

①

A definite statement could be this:

Suppose $E\left(\sum_{n=1}^{\infty} X_n\right) < \infty$. Then $\sum_{n=1}^{\infty} EX_n < \infty$ and the two quantities are equal.

Indeed, consider the random variables

$$S_N = \sum_{n=1}^N X_n \quad (\text{partial sums}).$$

Since $X_n \geq 0$, we have $S_N \geq 0$ and $S_N \uparrow S := \sum_{n=1}^{\infty} X_n \quad (N \rightarrow \infty)$

By the Monotone Convergence Theorem,

$$ES_N \rightarrow ES \quad (N \rightarrow \infty)$$

||

$$\sum_{n=1}^{\infty} EX_n.$$

Q.E.D.

②

Let $t \geq 1$.

Let $p \in (0, 1)$ be a parameter to be chosen later, and define a r.v.

$$X = \begin{cases} 1 & \text{with probability } p/2 \\ -1 & \text{with probability } p/2 \\ 0 & \text{with probability } 1-p \end{cases}$$

Then $\mu = EX = 0$, $EX^2 = 1^2 p/2 + (-1)^2 p/2 = p$, so $\sigma^2 = \text{var}(X) = p$.

Now, $P(|X - \mu| \geq t\sigma) = P(X \neq 0) = p$

To make the right hand side as in Chebychev's inequality, we choose $p := 1/t^2 \leq 1$ (by the assumption). Q.E.D.

③

$$\int_{\mathbb{R}} P(x < X < x+a) dx = \int_{\mathbb{R}} E \mathbb{1}_{\{x < X < x+a\}} dx$$

$$= E \int_{\mathbb{R}} \mathbb{1}_{\{x < X < x+a\}}(x) dx \quad \text{by Fubini}$$

$$= E a = a.$$

④

First, we observe that

$$E|X - \frac{1}{2}| = \int_0^1 |x - \frac{1}{2}| dx = \frac{1}{4},$$

and similarly $E|Y - \frac{1}{2}| = \frac{1}{4}.$

Since by the triangle inequality,

$$|X - Y| = |(X - \frac{1}{2}) - (Y - \frac{1}{2})| \leq |X - \frac{1}{2}| + |Y - \frac{1}{2}|,$$

we have

$$E|X - Y| \leq E|X - \frac{1}{2}| + E|Y - \frac{1}{2}| \leq \frac{1}{4} + \frac{1}{4} = \frac{1}{2}.$$

(5)

I will explain the solution for $n=2$; the general case is similar.

So, let X, Y be random variables such that

$$P((X, Y) \in D) = \int_D f(x)g(y) dx dy \quad \text{for } D \subseteq \mathbb{R}^2.$$

Use this for $D = A \times B \Rightarrow$ by Fubini

$$(*) \quad P(X \in A, Y \in B) = \int_A f(x) dx \cdot \int_B g(y) dy \quad \text{for } A, B \subseteq \mathbb{R}.$$

As we are told, f and g may not be densities, that is $a = \int_{\mathbb{R}} f \neq 1$, $b = \int_{\mathbb{R}} g \neq 1$, but nevertheless for $A=B=\mathbb{R}$ we get

$$1 = P(X \in \mathbb{R}, Y \in \mathbb{R}) = ab.$$

For $B=\mathbb{R}$, we get from (*).

$$P(X \in A) = \int_A f(x) dx \cdot b;$$

and for $A=\mathbb{R}$, we get

$$P(Y \in B) = \int_B g(y) dy \cdot a.$$

Multiplying these two identities, and using $ab=1$, we get

$$P(X \in A)P(Y \in B) = \int_A f(x) dx \cdot \int_B g(y) dy.$$

Together with (*) this completes the proof.