

# MATH 451/551

## Chapter 6. Joint Distribution

### 6.3 Joint Moment Generating Functions

GuanNan Wang  
[gwang01@wm.edu](mailto:gwang01@wm.edu)



# Joint Moment Generating Functions



## Joint Moment Generating Functions

Let  $(X_1, X_2)$  be a random vector. The **joint moment generating function** of  $(X_1, X_2)$  is

$$M(t_1, t_2) = E(e^{t_1 X_1 + t_2 X_2})$$

provided that the expected value exists on  $-h_1 < t_1 < h_1$  and  $-h_2 < t_2 < h_2$  for some positive constants  $h_1$  and  $h_2$ .

- $E(X_1) = \frac{\partial M(t_1, t_2)}{\partial t_1} \Big|_{(t_1, t_2) = (0, 0)}, \quad E(X_2) = \frac{\partial M(t_1, t_2)}{\partial t_2} \Big|_{(t_1, t_2) = (0, 0)}$
- $E(X_1^2) = \frac{\partial^2 M(t_1, t_2)}{\partial t_1^2} \Big|_{(t_1, t_2) = (0, 0)}, \quad E(X_2^2) = \frac{\partial^2 M(t_1, t_2)}{\partial t_2^2} \Big|_{(t_1, t_2) = (0, 0)}$
- $E(X_1 X_2) = \frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} \Big|_{(t_1, t_2) = (0, 0)}$
- Marginal moment generating functions:  
 $M_{X_1}(t_1) = E(e^{t_1 X_1}) = M(t_1, 0), \quad M_{X_2}(t_2) = E(e^{t_2 X_2}) = M(0, t_2)$
- The random variables  $X_1$  and  $X_2$  are independent iff

$$M(t_1, t_2) = M_{X_1}(t_1)M_{X_2}(t_2)$$

# Example 1



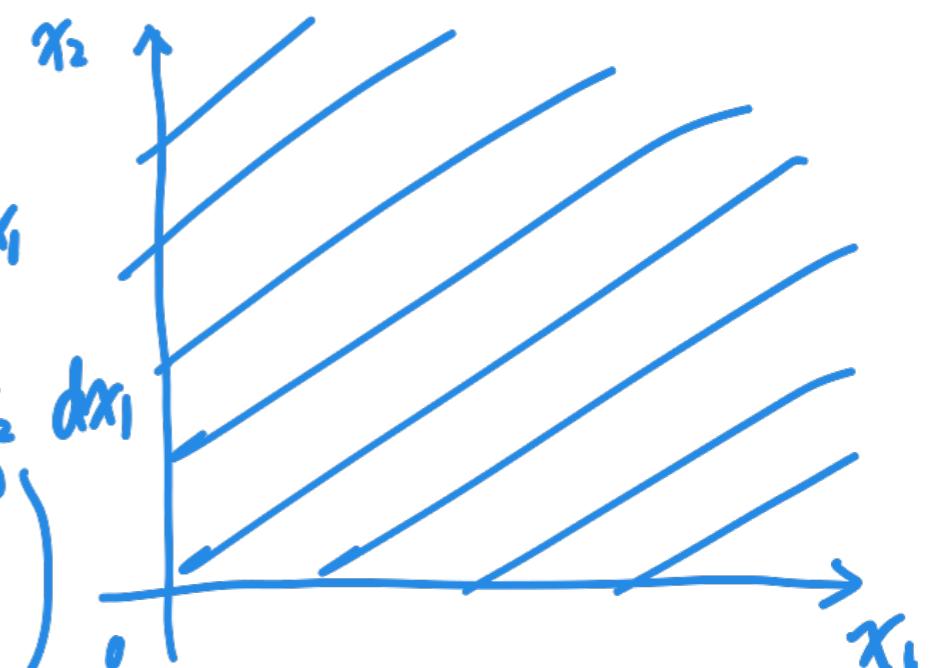
## Example 1

Let  $X_1$  and  $X_2$  be continuous random variables whose distribution is described by the joint probability density function

$$f(x_1, x_2) = \lambda_1 e^{-\lambda_1 x_1} \lambda_2 e^{-\lambda_2 x_2}, \quad x_1 > 0, x_2 > 0$$

for real positive parameters  $\lambda_1$  and  $\lambda_2$ . Find the joint moment generating function for  $X_1$  and  $X_2$  and use it to find  $E(X_1 X_2)$ .

$$\begin{aligned}
 M(t_1, t_2) &= E\left(e^{t_1 X_1 + t_2 X_2}\right) = \int_0^\infty \int_0^\infty e^{t_1 x_1 + t_2 x_2} f(x_1, x_2) dx_2 dx_1 \\
 &= \int_0^\infty \int_0^\infty e^{t_1 x_1} e^{t_2 x_2} \lambda_1 e^{-\lambda_1 x_1} \lambda_2 e^{-\lambda_2 x_2} dx_2 dx_1 \\
 &= \int_0^\infty e^{t_1 x_1} \lambda_1 e^{-\lambda_1 x_1} \int_0^\infty \lambda_2 e^{-(\lambda_2 - t_2) x_2} dx_2 dx_1 \\
 &= \int_0^\infty \lambda_1 e^{-(\lambda_1 - t_1) x_1} dx_1 \left( -\frac{\lambda_2 e^{-(\lambda_2 - t_2) x_2}}{\lambda_2 - t_2} \Big|_0^\infty \right) \\
 &\stackrel{t_1 < \lambda_1}{=} \frac{\lambda_1}{\lambda_1 - t_1} \frac{\lambda_2}{\lambda_2 - t_2}
 \end{aligned}$$



$$E(X_1 X_2) = \left. \frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} \right|_{t_1=0, t_2=0}$$

$$= \left. \frac{\partial^2}{\partial t_1 \partial t_2} \left( \frac{\lambda_1}{\lambda_1 - t_1} \frac{\lambda_2}{\lambda_2 - t_2} \right) \right|_{t_1=0, t_2=0}$$

$$= \left. \frac{+\lambda_1}{(\lambda_1 - t_1)^2} * \frac{\lambda_2}{(\lambda_2 - t_2)^2} \right|_{t_1=0, t_2=0}$$

$$= \frac{\lambda_1}{\lambda_1^2} \frac{\lambda_2}{\lambda_2^2} = \frac{1}{\lambda_1 \lambda_2}$$

# Thank You



**THANK YOU!**

