

# Mathematics for Economists: Lecture 3

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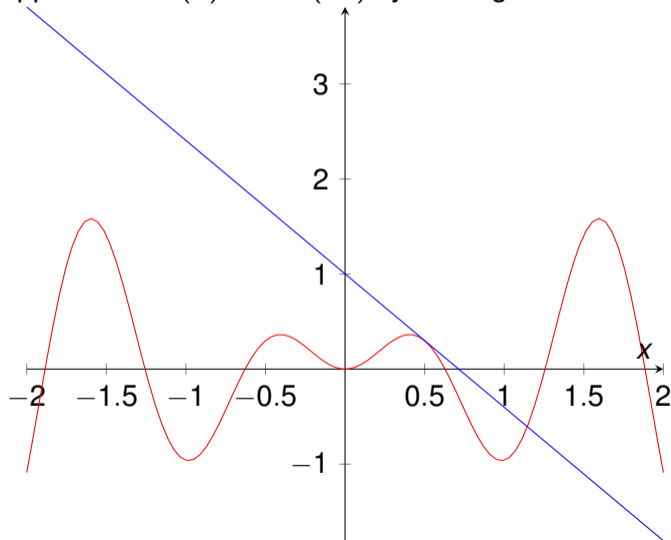
Spring 2022

## Content of Lecture 3

- ▶ In Lecture 2, Gaussian elimination and linear models in economics
- ▶ This Lecture:
  1. Linear approximation of functions of a real variable: the derivative
  2. Visualizing multivariate functions
  3. Linear approximations to multivariate functions
  4. Linear approximations and partial derivatives
  5. Directions of increase and level curves
  6. Non-linear models in economics: first examples

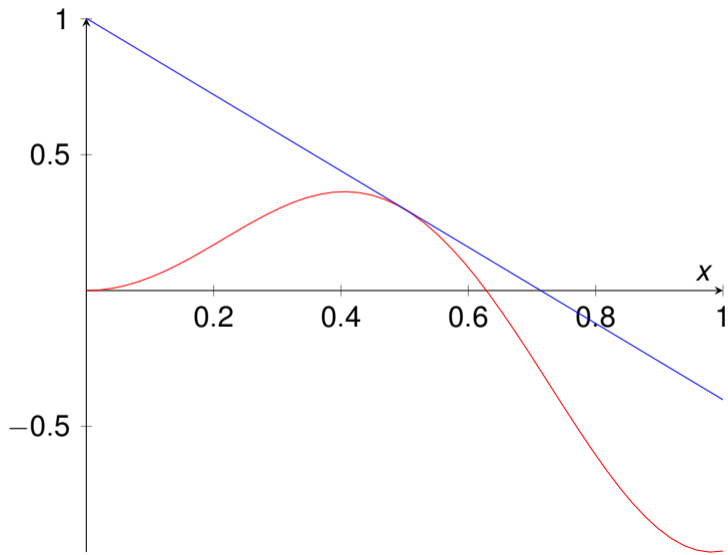
## Linear approximation of univariate functions

Approximate  $f(x) = x\sin(5x)$  by its tangent at  $x = 0.5$ . Not a great success:



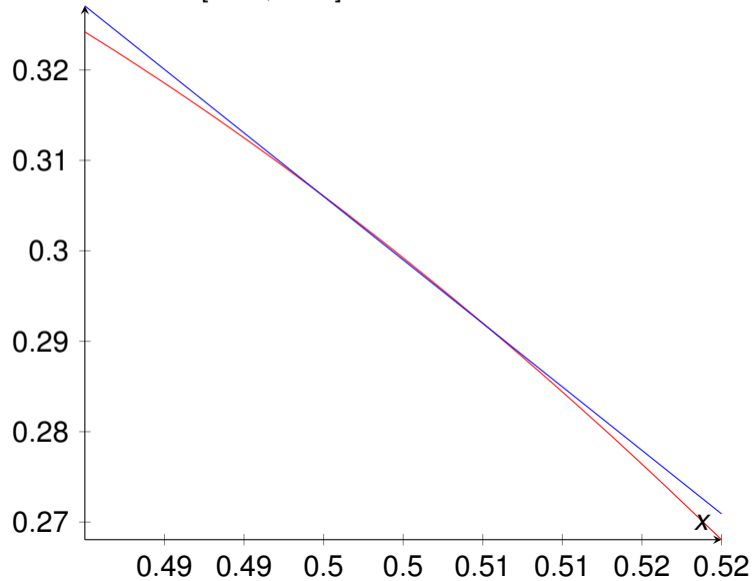
## Linear approximation of univariate functions

