

# Lecture 10: Lagrangians (cont'd) and Expenditure Minimization

## 1 Where are we?

- Last time, we stated the consumer problem,

$$\max u(x) \quad \text{subject to} \quad p \cdot x \leq w \quad \text{and} \quad x \geq 0$$

and introduced an auxiliary function, the Lagrangian,

$$\mathcal{L}(x, \lambda, \mu) = u(x) + \lambda(w - p \cdot x) + \mu \cdot x$$

defined for  $x \in \mathbb{R}^k$  and  $\lambda, \mu \geq 0$ ;

- And we showed that if  $(x^*, \lambda^*, \mu^*)$  is a saddle point of the Lagrangian –

$$\mathcal{L}(x, \lambda^*, \mu^*) \leq \mathcal{L}(x^*, \lambda^*, \mu^*) \leq \mathcal{L}(x^*, \lambda, \mu)$$

for every  $x \in \mathbb{R}^k$  and every  $\lambda, \mu \geq 0$  –

then  $x^*$  is a solution to the Consumer Problem;

and vice versa if  $u$  is differentiable and concave

- Today, we'll see how to actually use this stuff –  
the Kuhn-Tucker Conditions, examples, and onward

- First... any questions?

## 2 The Kuhn-Tucker Conditions

- So, we know that the solutions to the Consumer Problem pretty much correspond to the saddle points of the Lagrangian
- Next, let's think about what conditions  $(x^*, \lambda^*, \mu^*)$  must satisfy to be a saddle point, and therefore for  $x^*$  to solve the consumer problem
- First, we need  $x^*$  to be a solution to

$$\max_{x \in \mathbb{R}^k} \mathcal{L}(x, \lambda^*, \mu^*) = \max_{x \in \mathbb{R}^k} \{u(x) + \lambda^*(w - p \cdot x) + \mu^* \cdot x\}$$

- We're thinking of  $\mathcal{L}$  as defined for all  $x$  in  $\mathbb{R}^k$ , not just  $\mathbb{R}_+^k$ , so  $x^*$  must be an interior point;

if  $u$  is differentiable,  $\mathcal{L}$  is differentiable w.r.t.  $x$ , so we'll need the FOC to hold:

$$\frac{\partial \mathcal{L}}{\partial x_i}(x^*, \lambda^*, \mu^*) = 0 \quad \longrightarrow \quad \frac{\partial u}{\partial x_i}(x^*) - \lambda^* p_i + \mu_i^* = 0$$

- (If this was either positive or negative,  $x^*$  couldn't be maximizing  $\mathcal{L}$ )
- And second, we need  $(\lambda^*, \mu^*)$  to solve

$$\min_{\lambda, \mu \geq 0} \mathcal{L}(x^*, \lambda, \mu) = \min_{\lambda, \mu \geq 0} \{u(x^*) + \lambda(w - p \cdot x^*) + \mu \cdot x^*\}$$

- This means we need  $w - p \cdot x^* \geq 0$  – otherwise, we could always reduce  $\mathcal{L}$  by increasing  $\lambda$  – and we need  $\lambda^* = 0$  if  $w - p \cdot x^* > 0$ , otherwise we could reduce  $\mathcal{L}$  by reducing  $\lambda$
- And similarly, we need  $x_i^* \geq 0$  – if  $x_i^* < 0$ , we could reduce  $\mathcal{L}$  by increasing  $\mu_i$  – and we need  $\mu_i^* = 0$  if  $x_i^* > 0$ , or else we could reduce  $\mathcal{L}$  by reducing  $\mu_i$
- So basically, for  $(\lambda^*, \mu^*)$  to minimize  $\mathcal{L}$ , we need for  $x^*$  to be feasible, and we need complementary slackness to hold

