

CS–E4500 Advanced Course in Algorithms

Week 06 – Tutorial

1. The events A_1, A_2, A_3 are pairwise independent if, for all $i \neq j$, A_i is independent of A_j . However, pairwise independence is a weaker statement than mutual independence, which requires the additional condition that $P(A_1, A_2, A_3) = P(A_1)P(A_2)P(A_3)$. Construct an example where three events are pairwise independent but not mutually independent.

Solution. One example would concern two independent coin tosses. Let

- A_1 : first toss is Head
- A_2 : second toss is Head
- A_3 : both tosses are the same

Clearly $P(A_i | A_j) = P(A_i)$ for all $i, j \in \{1, 2, 3\}$ such that $i \neq j$. However, since events A_1 and A_2 determine event A_3 , they are not mutually independent.

2. Let $X = \sum_{i=1}^n X_i$, where the X_i are pairwise independent random variables. Prove that

$$\text{Var}(X) = \sum_{i=1}^n \text{Var}(X_i) .$$

This equality allows us to apply Chebyshev's inequality even when the random variables are only pairwise independent.

Solution. By a well known relation,

$$\text{Var} \left(\sum_{i=1}^n X_i \right) = \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{i < j} \text{Cov}(X_i, X_j) ,$$

where

$$\text{Cov}(X_i, X_j) = E((X_i - E(X_i))(X_j - E(X_j))) = E(X_i X_j) - E(X_i)E(X_j) .$$

Since $X_1, X_2, \dots, X_n$ are pairwise independent, it is clear that for any $i \neq j$ we have

$$E(X_i X_j) - E(X_i)E(X_j) = 0 ,$$

and therefore

$$\text{Var}(X) = \sum_{i=1}^n \text{Var}(X_i) .$$