The function is a bit less variable over the interval  $[0, 1]$ :



# Linear approximation of univariate functions

On the interval  $[0.48, 0.52]$ , it looks almost linear:



## Linear approximation of univariate functions

- ▶ Recall the definition of the derivative of a function  $f : \mathbb{R} \rightarrow \mathbb{R}$  at  $x_0$ :

$$Df(x_0) = \frac{df(x_0)}{dx} = f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h}.$$

- ▶ If the limit exists, we also have (from the definition of limits) that for all  $\varepsilon > 0$ , there is a  $\delta > 0$  such that:

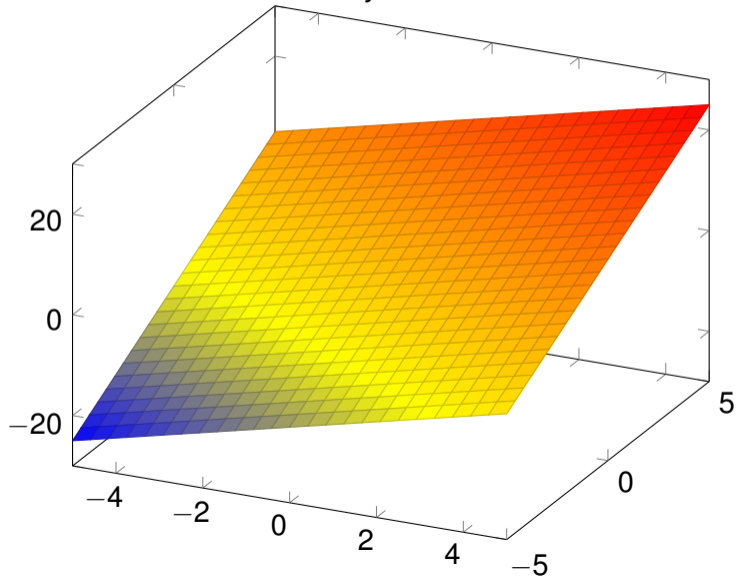
$$\frac{|f(x_0 + h) - f(x_0) - Df(x_0)h|}{|h|} < \varepsilon,$$

whenever  $|h| < \delta$ .

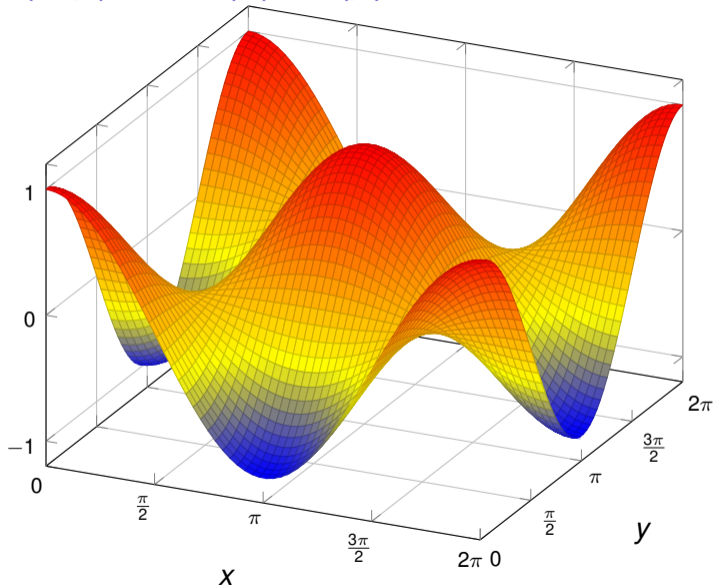
- ▶ We say then that  $Df(x_0)h$  approximates  $f(x) - f(x_0)$  well near  $x_0$ .
- ▶  $Df(x_0)h$  is a linear (in  $h$ ) approximation of the changes in the value of the function near  $x_0$ .
- ▶ We want to generalize this idea to multivariate functions.

# Graphing functions of two variables

A linear function:  $z = 2x + 3y$ .



