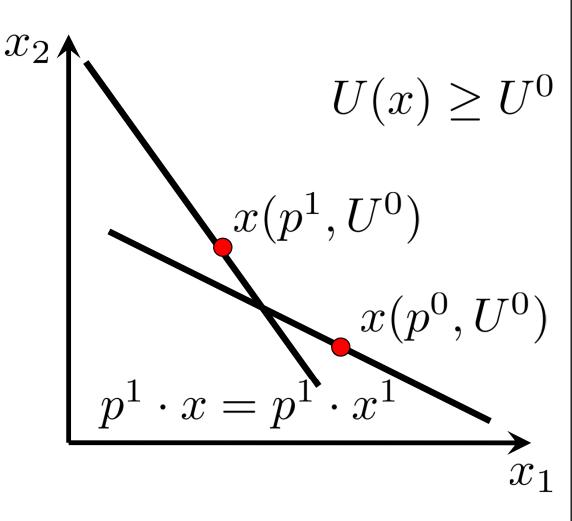
Another Example: Linear Model

- What if firm has n plants producing the same product q using m inputs $z = (z_1, \dots, z_m)$?
- Need to consider activity level x_j for plant j
 Produce output a_{0j}x_j with input a_{ij}x_j, i = 1, ..., m

• Total output
$$\sum_{j=1}^{n} a_{0j} x_j$$
; Total input $\sum_{j=1}^{n} a_{ij} x_j, \forall i$

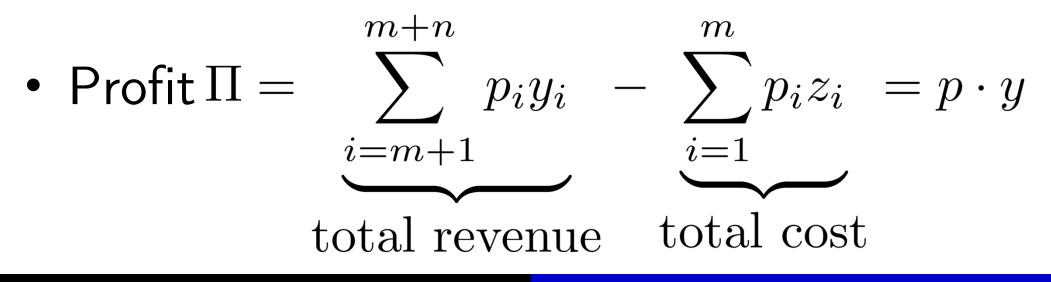
• Linear Production Set (convex, free disposal) $\mathcal{Y} = \{(-z,q) | x \ge 0, q \le a_0 \cdot x, Ax \le z\}$

Another Example: Linear Model



Production Set and Profits

- Production vector $y = (y_1, \cdots, y_{m+n}) = (-z_1, \cdots, -z_m, q_1, \cdots, q_n)$
- Production $\operatorname{Set} \mathcal{Y} \subset \mathbf{R}^{m+n}$ =Set of Feasible Production Plan
- Price vector $p = (p_1, \cdots, p_{m+n})$



Quasi-Concavity

f is quasi-concave if the upper contour set of *f* set are convex. Equivalently, for any y⁰, y¹ and convex combination

$$y^{\lambda} = \lambda \cdot y^{0} + (1 - \lambda) \cdot y^{1},$$

$$f(y^{\lambda}) \ge \min \left\{ f(y^{0}), f(y^{1}) \right\}.$$

- Why is this useful?
 - Because we have...

Separating Hyperplane Theorem

• Proposition 1.1-2:

Suppose *S* and *T* are convex sets with a common boundary point $s^0 = t^0$ and no common interior points. Then there is some *p* such that, for all $s \in S$ and $t \in T$, $p \cdot s \leq p \cdot t$.

(Inequality strict if either *s* or *t* is an interior.)

Separating Hyperplane Theorem

• Proof of Proposition 1.1-2: Define $\Upsilon = S - T$, then $s^0 - t^0 = 0 \in \Upsilon$ If Υ is convex (verify this!!!), then... Supporting Hyperplane Theorem says: there is some $p \neq 0$ such that, for all $y \in \Upsilon$, $p \cdot y \leq p \cdot (s^0 - t^0) = 0.$ Since y = s - t for some $s \in S, t \in T$, $p \cdot s \leq p \cdot t$ for all $s \in S, t \in T$.