

JTMS-12: Probability and Random Processes

Fall 2020

M. Bode

Lecture 12

Recap

Chapter 3: Functions of Random Variables

3.1 Functions of Random Variables

3.2 $Y = g(X)$

3.3 $Z = g(X, Y)$

3.4 $V = g(X, Y), W = h(X, Y)$

Functions of Random Variables

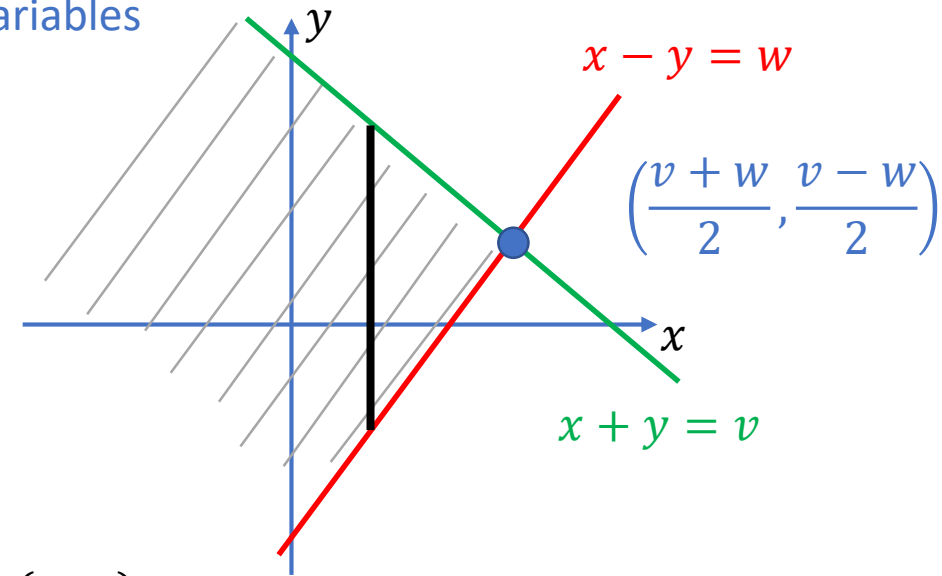
Example:

Consider continuous
r.v.s X, Y , and

$$V = g(X, Y) = X + Y$$

$$W = h(X, Y) = X - Y$$

Find $F_{VW}(v, w)$ and $f_{VW}(v, w)$.



Standard approach ...

$$F_{VW}(v, w) = P[V \leq v, W \leq w] = P[g(X, Y) \leq v, h(X, Y) \leq w]$$

$$= P[X + Y \leq v, X - Y \leq w] = \iint_{\text{shaded area}} f_{XY}(x, y) dx dy$$

$$= \int_{-\infty}^{\frac{v+w}{2}} \left(\int_{x-w}^{v-x} f_{XY}(x, y) dy \right) dx$$

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Functions of Random Variables

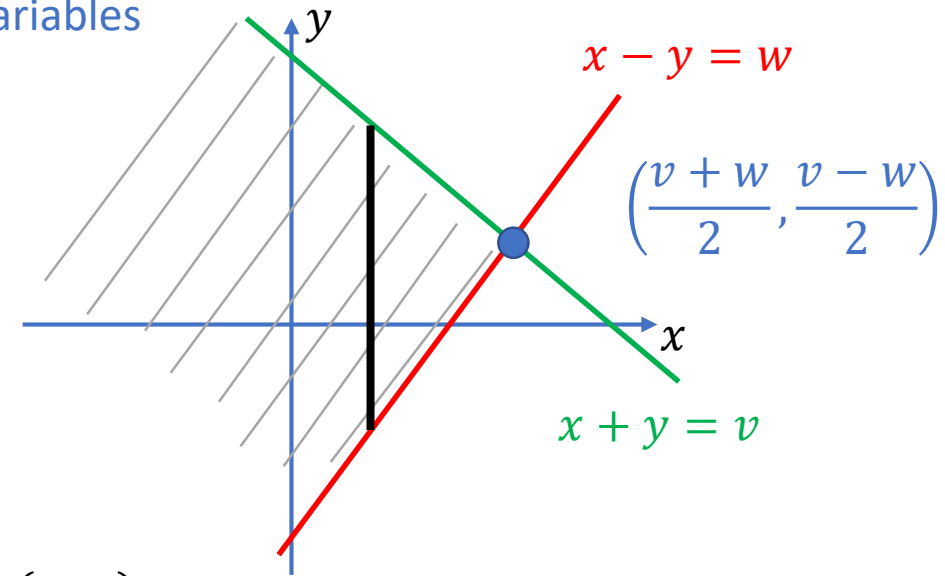
solve:

Consider continuous
r.v.s X, Y , and

$$V = g(X, Y) = X + Y$$

$$W = h(X, Y) = X - Y$$

Find $F_{VW}(v, w)$ and $f_{VW}(v, w)$.



Standard approach ...

$$F_{VW}(v, w) = \int_{-\infty}^{\frac{v+w}{2}} \left(\int_{x-w}^{v-x} f_{XY}(x, y) dy \right) dx$$

$$\Rightarrow f_{VW}(v, w) = \frac{\partial^2}{\partial v \partial w} F_{VW}(v, w) = \frac{1}{2} f_{XY} \left(\frac{v+w}{2}, \frac{v-w}{2} \right)$$

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Recap

Chapter 3: Functions of Random Variables

3.1 Functions of Random Variables

3.2 $Y = g(X)$

3.3 $Z = g(X, Y)$

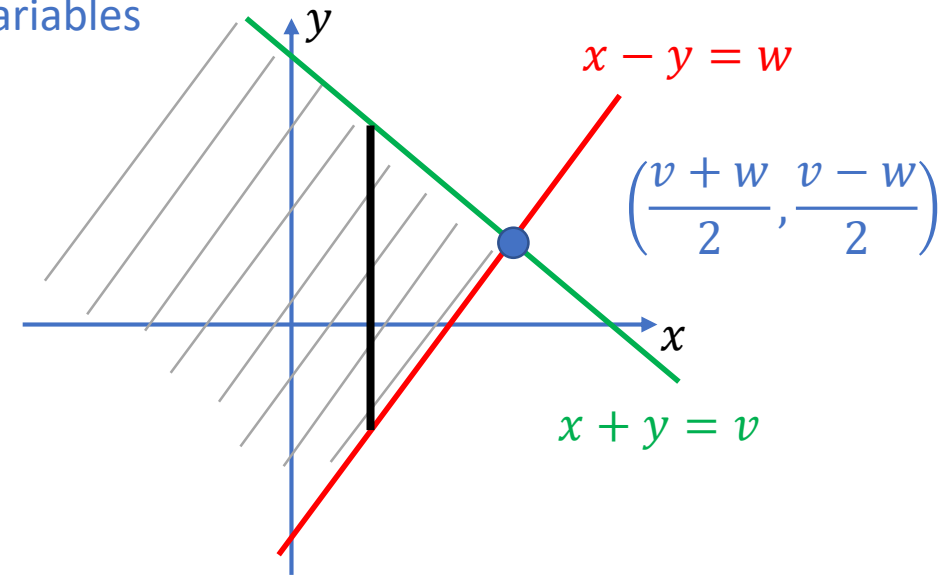
3.4 $V = g(X, Y), W = h(X, Y)$

Functions of Random Variables

solve:

Consider continuous
r.v.s X, Y , and

$$V = g(X, Y) = X + Y$$
$$W = h(X, Y) = X - Y$$



Details ...

$$f_{VW}(v, w) = \frac{\partial^2}{\partial v \partial w} \int_{-\infty}^{\frac{v+w}{2}} \left(\int_{x-w}^{v-x} f_{XY}(x, y) dy \right) dx$$

$$= \frac{\partial}{\partial v} \left[\frac{1}{2} \int_{\frac{v-w}{2}}^{\frac{v-w}{2}} f_{XY} \left(\frac{v+w}{2}, y \right) dy + \int_{-\infty}^{\frac{v+w}{2}} f_{XY}(x, x-w) dx \right] = \frac{1}{2} f_{XY} \left(\frac{v+w}{2}, \frac{v-w}{2} \right)$$