The graph of  $f(x, y) = \cos(x)\cos(y)$





## 2-d slices of the 3-d graph

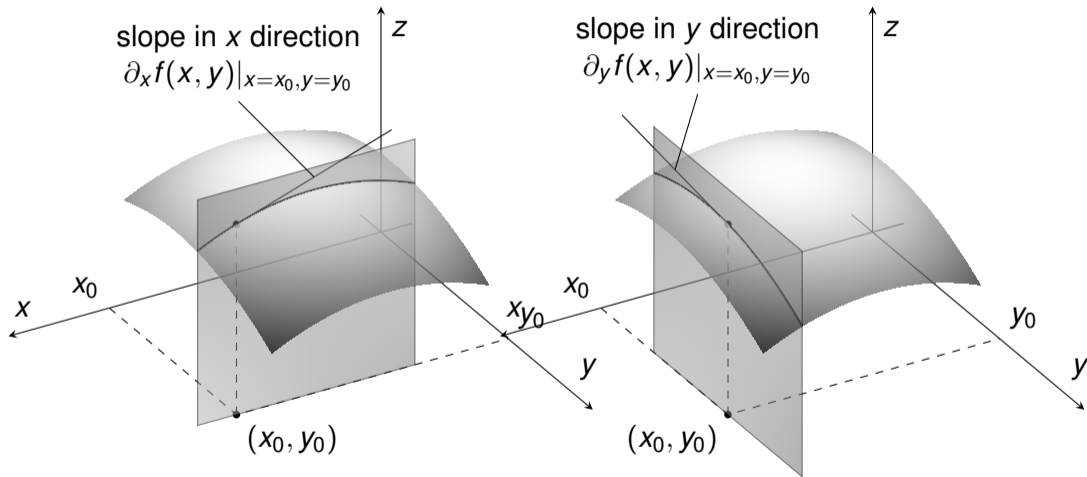


Figure: Cross sections of  $f(x, y)$  in the  $x$  and  $y$  direction at  $(x_0, y_0)$ .

# Level curves of a bivariate function

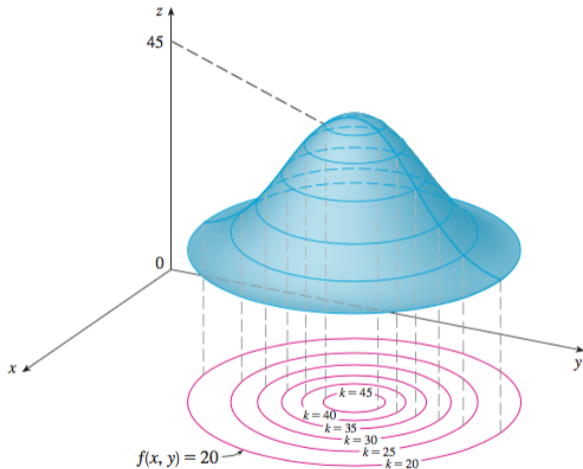


Figure: Some level curves of  $f$ .

All in one picture

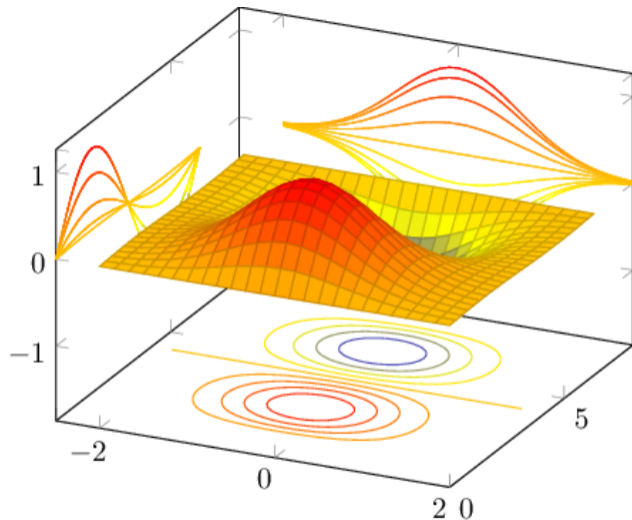


Figure: The graph of  $f$  together with some of its cross sections and level curves.