- And of course, we defined  $\mathcal{L}$  for non-negative values of the Lagrange multipliers, so we can think of  $\lambda \geq 0$  and  $\mu \geq 0$  as additional constraints
- That gives us a bunch of necessary conditions for  $(x^*, \lambda^*, \mu^*)$  to be a saddle point:
  1.  $\frac{\partial u}{\partial x_i}(x^*) - \lambda^* p_i + \mu_i^* = 0$  for each  $i$  (FOC w.r.t.  $x$ )
  2.  $p \cdot x^* \leq w$ ,  $x_i^* \geq 0$  for each  $i$  (original constraints on  $x$ )
  3.  $\lambda^* \geq 0$  and  $\mu_i^* \geq 0$  for each  $i$  (non-negative Lagrange multipliers)
  4.  $\lambda^*(w - p \cdot x^*) = 0$  and  $\mu_i^* x_i^* = 0$  for each  $i$  (complementary slackness)
- These are collectively known as the **Kuhn-Tucker Conditions**
- As expected, these are necessary conditions for a solution to the consumer problem, and also sufficient under some additional assumptions:

**Theorem 1.** *Assume  $u$  is differentiable and  $(p, w) \gg 0$ . If  $x^* > 0$  solves the Consumer Problem, then there exist  $\lambda^* \geq 0$  and  $\mu^* \geq 0$  such that  $(x^*, \lambda^*, \mu^*)$  satisfy the Kuhn-Tucker conditions.<sup>1</sup>*

**Theorem 2.** *Suppose  $u$  is differentiable and concave.<sup>2</sup> Then if  $(x^*, \lambda^*, \mu^*)$  satisfies the Kuhn-Tucker conditions,  $x^*$  solves the Consumer Problem.*

- Putting these results together with what we saw last time, if we let CP mean “ $x^*$  solves the consumer problem,”  $\mathcal{L}$  mean “ $(x^*, \lambda^*, \mu^*)$  is a saddlepoint of  $\mathcal{L}$  for some  $\lambda^*$  and  $\mu^*$ ,” and KT mean “ $(x^*, \lambda^*, \mu^*)$  satisfy the Kuhn-Tucker conditions for some  $\lambda^*$  and  $\mu^*$ ,” we have that if  $(p, w) \gg 0$ ...
  - $\mathcal{L} \rightarrow CP$  with no further restrictions
  - $CP \rightarrow \mathcal{L}$  if  $u$  is differentiable and concave
  - $CP \rightarrow KT$  if  $u$  is differentiable
  - $KT \rightarrow CP$  if  $u$  is differentiable and concave

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<sup>1</sup>For a more general constrained optimization problem, there is one additional restriction which would have to hold at  $x^*$ , in order to guarantee the Kuhn-Tucker conditions can be satisfied. This is called *constraint qualification*, and it basically says that whichever of the constraints of the problem bind at  $x^*$ , their gradients need to be linearly independent, i.e., there can’t be any locally “redundant” constraints. Given the constraints in our problem (budget and non-negativity), this condition automatically holds whenever  $p \gg 0$  and  $w > 0$ , so I don’t state it in the theorem; this is just something you need to know if you want to extend these techniques to other problems. See MWG page 956-964 for the more general treatment of constrained optimization.

<sup>2</sup>We can weaken concavity to (i) quasi-concavity and (ii) for any  $x, x' \in X$ ,  $u(x') > u(x)$  implies  $\nabla u(x) \cdot (x' - x) > 0$ , both of which hold when  $u$  is concave.

### 3 What's this good for?

- So, if  $u$  is differentiable,  
the Kuhn-Tucker conditions will find all the solutions to the Consumer Problem,  
although if  $u$  is not concave, it may find some points that aren't solutions too
- But, even if  $u$  is not concave  
(so the KT conditions aren't automatically sufficient),  
the KT conditions will often have a unique solution;  
since the CP must have a solution, and it must satisfy KT,  
if the KT conditions have a unique solution, we know it's the unique solution to the CP
- Let's use this to solve one example
- Let  $X = \mathbb{R}_+^3$ ,  $u(x_1, x_2, x_3) = x_1^{\alpha_1} x_2^{\alpha_2} x_3^{\alpha_3}$ , a generic price vector  $(p_1, p_2, p_3) \gg 0$ , and  $w > 0$
- First, it's helpful to try to eliminate some constraints
- Note that if any consumption quantity is 0, then  $u = 0$ , so this can't be optimal –  
so  $x^* \gg 0$ , which means (for complementary slackness)  $\mu^* = 0$ ,  
so we can ignore the non-negativity constraints
- (For many utility functions,  
the marginal utility of each good is infinite when you're consuming none of it –  
the first little bit of each good is extremely valuable –  
and in this case, it's never optimal to consume 0 of anything given positive wealth,  
and you can just solve the problem without the nonnegativity constraints)