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Chapter 3: Functions of Random Variables

3.1 Functions of Random Variables

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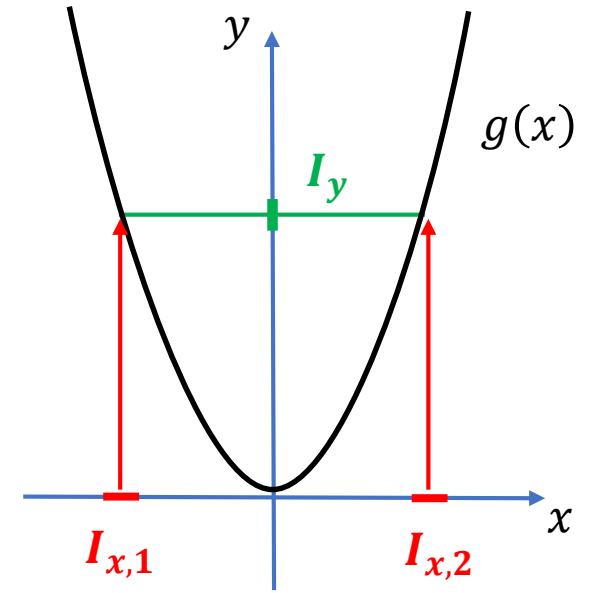
Functions of Random Variables

Determine densities like

$f_Y(y)$ directly

Recall:

$$f_Y(y)|\Delta y| \approx f_X(x_1)|\Delta x_1| + f_X(x_2)|\Delta x_2|$$



Basically, we try to calculate $\int f_X(x)dx$

via

$$\int f_X(x(y)) \frac{\Delta(x)}{\Delta(y)} dy = \int f_Y(y) dy$$

based on the ratio $\frac{\Delta(x)}{\Delta(y)}$ of corresponding “line/area-elements”.
Possibly with a need to sum over several pre-images ...

$$\left| \frac{\Delta x}{\Delta y} \right| = \left| \frac{1}{g'(x)} \right| = |\varphi'(x)|, \quad \text{if } \varphi = g^{-1}, x = \varphi(y)$$

We'll stick to this inverse perspective ...

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Functions of Random Variables

Determine densities like $f_{UV}(u, v)$, etc. directly

Chapter 3: Functions of Random Variables

3.1 Functions of Random Variables

3.2 $Y = g(X)$

3.3 $Z = g(X, Y)$

3.4 $V = g(X, Y), W = h(X, Y)$

We try to calculate:

$$\iint f_{XY}(x, y) dx dy$$

via

$$\iint f_{XY}(x(u, v), y(u, v)) \frac{\Delta(x, y)}{\Delta(u, v)} du dv = \iint f_{UV}(u, v) du dv$$

based on the ratio $\frac{\Delta(x, y)}{\Delta(u, v)}$ of corresponding “area-elements”.

So, how large is that ratio?

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Functions of Random Variables

Determine $f_{UV}(u, v)$... direct method

Special case:

Use $x = u$ and $y = u + v$

Start with a rectangle in the (u, v) – plane.

Consider the four points at positions:

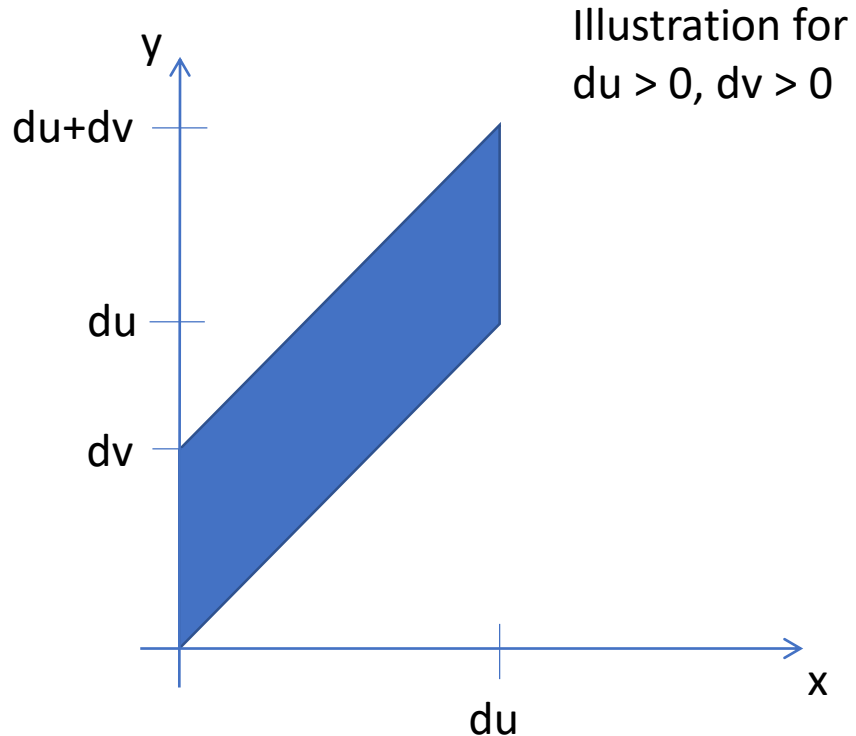
$$(u_i, v_i) = (0, 0), (du, 0), (du, dv), (0, dv)$$

The enclosed area is $du \cdot dv$... or its absolute value.

Transform to the (x, y) - plane:

$$(x_i, y_i) = (0, 0), (du, du), (du, du + dv), (0, dv)$$

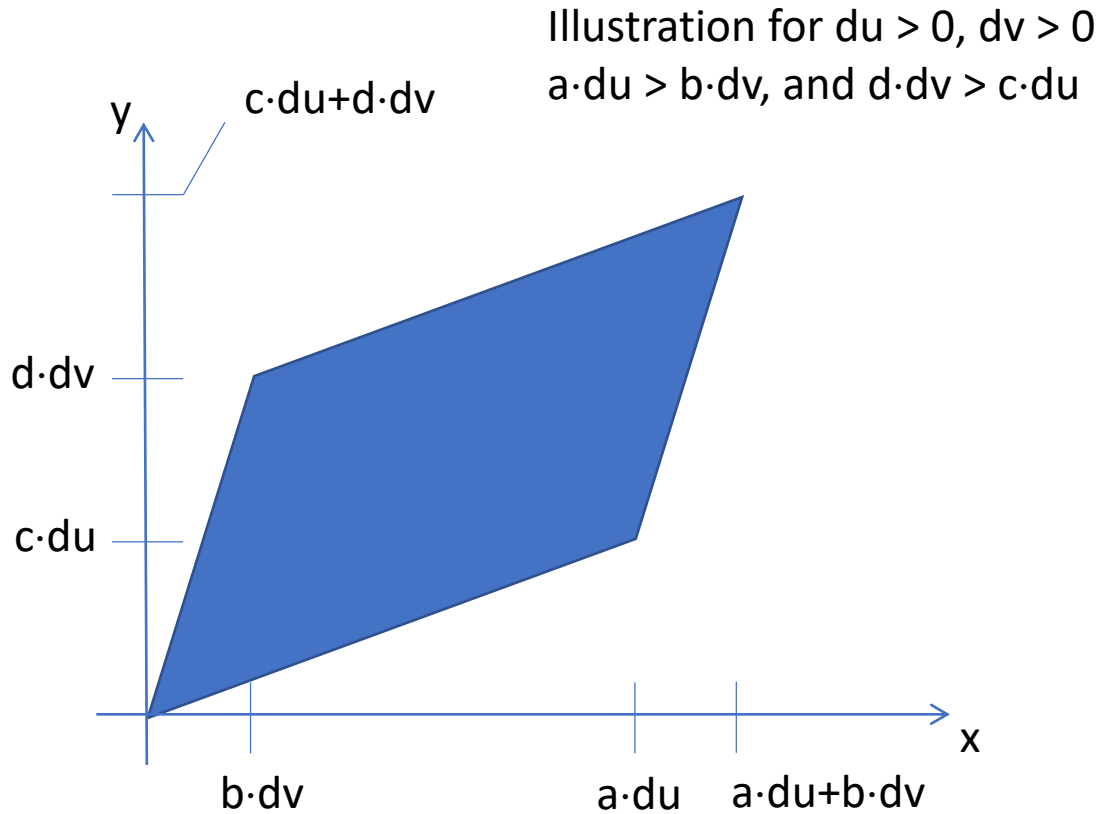
with the same area, and the ratio is 1.