## Linear functions

- ▶ Recall that a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is said to be linear if for all  $\lambda \in \mathbb{R}$  and for all  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ , i)  $f(\lambda\mathbf{x}) = \lambda f(\mathbf{x})$ , ii)  $f(\mathbf{x} + \mathbf{y}) = f(\mathbf{x}) + f(\mathbf{y})$ .
- ▶ For  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , let  $f(\mathbf{e}^i) = a_i$ , where  $\mathbf{e}^i = (e_1^i, \dots, e_n^i)$  is the  $i^{\text{th}}$  unit vector, i.e.  $e_j^i = 0$  if  $i \neq j$  and  $e_i^i = 1$ .
- ▶ Then we have

$$f(\mathbf{x}) = \sum_{i=1}^n f(x_i \mathbf{e}^i) = \sum_{i=1}^n x_i f(\mathbf{e}^i) = \sum_{i=1}^n a_i x_i = \mathbf{a} \cdot \mathbf{x},$$

where  $\mathbf{a} = (a_1, \dots, a_n)$ .

- ▶ For  $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , let  $f(\mathbf{e}^i) = \mathbf{a}^i \in \mathbb{R}^m$ . By the same reasoning as above, any linear  $\mathbf{f}$  takes the form:

$$\mathbf{f}(\mathbf{x}) = \mathbf{A}\mathbf{x},$$

where the  $ij^{\text{th}}$  element of  $\mathbf{A}$  is the  $i^{\text{th}}$  row element of  $\mathbf{a}^j$ .

# Linear approximation of multivariate functions

- ▶ If we want to find a linear approximation to  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  near  $\mathbf{x}_0$ , we look for a vector  $\mathbf{a}$  such that  $\frac{|f(\mathbf{x}_0+h) - f(\mathbf{x}_0) - \mathbf{a} \cdot \mathbf{x}|}{\|\mathbf{x} - \mathbf{x}_0\|}$  is small whenever  $\|\mathbf{x} - \mathbf{x}_0\|$  is small.
- ▶ If such an  $\mathbf{a}$  exists, we call it the derivative  $D_{\mathbf{x}}f(\mathbf{x}_0)$  of  $f$  at  $\mathbf{x}_0$  and we say that  $f$  is differentiable at  $\mathbf{x}_0$ .
- ▶ When the derivative exists, we have for small  $\|\mathbf{x} - \mathbf{x}_0\|$ :

$$f(\mathbf{x}) - f(\mathbf{x}_0) \approx D_{\mathbf{x}}f(\mathbf{x}_0)(\mathbf{x} - \mathbf{x}_0).$$

- ▶ Our next task is to identify the components of  $D_{\mathbf{x}}f(\mathbf{x}_0)$ .

# Partial derivatives

- ▶ Consider changes in the direction of a coordinate axis:  $\mathbf{x} = \mathbf{x}_0 + h\mathbf{e}^i$ .
- ▶ Since all the other coordinates of  $\mathbf{x}$  remain fixed, we can compute:

$$\lim_{h \rightarrow 0} \frac{f(\mathbf{x}_0 + h\mathbf{e}^i) - f(\mathbf{x}_0)}{h},$$

exactly as in the case of univariate functions,

- ▶ We call this limit the  $i^{\text{th}}$  partial derivative of  $f$  at  $\mathbf{x}_0$  and denote it by:

$$D_{x_i} f(\mathbf{x}_0) := \frac{\partial f(\mathbf{x}_0)}{\partial x_i} := \lim_{h \rightarrow 0} \frac{f(\mathbf{x}_0 + h\mathbf{e}^i) - f(\mathbf{x}_0)}{h}.$$

# Partial derivatives

Recall the picture from before:

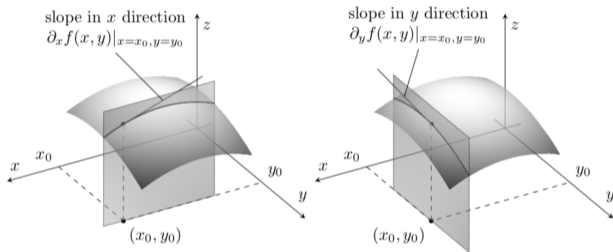


Figure: Partial derivatives of  $f$  at  $(x_0, y_0)$ .