- Next, rather than  $u = x_1^{\alpha_1} x_2^{\alpha_2} x_3^{\alpha_3}$ , let's maximize  $\ln u$ , since this is another utility function representing the same preferences, so we'll use

$$\tilde{u}(x, y, z) = \alpha_1 \log x_1 + \alpha_2 \log x_2 + \alpha_3 \log x_3$$

- The Lagrangian is therefore

$$\mathcal{L} = \alpha_1 \log x_1 + \alpha_2 \log x_2 + \alpha_3 \log x_3 + \lambda(w - p \cdot x)$$

and the Kuhn-Tucker FOC are therefore

$$\frac{\partial \mathcal{L}}{\partial x_i} = 0 \quad \longrightarrow \quad \frac{\alpha_i}{x_i} - \lambda p_i = 0 \quad \longrightarrow \quad x_i = \frac{\alpha_i}{\lambda p_i}$$

for each good  $i$

- Next, since  $u$  is strictly increasing everywhere, the budget constraint will hold with equality
- (We haven't yet proved Walras' Law – that's what I get for changing the order – but since  $u$  is strictly increasing in every good, the consumer would never choose not to spend the whole budget<sup>3</sup>)
- So

$$w = \sum_{j=1}^3 p_j x_j = \sum_{j=1}^3 p_j \frac{\alpha_j}{\lambda p_j} = \frac{\alpha_1 + \alpha_2 + \alpha_3}{\lambda}$$

so  $\lambda = \frac{\alpha_1 + \alpha_2 + \alpha_3}{w}$ , and therefore

$$x_i = \frac{\alpha_i}{\lambda p_i} = \frac{\alpha_i}{\alpha_1 + \alpha_2 + \alpha_3} \frac{w}{p_i}$$

or

$$p_i x_i = \frac{\alpha_i}{\alpha_1 + \alpha_2 + \alpha_3} w$$

- Since this is the only way to satisfy the K-T conditions, we know this is the (unique) solution to the consumer problem!
- (also note that we didn't need concavity of  $u$  – we never even specified that  $\alpha_1 + \alpha_2 + \alpha_3 \leq 1$ !)

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<sup>3</sup>Or to be more formal, recall that for any  $j$  with  $x_j^* > 0$ ,  $\lambda^* = \frac{1}{p_j} \frac{\partial u}{\partial x_j}(x^*) > 0$ ; and once we know  $\lambda^* \neq 0$ , then for complimentary slackness to hold,  $w - p \cdot x^* = 0$ .

### 3.1 one other thing about Lagrangians

- Now,  $(x^*, \lambda^*, \mu^*)$  being a saddle point of  $\mathcal{L}$  is really two separate conditions:
  - given  $\lambda^*$  and  $\mu^*$ ,  $x^*$  maximizes  $\mathcal{L}$
  - and given  $x^*$ , the multipliers  $(\lambda^*, \mu^*)$  minimize  $\mathcal{L}$
- And we've seen that the latter condition is really equivalent to two things: feasibility of  $x^*$ , and complementary slackness
- So when I learned this stuff in grad school, I didn't learn it as saddle points, I learned it as, you solve a constrained optimization problem by solving

$$\max_x \{u(x) + \lambda(w - p \cdot x) + \mu \cdot x\}$$

and imposing the additional conditions  $p \cdot x \leq w$ ,  $x \geq 0$ ,  $\lambda(w - p \cdot x) = 0$ , and  $\mu \cdot x = 0$