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Functions of Random Variables

Determine $f_{UV}(u, v)$... direct method



General case: Use $x = au + bv$ and $y = cu + dv$

Again, start with a rectangle in the (u, v) – plane.

Consider the same four points as before:

$$(u_i, v_i) = (0, 0), (du, 0), (du, dv), (0, dv)$$

The enclosed area is $du \cdot dv$.

Now, transform to the (x, y) - plane:

$$(x_i, y_i) = (0, 0), (a \cdot du, c \cdot du),$$

$$(a \cdot du + b \cdot dv, c \cdot du + d \cdot dv), (b \cdot dv, d \cdot dv)$$

Assume that a, b, c, d are all positive.

What's the area in this case ?

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Functions of Random Variables

Determine $f_{UV}(u, v)$... direct method

General case: Use $x = au + bv$ and $y = cu + dv$

Area of the blue parallelogram:

$$(a \cdot du + b \cdot dv) \cdot (c \cdot du + d \cdot dv)$$

$$-2b \cdot dv \cdot c \cdot du - b \cdot dv \cdot d \cdot dv - a \cdot du \cdot c \cdot du$$

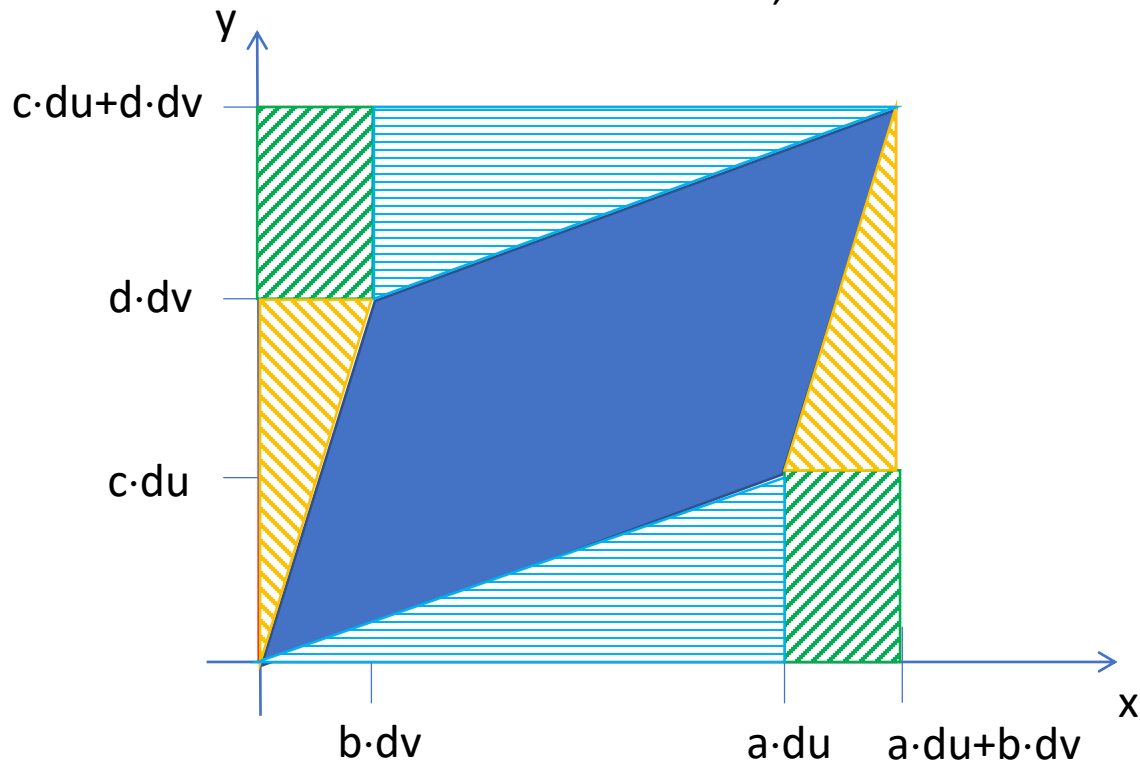
$$= (ad - bc) \cdot du \cdot dv = \begin{vmatrix} a & b \\ c & d \end{vmatrix} \cdot du \cdot dv$$