## From partial derivatives to linear approximation

- ▶ But this is all we need if a linear approximation exists!
- ▶ A linear approximation in the direction  $\Delta \mathbf{x} = \mathbf{e}^i$  must coincide with  $\frac{\partial f(\mathbf{x}_0)}{\partial x_i}$ .
- ▶ But each direction  $\Delta \mathbf{x}$  can be written as:

$$\Delta \mathbf{x} = \sum_{i=1}^n x_i \mathbf{e}^i,$$

so by linearity we get for all  $\Delta \mathbf{x}$  that

$$f(\mathbf{x}_0 + \Delta \mathbf{x}) - f(\mathbf{x}_0) = D_{\mathbf{x}} f(\mathbf{x}_0) \cdot \Delta \mathbf{x} + \text{h.o.t.},$$

where  $D_{\mathbf{x}} f(\mathbf{x}_0)$  is the row vector of partial derivatives

$$D_{\mathbf{x}} f(\mathbf{x}) = \left( \frac{\partial f(\mathbf{x}_0)}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x}_0)}{\partial x_n} \right).$$

- ▶ A linear approximation exists and  $f$  is differentiable at  $\mathbf{x}_0$  if all of its partial derivatives exist at  $\mathbf{x}_0$  and are continuous in  $\mathbf{x}$ .



## Planar approximation to a non-linear surface

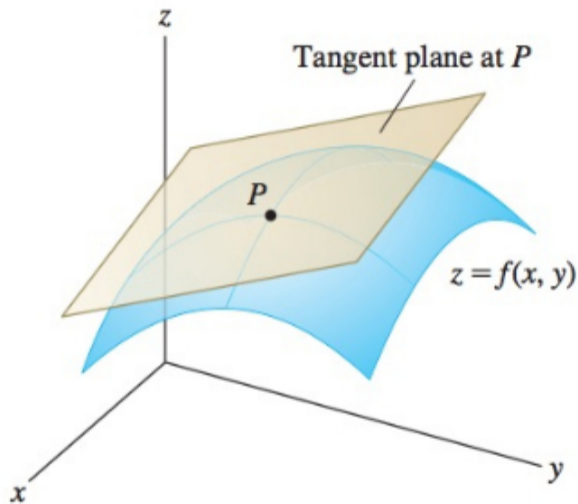


Figure: Linear approximation to  $f$  at point  $P$ .

## Computing the derivative: an example

- ▶ Compute at  $(x_1, x_2, x_3) = (1, 2, 1)$  the derivative of the following function:

$$f(x_1, x_2, x_3) = x_1 \ln x_2 + \sqrt{x_2 x_3}.$$

- ▶ Since we have a real-valued function  $f$ , its derivative is the row vector of its partial derivatives evaluated  $x = (x_1, x_2, x_3)$ :

$$\begin{aligned} D_{\mathbf{x}}f(\mathbf{x}) &= \left( \frac{\partial f(x_1, x_2, x_3)}{\partial x_1}, \frac{\partial f(x_1, x_2, x_3)}{\partial x_2}, \frac{\partial f(x_1, x_2, x_3)}{\partial x_3} \right) \\ &= \left( \ln x_2, \frac{x_1}{x_2} + \frac{1}{2}x_2^{-\frac{1}{2}}x_3^{\frac{1}{2}}, \frac{1}{2}x_2^{\frac{1}{2}}x_3^{-\frac{1}{2}} \right). \end{aligned}$$

Evaluating at  $(1, 2, 1)$

$$D_{\mathbf{x}}f(1, 2, 1) = \left( \ln 2, \frac{1}{2} + \frac{1}{2\sqrt{2}}, \frac{\sqrt{2}}{2} \right).$$

## Utility functions: marginal utilities

- ▶ Utility function  $u : \mathbb{R}_+^n \rightarrow \mathbb{R}$  assigns a numerical value  $u(\mathbf{x})$  for each possible (positive) consumption vector  $\mathbf{x} \in \mathbb{R}_+^n := \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{x} \geq 0\}$ .
- ▶  $u(\mathbf{x}) \geq u(\mathbf{y})$  if and only if the consumer considers  $\mathbf{x}$  at least as good as  $\mathbf{y}$
- ▶ Consider first the case with two goods, i.e.  $n = 2$ .
- ▶ Partial derivatives of the utility function are called the *marginal utilities* denoted by  $MU_{x_i}$ .

$$MU_{x_i}(\hat{x}_1, \hat{x}_2) := \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_i}.$$

- ▶ If  $\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1} > 0$ , then  $u(\hat{x}_1 + h, \hat{x}_2) > u(\hat{x}_1, \hat{x}_2)$  for small  $h > 0$ , and we say that utility is strictly increasing in good 1 at  $(\hat{x}_1, \hat{x}_2)$ .
- ▶ If this holds at all  $(x_1, x_2)$ , we say simply that utility is strictly increasing. Typically it is assumed that utility is strictly increasing in all goods.