- As we saw, the FOC from maximization over  $x$ , combined with these other conditions (and non-negativity of the Lagrange multipliers), are exactly the Kuhn-Tucker conditions

### 3.2 end of Lagrangians

- Before we move on, I'll repeat two cautionary points I made earlier
- First of all, while this is a great way to solve the consumer problem when  $u$  is differentiable, please don't think of it as a cookbook, where you get a problem and immediately plug it into the Kuhn-Tucker conditions without thinking about it
- There's a lot to gain by thinking a little about the problem before you start to solve it – whether to expect a corner solution, an interior solution, and other things like that
- And finally, if you use this for other constrained optimization problems beyond CP, there's one additional restriction that has to hold – called “constraint qualification” – which we didn't go into.<sup>4</sup>

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<sup>4</sup>Roughly, constraint qualification requires that all the constraints that bind at the same point have gradients at that point that are not linear combinations of each other – in some sense, there are no locally “redundant” constraints. With the Consumer Problem, this holds automatically, since the gradient of  $w - p \cdot x$  is  $(-p_1, -p_2, \dots, -p_k)$  and the gradient of  $x_i$  is  $(0, \dots, 0, 1, 0, \dots, 0)$ . But if you were maximizing a function subject to the constraints  $x_1 \geq 0$  and  $-x_1 + (x_2 - 3)^2 \geq 0$ , you'd be in trouble, because at the point  $(0, 3)$ , the two constraints both bind and have gradients  $(1, 0)$  and  $(-1, 0)$ . See MWG section M.K in the math appendix for more detail.

## 4 jump back – a few properties of demand

- We defined  $x(p, w)$  as the solution to the consumer problem, or Marshallian demand
- **Proposition.** Marshallian demand is homogeneous of degree zero:  $x(\lambda p, \lambda w) = x(p, w)$  for any  $p \gg 0$ ,  $w > 0$ ,  $\lambda > 0$ .
  - Proof: we already showed  $B(\lambda p, \lambda w) = B(p, w)$ , so  $x(\lambda p, \lambda w)$  and  $x(p, w)$  are literally the solutions to the same optimization problem.
- Again, the problem is invariant to the unit of currency
- **Proposition [Walras' Law].** If preferences are locally non-satiated, then for any  $(p, w)$  and any  $x \in x(p, w)$ ,  $p \cdot x = w$ .

– Suppose not, i.e., suppose  $p \cdot x < w$  for some  $x \in x(p, w)$

– If we make  $\varepsilon$  small enough, then for every  $y$  within  $\varepsilon$  of  $x$ ,  $p \cdot y \leq w$

\* specifically, let  $\varepsilon = (w - p \cdot x) / \|p\|$

\* then

$$p \cdot y = p \cdot (x + (y - x)) = p \cdot x + p \cdot (y - x)$$

\* since  $a \cdot b = \|a\| \|b\| \cos \theta \leq \|a\| \|b\|$  (where  $\theta$  is the angle between vectors  $a$  and  $b$ ), this is

$$\leq p \cdot x + \|p\| \|y - x\| \leq p \cdot x + \|p\| \frac{w - p \cdot x}{\|p\|} = w$$

so if  $\|y - x\| \leq \varepsilon$ ,  $p \cdot y \leq w$

– but if preferences are LNS, then there must be some  $y$  within  $\varepsilon$  of  $x$  such that  $y \succ x$

– since we just showed that  $y$  was affordable, this contradicts  $x$  being the maximizer

- so if preferences are LNS, optimizing consumers always spend their entire budget;
- so if prefs are LNS, we can rewrite the consumer problem as

$$\max u(x) \quad s.t. \quad p \cdot x = w$$

- We said that  $x(p, w)$  could be a set, i.e., the consumer problem could have multiple solutions

- **Proposition.**

1. If preferences are convex, then for each  $p$  and  $w$ ,  $x(p, w)$  is a convex set
2. If preferences are strictly convex, then for each  $p$  and  $w$ ,  $x(p, w)$  is a single point, i.e.,  $x$  is a function.

