

Discussion Session Week 5 (2 Nov 20)

$$1) f(x, y) = \begin{cases} 1 & 0 \leq x \leq 1 \text{ and } 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$z = xy$$

$$x = u$$

 \Rightarrow

$$u = x$$

$$y = \frac{z}{u}$$

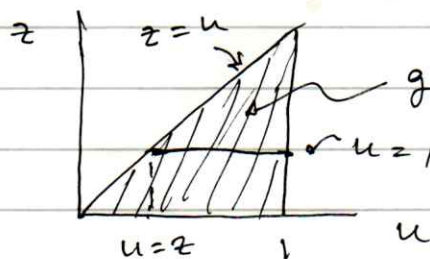
$$J = \begin{vmatrix} \frac{\partial x}{\partial z} & \frac{\partial x}{\partial u} \\ \frac{\partial y}{\partial z} & \frac{\partial y}{\partial u} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ \frac{1}{u} & -\frac{z}{u^2} \end{vmatrix} = -\frac{1}{u}$$

$$g(z, u) = |J| f(x(z, u), y(z, u))$$

$$= \begin{cases} \frac{1}{u} & 0 \leq u \leq 1 \text{ and } 0 \leq z \leq u \\ 0 & \text{otherwise} \end{cases}$$

$$0 \leq x \leq 1 \Rightarrow 0 \leq u \leq 1$$

$$0 \leq y \leq 1 \Rightarrow 0 \leq xy \leq x \Rightarrow$$



$g(z, u) = \frac{1}{u}$ here, 0 otherwise

$$g_z(z) = \int g(z, u) du = \int_z^1 \frac{du}{u} = -\ln z$$

$$y \leq 1 \Rightarrow xy \leq x \Rightarrow z \leq u$$

$$1b) \quad g(z) = \int f(x, \frac{z}{x}) \frac{dx}{x}$$

$$= \int_{x_{\min}}^{x_{\max}} 1 \cdot \frac{dx}{x}$$

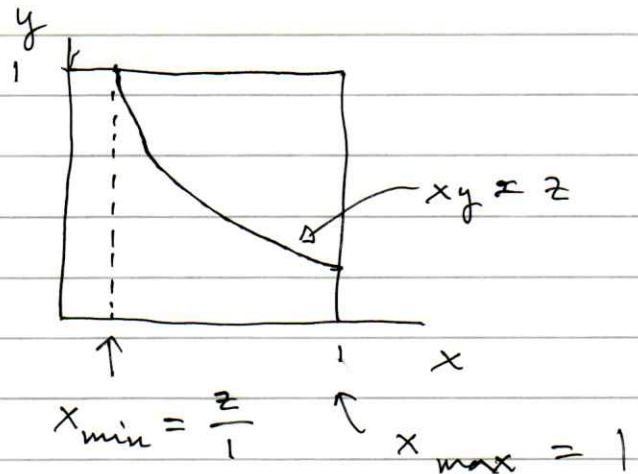
$f(x, y)$ is nonzero for

$$0 \leq x \leq 1$$

$$0 \leq y \leq 1 \quad \Rightarrow \quad 0 \leq \frac{z}{x} \leq 1$$

$$\Rightarrow \quad 0 \leq z \leq x$$

i.e. for a given
value z



$$\Rightarrow g_z(z) = \int_z^1 \frac{dx}{x} = \ln x \Big|_z^1$$

$$= -\ln z$$

$$2) \quad E[\alpha x + \beta] = \int (\alpha x + \beta) f(x) dx$$

$$= \alpha \int x f(x) dx + \beta$$

$$= \alpha E[x] + \beta$$

$$V[\alpha x + \beta] = E[(\alpha x + \beta)^2] - (E[\alpha x + \beta])^2$$

$$= E[\alpha^2 x^2 + 2\alpha\beta x + \beta^2] - (\alpha^2 \mu^2 + 2\alpha\beta\mu + \beta^2)$$

\uparrow
 $\mu \equiv E[x]$

$$= \alpha^2 E[x^2] + \cancel{2\alpha\beta\mu} + \cancel{\beta^2} - \alpha^2 \mu^2 - \cancel{2\alpha\beta\mu} - \cancel{\beta^2}$$

$$= \alpha^2 (E[x^2] - \mu^2)$$

$$= \alpha^2 V[x]$$

$$3a) \quad V[\alpha x + y] = E[(\alpha x + y)^2] - (E[\alpha x + y])^2$$

$$= E[\alpha^2 x^2 + 2\alpha xy + y^2] - (\alpha \mu_x + \mu_y)^2$$

$$= \alpha^2 E[x^2] + 2\alpha E[xy] + E[y^2]$$

$$- \alpha^2 \mu_x^2 - 2\alpha \mu_x \mu_y - \mu_y^2$$

$$= \alpha^2 (E[x^2] - \mu_x^2) + E[y^2] - \mu_y^2$$

$$+ 2\alpha (E[xy] - \mu_x \mu_y)$$

$$= \alpha^2 V[x] + V[y] + 2\alpha \underbrace{\text{cov}[x, y]}_{= \rho \sigma_x \sigma_y}$$

$$b) \quad V[\alpha x + y] = \alpha^2 \sigma_x^2 + \sigma_y^2 + 2\alpha \rho \sigma_x \sigma_y \geq 0$$

$$\text{Let } \alpha = \sigma_y / \sigma_x$$

$$\frac{\sigma_y^2}{\sigma_x^2} \sigma_x^2 + \sigma_y^2 + 2 \left(\frac{\sigma_y}{\sigma_x} \right) \rho \sigma_x \sigma_y \geq 0$$

$$\Rightarrow 2\sigma_y^2 + 2\rho \sigma_y^2 \geq 0 \Rightarrow \rho \geq \underline{\underline{-1}}$$

$$\text{Let } \alpha = -\sigma_y / \sigma_x$$

$$\left(\frac{-\sigma_y}{\sigma_x} \right)^2 \sigma_x^2 + \sigma_y^2 - 2 \frac{\sigma_y}{\sigma_x} \rho \sigma_x \sigma_y \geq 0$$

$$\sigma_y^2 + \sigma_y^2 - 2\rho \sigma_y^2 \geq 0 \Rightarrow \rho \leq \underline{\underline{+1}}$$

$$4) \quad x_1 : \mu_1 = 10, \sigma_1^2 = 1$$

$$x_2 : \mu_2 = 10, \sigma_2^2 = 1$$

$$\text{cov}[x_1, x_2] = 0$$

$$\text{Let } y = \frac{x_1^2}{x_2}$$

$$V[y] = \left(\frac{\partial y}{\partial x_1} \right)^2 \Big|_{\mu} \sigma_1^2 + \left(\frac{\partial y}{\partial x_2} \right)^2 \Big|_{\mu} \sigma_2^2$$

$$= \left(\frac{2x_1}{x_2} \right)^2 \Big|_{\mu} \sigma_1^2 + \left(-\frac{x_1^2}{x_2^2} \right)^2 \Big|_{\mu} \sigma_2^2$$

$$= 4 \cdot 1 + 1 \cdot 1 = 5$$

$$\Rightarrow \sigma_y = \sqrt{5} = 2.2$$

If $\mu_2 = 1$, then $y = x_1^2/x_2$ is

significantly non linear in a region

$x_2 \in \mu_2 \pm \sigma_2$ & therefore linear error

prop. is a poor approx.