## Utility functions: marginal utilities

- ▶ For small consumption changes  $(\Delta x_1, \Delta x_2)$ , we can approximate the change in utility by using the derivative  $D_{\mathbf{x}}u(\hat{x}_1, \hat{x}_2)$ :

$$\begin{aligned}u(\hat{x}_1 + \Delta x_1, \hat{x}_2 + \Delta x_2) - u(\hat{x}_1, \hat{x}_2) &= D_{\mathbf{x}}u(\hat{x}_1, \hat{x}_2)(\Delta x_1, \Delta x_2) \\ &= \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1} \Delta x_1 + \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2} \Delta x_2.\end{aligned}$$

- ▶ Recall from Principles 1 that  $(\hat{x}_1 + \Delta x_1, \hat{x}_2 + \Delta x_2)$  and  $(\hat{x}_1, \hat{x}_2)$  are on the same indifference curve if they are equally good to the consumer:  
 $u(\hat{x}_1 + \Delta x_1, \hat{x}_2 + \Delta x_2) = u(\hat{x}_1, \hat{x}_2)$ .

- ▶ But then we have:

$$\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1} \Delta x_1 + \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2} \Delta x_2 = 0,$$

or

$$\Delta x_2 = -\frac{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}}{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2}} \Delta x_1.$$

## Utility functions: MRS

- ▶ The consumer is willing to give up  $\frac{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}}{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2}}$  units of good 2 to get an additional unit of good 1 at  $(\hat{x}_1, \hat{x}_2)$ .
- ▶ Hence marginal rate of substitution at  $(\hat{x}_1, \hat{x}_2)$  is captured in the ratio of marginal utilities:

$$MRS_{x_1, x_2}(\hat{x}_1, \hat{x}_2) = \frac{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}}{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2}} = \frac{MU_{x_1}(\hat{x}_1, \hat{x}_2)}{MU_{x_2}(\hat{x}_1, \hat{x}_2)}.$$

## Utility functions: MRS

- ▶ If  $n > 2$  we can ask how many (small) units of good  $j$  the consumer would be willing to give up in order to get an additional (small) unit of good  $i$ .
- ▶ If all the other goods remain fixed at  $\hat{\mathbf{x}}$  and

$$u(\hat{\mathbf{x}} + \Delta x_i \mathbf{e}^i + \Delta x_j \mathbf{e}^j) = u(\hat{\mathbf{x}}),$$

then

$$\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_i} \Delta x_i + \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_j} \Delta x_j = 0,$$

and we have:

$$MRS_{x_i, x_j}(\hat{\mathbf{x}}) = \frac{\frac{\partial u(\hat{\mathbf{x}})}{\partial x_i}}{\frac{\partial u(\hat{\mathbf{x}})}{\partial x_j}} = \frac{MU_{x_i}(\hat{\mathbf{x}})}{MU_{x_j}(\hat{\mathbf{x}})}.$$

# The gradient

- ▶ The gradient of the utility function denoted by  $\nabla u(\mathbf{x})$  is the transpose of its derivative:

$$\nabla u(\mathbf{x}) = \left( \frac{\partial u(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial u(\mathbf{x})}{\partial x_n} \right).$$

- ▶ Does the gradient have any particular interpretation?
- ▶ A first observation is that when  $n = 2$ , the gradient is orthogonal to the indifference curve:

$$\left( 1, -\frac{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}}{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2}} \right) \cdot \left( \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}, \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2} \right) = \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1} - \frac{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_1}}{\frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2}} \frac{\partial u(\hat{x}_1, \hat{x}_2)}{\partial x_2} = 0.$$