So, here the ratio is

$$\frac{\Delta(x, y)}{\Delta(u, v)} = \begin{vmatrix} a & b \\ c & d \end{vmatrix}$$

If the determinant happens to be negative, we need to take its absolute value.

Illustration for $du > 0, dv > 0$
 $a \cdot du > b \cdot dv$, and $d \cdot dv > c \cdot du$



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Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

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Jacobian matrix perspective:

Use $x_i = \varphi_i(y_1, \dots, y_n)$... locally, we have a linear mapping

$$\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y})$$

$$\mathbf{x} + d\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y} + d\mathbf{y}) \approx \boldsymbol{\varphi}(\mathbf{y}) + \mathbf{J} d\mathbf{y}$$

$$d\mathbf{x} = \mathbf{J} d\mathbf{y}$$

$$\mathbf{J} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial y_1} & \dots & \frac{\partial \varphi_1}{\partial y_n} \\ \vdots & & \vdots \\ \frac{\partial \varphi_n}{\partial y_1} & \dots & \frac{\partial \varphi_n}{\partial y_n} \end{bmatrix}$$

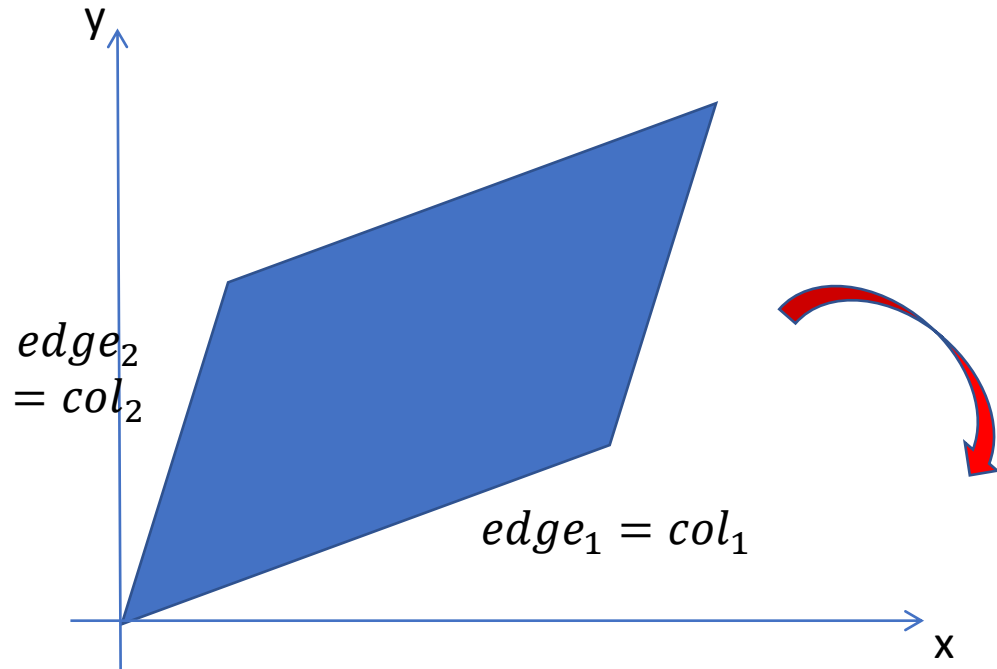
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What did we do?

