



6. Mutual Information

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Mutual Information (MI) is a measure of the dependency between two variables. It quantifies the amount of information obtained about one random variable through observing the other random variable. It is a fundamental quantity in information theory.

$$MI = \sum_{x \in X} \sum_{y \in Y} p(x, y) \log \left[\frac{p(x, y)}{p(x) p(y)} \right]$$

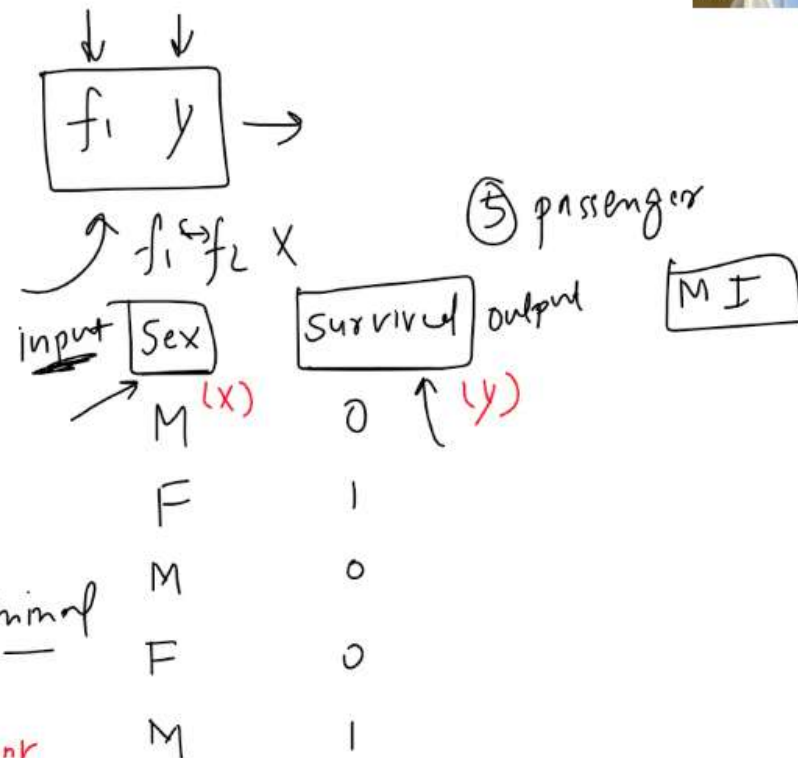
where

$p(x, y) \rightarrow$ Joint prob of X and Y

$p(x) \rightarrow$ marginal prob of X

$p(y) \rightarrow$ marginal prob of Y

$p(x, y)$ $p(x)$
 \downarrow
 marginal
 $p(x, y)$ \rightarrow joint
 $p(x) p(y)$ \rightarrow marginal



1. **Joint Probability:** This is the probability of two (or more) simultaneous events. For example, if we have two random variables, X and Y, the joint probability of X and Y is denoted as $P(X, Y)$, and it represents the probability that X takes on a specific value and Y takes on a specific value at the same time. In other words, it represents the probability of both events happening at the same time.

2. **Marginal Probability:** This is the probability of an event occurring regardless of the

ity of an event occurring regardless of the



$P(X=m, Y=0)$

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	0	1	
M	$\frac{2}{5} +$	$\frac{1}{5}$	$\frac{3}{5}$ ←
F	$\frac{1}{5} +$	$\frac{1}{5}$	$\frac{2}{5}$ ↗
	$\frac{3}{5}$ ↘	$\frac{2}{5}$ ↗	



the process of summing or integrating over the distribution of the other variable(s) to obtain the distribution of the variable of interest.

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Mutual Information has several properties that make it useful for feature selection:

1. It is non-negative: MI is always zero or positive, with zero indicating that the variables are independent (i.e., no information about one variable can be obtained by observing the other variable).
2. It is symmetric: $MI(X, Y) = MI(Y, X)$. The mutual information from X to Y is the same as from Y to X.
3. It can capture any kind of statistical dependency: Unlike correlation, which only captures linear relationships, mutual information can capture any kind of relationship, including nonlinear ones.

How to deal with numerical variables