– Proof.  $x(p, w)$  is a convex set if for any  $x$  and  $x' \in x(p, w)$ ,  $tx + (1 - t)x' \in x(p, w)$  as well. So let's show it.

– If  $x$  and  $x'$  are both in  $B(p, w)$ , then  $p \cdot x \leq w$  and  $p \cdot x' \leq w$ , so

$$p \cdot (tx + (1 - t)x') = t(p \cdot x) + (1 - t)(p \cdot x') \leq tw + (1 - t)w = w$$

so  $tx + (1 - t)x'$  is in the budget set as well

– And if  $x$  and  $x'$  are in  $x(p, w)$ , then for any other  $y$  in the budget set,  $x \succsim y$  and  $x' \succsim y$ ; if preferences are convex, this means  $tx + (1 - t)x' \succsim y$  as well, so  $tx + (1 - t)x' \in x(p, w)$

– for the second part, if preferences are strictly convex, suppose there were two separate points  $x$  and  $x'$  in  $x(p, w)$

– since  $x \succ x'$  and  $x' \succ x'$ , if preferences are strictly convex, for any  $t \in (0, 1)$ ,

$$tx + (1 - t)x' \succ x'$$

– as we argued above,  $tx + (1 - t)x'$  is in the budget set, so this contradicts  $x'$  being optimal

- so now we know that Marshallian demand...

– exists

– is homogeneous of degree 0

– exhausts the budget if preferences are LNS

– is single-valued if preferences are strictly convex

- onward!



## 5 The Indirect Utility Function

- We stated the consumer problem as

$$\max u(x) \quad \text{subject to} \quad x \in B(p, w)$$

and called the solution  $x(p, w)$ ; now define  $v(p, w)$  as the value function, i.e.,

$$v(p, w) = \max u(x) \quad \text{subject to} \quad x \in B(p, w)$$

- by definition,  $v(p, w) = u(x)$  for any  $x \in x(p, w)$  –  
 $v$  is the “utility value” of having wealth  $w$  and facing prices  $p$
- note that  $v : \mathbb{R}_{++}^k \times \mathbb{R}_+ \rightarrow \mathbb{R}$  is a function:  
if the CP has multiple solutions, they all give the same utility, so there’s a unique max value
- $v$  is called the *indirect utility function* (because it’s a function of  $p$  and  $w$ , not consumption)
- Some properties of  $v$ :
- **Proposition.** Suppose  $u$  is a continuous utility function representing a LNS preference relation  $\succsim$  on  $\mathbb{R}_+^k$ . Then  $v(p, w)$  is...
  1. homogeneous of degree 0:  $v(\lambda p, \lambda w) = v(p, w)$
  2. continuous on  $\{(p, w) : p \gg 0, w \geq 0\}$
  3. nonincreasing in  $p$ , and strictly increasing in  $w$
  4. quasi-convex, i.e., the set  $\{(p, w) : v(p, w) \leq \bar{v}\}$  is a convex set for each  $\bar{v}$
- I’ll skip the formal proofs of most parts.
  - Part 1 is the usual – if you scale up prices and wealth,  $B$  is unchanged, so you’re solving the same optimization problem.
  - Part 2, I’ll skip the formal proof, but the intuition is this. If  $(p, w)$  is close to  $(p', w')$ , then the budget sets are close to being the same set; so if  $u$  is continuous, the utility you can achieve on one set is close to the utility you can achieve on the other.
  - Part 3, if  $p$  increases, the budget set shrinks, so you’re maximizing the same function over a smaller set, so the value goes weakly down. If  $w$  goes up, then relative to what used to be your optimal point, you can afford strictly more, and can therefore afford any nearby bundle; by LNS, there’s a nearby bundle that’s strictly better, so you can afford at least one bundle better than your old optimum.