$$J = \begin{bmatrix} * & * \\ * & * \end{bmatrix} \xrightarrow{\text{rotation}} \begin{bmatrix} * & * \\ 0 & * \end{bmatrix} = RJ$$

base height

Notice: Area = $|\det(RJ)| = \underbrace{\det(R)}_{=1} |\det(J)| = |\det(J)|$



Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

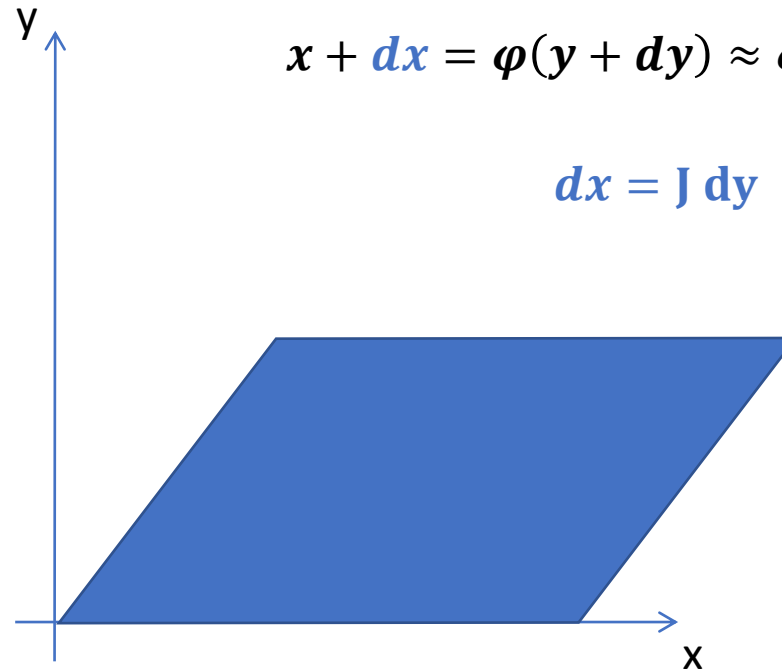
Jacobian matrix perspective – different perspectives even:

Use $x_i = \varphi_i(y_1, \dots, y_n)$... locally, we have a linear mapping

$$x = \varphi(y)$$

$$x + dx = \varphi(y + dy) \approx \varphi(y) + J dy$$

$$dx = J dy$$



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Higher dimensions ...

$$J = \begin{bmatrix} * & * & * \\ * & * & * \\ * & * & * \end{bmatrix} \xrightarrow{\text{rotation}} \begin{bmatrix} * & * & * \\ 0 & * & * \\ 0 & * & * \end{bmatrix} \xrightarrow{\text{rotation}} \begin{bmatrix} * & * & * \\ 0 & * & * \\ 0 & 0 & * \end{bmatrix}$$

Edge 1 to axis 1

Edge 2 to plane (1,2) ... keep angle with edge1 = axis 1

...

Area / Volume

$$= |\det(\dots R_2 R_1 J)| = |\det(J)|$$

Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

Jacobian matrix perspective:

Use $x_i = \varphi_i(y_1, \dots, y_n)$... locally, we have a linear mapping

$$\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y})$$

$$\mathbf{x} + d\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y} + d\mathbf{y}) \approx \boldsymbol{\varphi}(\mathbf{y}) + \mathbf{J} d\mathbf{y}$$

$$d\mathbf{x} = \mathbf{J} d\mathbf{y}$$

$$\mathbf{J} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial y_1} & \dots & \frac{\partial \varphi_1}{\partial y_n} \\ \vdots & & \vdots \\ \frac{\partial \varphi_n}{\partial y_1} & \dots & \frac{\partial \varphi_n}{\partial y_n} \end{bmatrix}$$

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Higher dimensions ...

