

# Lecture Summary: Multiple Random Variables and Joint PMFs

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## Key Points

- **Transition to Multiple Random Variables:**

- Random variables simplify probability spaces by mapping complex outcomes to numerical values.
- Multiple random variables often arise in experiments with interconnected outcomes, requiring tools to analyze their relationships.

- **Examples of Multiple Random Variables:**

- **Coin Toss Experiment:**

- \* Toss a fair coin three times, defining random variables  $X_1$ ,  $X_2$ , and  $X_3$  to indicate whether the first, second, and third tosses are heads.
- \* These are indicator random variables, taking values 1 for heads and 0 for tails.
- \* Events based on  $X_1$  (e.g.,  $X_1 = 1$ ) are independent of events based on  $X_2$  or  $X_3$ .

- **Two-Digit Lottery:**

- \* Random variable  $X$ : Units digit of the number.
- \* Random variable  $Y$ : Remainder when the number is divided by 4.
- \*  $X$  and  $Y$  are uniformly distributed over their respective ranges, but they are not independent.
- \* Example: If  $X = 1$ ,  $Y = 0$  is impossible because numbers ending in 1 are not multiples of 4.

- **IPL Powerplay Data:**

- \* Random variable  $X$ : Total runs scored in an over.
- \* Random variable  $Y$ : Number of wickets lost in the over.
- \* A relationship exists: Overs with more wickets typically have fewer runs.

- **Modeling Multiple Random Variables:**

- Relationships between random variables help model complex experiments (e.g., IPL data).
- Dependencies are modeled probabilistically rather than deterministically.

- **Joint PMFs (Probability Mass Functions):**

- For discrete random variables  $X$  and  $Y$  defined on the same probability space:

$$f_{XY}(t_1, t_2) = P(X = t_1 \text{ and } Y = t_2).$$

- Properties:

- \*  $f_{XY}(t_1, t_2) \geq 0$  for all  $t_1, t_2$ .
- \*  $\sum_{t_1, t_2} f_{XY}(t_1, t_2) = 1$ .

- **Examples of Joint PMFs:**

– **Coin Toss:**

- \*  $X_1$ : First toss indicator,  $X_2$ : Second toss indicator.
- \* Joint PMF:  $f_{X_1 X_2}(t_1, t_2) = P(X_1 = t_1 \text{ and } X_2 = t_2) = (1/2) \cdot (1/2) = 1/4$  for independent tosses.

– **Two-Digit Lottery:**

- \*  $X$ : Units digit,  $Y$ : Remainder modulo 4.
- \* Example:  $f_{XY}(0, 0) = P(X = 0 \text{ and } Y = 0)$  includes numbers like 00, 20, 40, ..., totaling 5 outcomes:  $5/100 = 1/20$ .

• **Properties of Joint PMFs:**

- All probabilities in the PMF lie between 0 and 1.
- The sum of all probabilities in the PMF equals 1, covering all possible outcomes.
- PMFs can be represented as tables or matrices for easier interpretation and computation.

## Simplified Explanation

**Multiple Random Variables:** These arise when experiments have interconnected outcomes. Examples include: - Coin tosses (independent variables). - Two-digit lottery numbers (dependent variables). - IPL data (complex dependencies).

**Joint PMFs:** Functions that assign probabilities to pairs of values taken by two random variables, satisfying:

$$f_{XY}(t_1, t_2) \geq 0, \quad \sum f_{XY}(t_1, t_2) = 1.$$

## Conclusion

In this lecture, we:

- Explored multiple random variables and their applications.
- Defined joint PMFs and derived examples.
- Discussed how dependencies between variables help model real-world scenarios.

Joint PMFs are crucial for analyzing and modeling relationships between random variables in complex experiments.