- Skipped in class, but here's the formal proof for part 4, that  $v$  is quasi-convex:
  - We need to show that for any  $\bar{v}$ , the set  $\{(p, w) : v(p, w) \leq \bar{v}\}$  is a convex set
  - Or, if  $(p, w)$  and  $(p', w')$  are both in that set,
    - so is  $t(p, w) + (1 - t)(p', w')$ , which we'll abbreviate  $(p^t, w^t)$
  - so we need to show that if  $v(p, w) \leq \bar{v}$  and  $v(p', w') \leq \bar{v}$ , then  $v(p^t, w^t) \leq \bar{v}$  as well
  
  - To show this, note that for any  $x \in B(p^t, w^t)$ ,
    - $x$  must also be in at least one of  $B(p, w)$  or  $B(p', w')$
  - If it weren't, then we'd know that both  $p \cdot x > w$  and  $p' \cdot x > w'$ , which would mean
 
$$t(p \cdot x) + (1 - t)(p' \cdot x) > tw + (1 - t)w' \quad \longrightarrow \quad p^t \cdot x > w^t$$
    - which would give a contradiction
  - So if  $v(p, w)$  and  $v(p', w')$  are both below  $\bar{v}$ ,
    - then any bundle with  $u(x) > \bar{v}$  must be outside of both  $B(p, w)$  and  $B(p', w')$
    - which we just said means it's outside of  $B(p^t, w^t)$
  - so there's no way to achieve utility more than  $\bar{v}$  in  $B(p^t, w^t)$ , so  $v(p^t, w^t) \leq \bar{v}$ .

## 6 Using the Lagrangian to prove one more cool property of $v$

- We noted last time that a saddle point is the solution to the min-max problem

$$\min_{\lambda, \mu \geq 0} \max_x \mathcal{L}(x, \lambda, \mu)$$

- Further, since we know complementary slackness holds at a saddle point, the value of that problem is

$$\min_{\lambda, \mu} \max_x \mathcal{L}(x, \lambda, \mu) = \mathcal{L}(x^*, \lambda^*, \mu^*) = u(x^*) + \lambda^*(w - p \cdot x^*) + \mu^* \cdot x^* = u(x^*) = v(p, w)$$

since  $x^*$  is the solution to the consume rproblem

- So we can write

$$v(p, w) = \min_{\lambda, \mu \geq 0} \max_x \mathcal{L}(x, \lambda, \mu)$$

- Which gives us an elegant way to prove a cool result:

**Theorem (Roy's Identity).** *Suppose  $v$  is differentiable at  $(p, w) \gg 0$ , and  $\partial v / \partial w > 0$ . Then  $x(p, w)$  is a singleton and*

$$x_i(p, w) = - \frac{\partial v}{\partial p_i}(p, w) \Big/ \frac{\partial v}{\partial w}(p, w)$$

- Originally, we defined  $v(p, w)$  as  $\max u(x)$  subject to  $p \cdot x \leq w$  and  $x \geq 0$ , which is a constrained problem – and in particular, a change in a parameter  $p$  or  $w$  changes the constraints, not the objective function, so it was hard to see how changes in parameters would change the outcome
- But now, the Lagrangian lets us move the constraints into the objective function, which gives us a way to apply the Envelope Theorem to a constrained problem

- Starting with

$$v(p, w) = \min_{\lambda, \mu \geq 0} \max_x \{u(x) + \lambda(w - p \cdot x) + \mu \cdot x\}$$

and define

$$\Phi(\lambda, \mu, p, w) = \max_x \{u(x) + \lambda(w - p \cdot x) + \mu \cdot x\}$$

as the result of the inner maximization problem, so that

$$v(p, w) = \min_{\lambda, \mu \geq 0} \Phi(\lambda, \mu, p, w)$$

- By the envelope theorem,

$$\frac{\partial v}{\partial w} = \left. \frac{\partial \Phi}{\partial w} \right|_{\lambda=\lambda^*, \mu=\mu^*}$$

- And applying the envelope theorem again,

$$\frac{\partial \Phi}{\partial w} = \left. \frac{\partial \mathcal{L}}{\partial w} \right|_{x=x^*}$$

- So now

$$\frac{\partial v}{\partial w} = \left. \frac{\partial}{\partial w} \left( u(x) + \lambda(w - p \cdot x) + \sum_i \mu_i x_i \right) \right|_{\lambda=\lambda^*, \mu=\mu^*, x=x^*} = \lambda|_{\lambda=\lambda^*} = \lambda^*$$