Or via induction:

Just rotate the base ``area`` away from the last axis.

$$\begin{bmatrix} * & \dots & * & * \\ \vdots & & \vdots & \vdots \\ \vdots & & \vdots & \vdots \\ * & \dots & * & * \end{bmatrix} \xrightarrow{\text{rotation}} \begin{bmatrix} * & \dots & * & * \\ \vdots & & \vdots & \vdots \\ * & \dots & * & \vdots \\ 0 & \dots & 0 & * \end{bmatrix}$$

Intuition: Real 3-d box ...

Reasoning: Base is (n-1)-dim. ... rotate into the (n-1)-dim subspace spanned by axes 1, ..., n-1

To see the rotation it takes, consider the normal vectors (orth. complements) of those two (n-1)-dim. subspaces, and the angle in between.

Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

Jacobian matrix perspective:

Use $x_i = \varphi_i(y_1, \dots, y_n)$... locally, we have a linear mapping

$$\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y})$$

$$\mathbf{x} + d\mathbf{x} = \boldsymbol{\varphi}(\mathbf{y} + d\mathbf{y}) \approx \boldsymbol{\varphi}(\mathbf{y}) + \mathbf{J} d\mathbf{y}$$

$$d\mathbf{x} = \mathbf{J} d\mathbf{y}$$

$$\mathbf{J} = \begin{bmatrix} \frac{\partial \varphi_1}{\partial y_1} & \dots & \frac{\partial \varphi_1}{\partial y_n} \\ \vdots & & \vdots \\ \frac{\partial \varphi_n}{\partial y_1} & \dots & \frac{\partial \varphi_n}{\partial y_n} \end{bmatrix}$$

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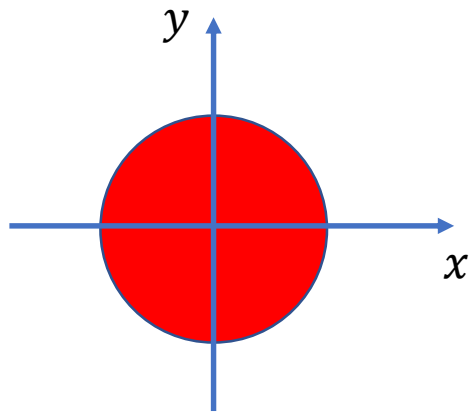
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Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

Example 1:

Recall our earlier example:

... Change integration variables:

$$\begin{aligned} x &= r \cos \theta \\ y &= r \sin \theta \end{aligned} \Rightarrow dx dy \rightarrow r dr d\theta$$

Jacobian matrix:

$$J = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix} \Rightarrow \det(J) = r(\cos^2 \theta + \sin^2 \theta) = r$$

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Jacobian matrix:

$$J = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix} \Rightarrow \det(J) = -\frac{1}{2} \Rightarrow |\det(J)| = \frac{1}{2}$$

Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

Example 2:

$$V = g(X, Y) = X + Y, \quad W = h(X, Y) = X - Y$$

$$\Leftrightarrow X = \frac{V + W}{2}, \quad Y = \frac{V - W}{2}$$

→

$$f_{VW}(v, w) = \frac{1}{2} f_{XY}\left(\frac{v + w}{2}, \frac{v - w}{2}\right)$$

Mind:

$$f_{VW}(v, w) = f_{XY}(x(v, w), y(v, w)) \frac{\Delta(x, y)}{\Delta(v, w)}$$

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Jacobian matrix:

$$\frac{\Delta(v, w)}{\Delta(x, y)} = \left| \det \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \right| = 2$$

Functions of Random Variables

Determine densities like $f_{UV}(u, v)$... direct method

Example 2 - Alternative Perspective:

$$V = g(X, Y) = X + Y, \quad W = h(X, Y) = X - Y$$

$$\Leftrightarrow X = \frac{V + W}{2}, \quad Y = \frac{V - W}{2}$$

→

$$f_{VW}(v, w) = \frac{1}{2} f_{XY} \left(\frac{v + w}{2}, \frac{v - w}{2} \right)$$

Mind:

$$f_{VW}(v, w) = f_{XY}(x(v, w), y(v, w)) \frac{\Delta(x, y)}{\Delta(v, w)}$$

The End

Next time: Recap Chp. 3
& Exam-type tasks