

Study aid Probability 2 (the formulas on the formula sheet are given)

**Definitions:**

<b>Discrete:</b> $E(X) =$ <b>Continuous:</b> $E(X) =$	$cov(X, Y) =$
$var(X) =$	$\rho(X, Y) =$

**Calculation formulas :**

<b>Discrete:</b> $E(X^2) =$ <b>Continuous:</b> $E(X^2) =$	$var(X) =$
$E(XY) =$	$cov(X, Y) =$
$E(X Y = y) =$	$E(X) = E[ \quad ]$ $= \sum_{\dots} \dots \dots \dots \dots \dots \dots \dots$

**Properties  $E(X)$ ,  $var(X)$ ,  $cov(X, Y)$  and  $\rho(X, Y)$ :**

$E(aX + b) =$	$var(aX + b) =$
$E(X + Y) =$ $E(X - Y) =$	$var(X + Y) =$ $var(X - Y) =$
$E(\sum_{i=1}^n X_i) =$	$var\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n var(X_i) + \sum_{i \neq j} \sum_j cov(X_i, X_j)$
$cov(aX, Y) =$	$\rho(aX, Y) =$ $cov(X, X) =$
$cov(X, Y + Z) =$	

**Independence**

**Discrete:**  $X$  and  $Y$  are independent  $\Leftrightarrow P(X = x \text{ and } Y = y) = \dots \dots \dots$

**General:**  $X$  and  $Y$  are independent  $\Leftrightarrow P(X \in A \text{ and } Y \in B) = \dots \dots \dots$

<b>If <math>X</math> and <math>Y</math> are independent, then we have:</b>	$E(XY) =$ $cov(X, Y) =$ and $\rho(X, Y) =$ $var(X \pm Y) =$
<b>If <math>X_1, \dots, X_n</math> are independent and identically distributed with expectation <math>\mu</math> and variance <math>\sigma^2</math>, then:</b>	$var(\sum_{i=1}^n X_i) =$  $E(\bar{X}) =$ and $var(\bar{X}) =$

**Convolution-sum (Ch. 5):**  $P(X + Y = k) =$

**Convolution-integral (Ch. 7):**  $f_{X+Y}(z) =$

## Study aid 2

### Definitions:

Discrete: $E(X) = \sum_x xP(X = x)$	$cov(X, Y) = E(X - EX)(Y - EY)$
Continuous: $E(X) = \int_{-\infty}^{\infty} xf(x)dx$	
$var(X) = E(X - EX)^2$	$\rho(X, Y) = \frac{cov(X, Y)}{\sigma_X \sigma_Y}$

### Calculation formulas:

Discrete: $E(X^2) = \sum_x x^2 P(X = x)$	$var(X) = E(X^2) - (EX)^2$
Continuous: $E(X^2) = \int_{-\infty}^{\infty} x^2 f(x)dx$	
$E(XY) = \sum_x \sum_y x \cdot y \cdot P(X = x \text{ and } Y = y)$	$cov(X, Y) = E(XY) - E(X)E(Y)$
$E(X Y = y) = \sum_x x \cdot P(X = x Y = y)$	$E(X) = E[E(X Y)]$ $= \sum_y E(X Y = y) \cdot P(Y = y)$

### Properties $E(X)$ , $var(X)$ , $cov(X, Y)$ and $\rho(X, Y)$ :

$E(aX + b) = aE(X) + b$	$var(aX + b) = a^2 var(X)$
$E(X + Y) = E(X) + E(Y)$ $E(X - Y) = E(X) - E(Y)$	$var(X + Y) = var(X) + var(Y) + 2cov(X, Y)$ $var(X - Y) = var(X) + var(Y) - 2cov(X, Y)$
$E(\sum_{i=1}^n X_i) = \sum_{i=1}^n E(X_i)$	$var\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n var(X_i) + \sum_{i \neq j} cov(X_i, X_j)$
$cov(aX, Y) = a \cdot cov(X, Y)$	$\rho(aX, Y) = \rho(X, Y), a > 0$   $cov(X, X) = var(X)$
$cov(X, Y + Z) = cov(X, Y) + cov(X, Z)$	

### Independence Discrete:

$X$  and  $Y$  are independent  $\Leftrightarrow P(X = x \text{ and } Y = y) = P(X = x) \cdot P(Y = y)$ ,

### General:

$X$  and  $Y$  are independent  $\Leftrightarrow P(X \in A \text{ and } Y \in B) = P(X \in A) \cdot P(Y \in B)$

If $X$ and $Y$ are independent, then we have:	$E(XY) = E(X)E(Y)$ $cov(X, Y) = 0$ and $\rho(X, Y) = 0$ $var(X \pm Y) = var(X) + var(Y)$
If $X_1, \dots, X_n$ are independent and identically distributed with expectation $\mu$ and variance $\sigma^2$ , then:	$var(\sum_{i=1}^n X_i) = \sum_{i=1}^n var(X_i) = n\sigma^2$ $E(\bar{X}) = \mu$ and $var(\bar{X}) = \frac{\sigma^2}{n}$

Convolution-sum:  $P(X + Y = k) = \sum_k P(X = k) \cdot P(Y = n - k)$

Convolution-integral:  $f_{X+Y}(z) = \int_{-\infty}^{\infty} f_X(x)f_Y(z - x)dx$