- This is interesting in its own right – if  $(x^*, \lambda^*, \mu^*)$  is a saddlepoint,  $dv/dw = \lambda^*$  – the value of the multiplier on the budget constraint, is the marginal value of wealth!  
(That is, the Lagrange multiplier tells you the marginal benefit of relaxing that constraint!)
- (This makes sense – for any good where  $x_i^* > 0$ , the KT FOC is  $\lambda = \frac{1}{p_i} \frac{\partial u}{\partial x_i}$ , so at a saddlepoint,  $\lambda$  is the marginal utility you'd get from spending an incremental amount of money on good  $i$ )
- If you're already at the optimum, this is equalized across all goods, and so a marginal increase in  $w$  gives you this much additional utility
- But it's nice that the envelope theorem let us show this formally)

- Then we can do the same thing differentiating with respect to price  $p_i$ , and find

$$\frac{\partial v}{\partial p_i} = \frac{\partial}{\partial p_i} \left( u(x) + \lambda(w - p \cdot x) + \sum_i \mu_i x_i \right) \Big|_{\lambda=\lambda^*, \mu=\mu^*, x=x^*} = -\lambda x_i \Big|_{\lambda=\lambda^*, x=x^*} = -\lambda^* x_i^*$$

- (This also makes sense – if  $p_i$  goes up incrementally, then before you adjust your consumption, you're over budget by an amount proportional to  $x_i^*$ , and scaling down consumption to get back under budget costs you that amount times  $\lambda^*$ )

- So

$$-\frac{\partial v / \partial p_i}{\partial v / \partial w} = -\frac{-\lambda^* x_i^*}{\lambda^*} = x_i^*$$

which is Roy's Identity

## 7 The Expenditure Minimization Problem

- Next, we're going to consider a different formulation of the consumer problem, look at its properties, and then look at the link between the two problems
- In producer theory, we considered the general problem of profit maximization, and we also considered the more specialized problem of cost minimization
- Here, we're going to do something very analogous

- We defined the consumer problem

$$\max_{x \geq 0} u(x) \quad \text{subject to} \quad p \cdot x \leq w$$

with  $v(p, w)$  its value and  $x(p, w)$  (Marshallian demand) its solution

- The Expenditure Minimization Problem – or dual consumer problem – is defined as

$$\min_{x \geq 0} p \cdot x \quad \text{subject to} \quad u(x) \geq u$$

- That is, rather than asking how much utility we can buy for wealth  $w$ , we ask how cheaply we can achieve a target utility level  $u$
- We call  $e(p, u)$  the min – the smallest amount of money that can achieve utility  $u$  – this is the *expenditure function*  
And  $h(p, u)$ , the solution to the problem,  
we'll call Hicksian demand, or compensated demand

## 7.1 Why is this problem interesting?

- Our original consumer problem seems like a reasonable proxy for how people might actually make decisions
- The expenditure minimization problem seems more contrived – less related to a problem people really solve
- So why do we like it?
  
- Three reasons
- First, Marshallian Demand faces two effects when a price changes – substitution and wealth effects  
(If  $p_1$  goes down, you might buy less of another good and more of good 1  
But in addition, if  $p_1$  goes down, this makes you effectively richer – which can change your demand in any direction)  
Hicksian demand eliminates the wealth effect –  
so it gives a way to kind of decompose price effects into their separate components
  
- Second, because Hicksian demand doesn't have wealth effects,  
we can get stronger, testable predictions  
And importantly, we can infer Hicksian demand from Marshallian demand –  
so if we observe data from the regular consumer problem,  
this gives us a way to test the predictions of the model
  
- Third, we can't really do welfare analysis using indirect utility functions,  
because utility has no natural scale and people aren't comparable  
But the Expenditure Function is in dollars –  
we can measure welfare changes by how much money it would take to compensate people,  
and then aggregate across people in a more valid way
  
- (It also creates a nice parallel with what we did in producer theory)