# The gradient

- ▶ The gradient at  $\mathbf{x}$  gives the direction in which the utility function increases the fastest near  $\mathbf{x}$ .
- ▶ To see this, consider the change in the utility using the linear approximation:

$$u(\hat{\mathbf{x}} + \Delta\mathbf{x}) - u(\hat{\mathbf{x}}) = D_{\mathbf{x}}(\hat{\mathbf{x}})\Delta\mathbf{x} + h.o.t.$$

- ▶ For a unit length (or norm) of  $\Delta\mathbf{x}$ , the change in utility is maximized at

$$\Delta\mathbf{x} = \frac{1}{\|\nabla u(\hat{\mathbf{x}})\|} \nabla u(\hat{\mathbf{x}})$$

by Cauchy's inequality.



# Computing marginal utilities and the MRS

- ▶ Linear utility:

$$u(\mathbf{x}) = \mathbf{a} \cdot \mathbf{x} = \sum_{i=1}^n a_i x_i.$$

Then  $MU_{x_i}(\mathbf{x}) = a_i$  for all  $i$  and all  $\mathbf{x}$ , and  $MRS_{x_i, x_j} = \frac{a_i}{a_j}$  for all  $i \neq j$  and all  $\mathbf{x}$ .

- ▶ Quasilinear utility:

$$u(x_1, x_2) = v(x_1) + x_2,$$

for some increasing function  $v$ .

$$MU_{x_2} = 1, MU_{x_1} = MRS_{x_1, x_2} = v'(x_1).$$

For example if  $v(x_1) = \ln(x_1)$ , then

$$MU_{x_1} = MRS_{x_1, x_2} = \frac{1}{x_1}.$$

# Computing marginal utilities and the MRS

- ▶ Cobb-Douglas utility:

$$u(x_1, x_2) = x_1^\alpha x_2^{1-\alpha} \text{ for } \alpha \in (0, 1).$$

In this case,

$$MU_{x_1}(x_1, x_2) = \alpha \left(\frac{x_2}{x_1}\right)^{1-\alpha}, \quad MU_{x_2}(x_1, x_2) = (1 - \alpha) \left(\frac{x_1}{x_2}\right)^\alpha,$$

and therefore:

$$MRS_{x_1, x_2} = \frac{\alpha \left(\frac{x_2}{x_1}\right)^{1-\alpha}}{(1 - \alpha) \left(\frac{x_1}{x_2}\right)^\alpha} = \frac{\alpha x_2}{(1 - \alpha) x_1}.$$

# Computing marginal utilities and the MRS

- ▶ For  $n > 2$ , we have:

$$u(\mathbf{x}) = \prod_{i=1}^n x_i^{\alpha_i} \text{ for } \alpha_i \geq 0, \sum_{i=1}^n \alpha_i = 1.$$

Denote  $y = \prod_{i=1}^n x_i^{\alpha_i}$ . Then

$$MU_{x_i} = \alpha_i \frac{y}{x_i},$$

and therefore

$$MRS_{x_i, x_j}(\mathbf{x}) = \frac{\alpha_i x_j}{\alpha_j x_i}.$$

Exercise: Compute the marginal utilities for  $u(\mathbf{x}) = \sum_{i=1}^n \alpha_i \ln x_i$ .

## Computing marginal utilities and the MRS

- ▶ Constant elasticity of substitution utility (CES utility function): for  $\rho < 1$  and  $\rho \neq 0$ ,

$$u(x_1, x_2) = (a_1 x_1^\rho + a_2 x_2^\rho)^{\frac{1}{\rho}}.$$

Denote  $y(x_1, x_2) = (a_1 x_1^\rho + a_2 x_2^\rho)$  and we have  $u(x_1, x_2) = y(x_1, x_2)^{\frac{1}{\rho}}$ . We get by chain rule that

$$MU_{x_i}(x_1, x_2) = \frac{1}{\rho} y(x_1, x_2)^{\frac{1-\rho}{\rho}} \rho a_i x_i^{\rho-1}.$$

Therefore the marginal rates of substitution are quite simple:

$$MRS_{x_1, x_2}(x_1, x_2) = \frac{a_1}{a_2} \left( \frac{x_1}{x_2} \right)^{\rho-1}.$$

## Next Lecture

- ▶ Vector valued multivariate functions
- ▶ Implicit function theorem
- ▶ Comparative statics in economic models