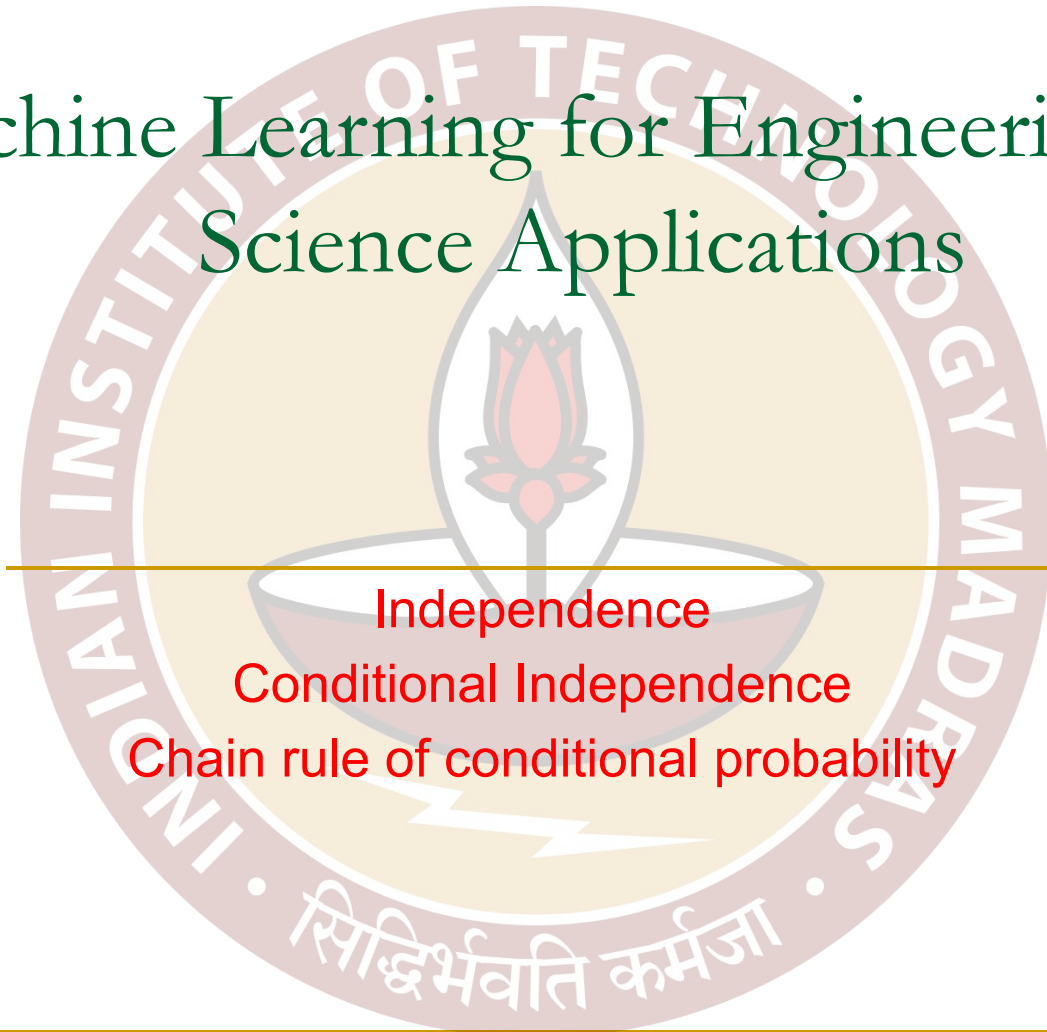


Machine Learning for Engineering and Science Applications



Independence

Conditional Independence

Chain rule of conditional probability

Independence

- **Independent random variables** – Two random variables X and Y are said to be *statistically independent* if and only if

$$p(x, y) = p(x)p(y)$$

- More precisely, X and Y are independent iff
$$p(X = x, Y = y) = p(X = x)p(Y = y) \quad \forall x \in X, y \in Y$$
- Examples
 - Independent – X : Throw of a dice, Y : Toss of a coin
 - Not independent – X : Height, Y : Weight
- Independence is equivalent to saying
$$p(y|x) = p(y) \text{ OR } p(x|y) = p(x)$$
- Can be seen from product rule $p(x, y) = p(y|x)p(x) = p(x)p(y)$
$$\Rightarrow p(y|x) = p(y)$$

Conditional Independence

- Two random variables **X** and **Y** are said to be *independent given z* if and only if

$$p(x, y | z) = p(x|z)p(y|z)$$

- More precisely, **X** and **Y** are independent given **Z** iff

$$p(X = x, Y = y | Z = z) = p(X = x | Z = z)p(Y = y | Z = z) \quad \forall x \in X, y \in Y, z \in Z$$

- Examples

- *Ind & Cond Ind* – **X**: Throw of a dice, **Y**: Toss of a coin, **Z**: Card from deck
- *Not Ind BUT Cond Ind* – **X**: Height, **Y**: Vocabulary, **Z**: Age
- *Ind BUT Cond Not Ind* – **X**: Dice Throw 1, **Y**: Dice Throw 2, **Z**: Sum of dice

Conditional Probability – Chain rule

- Recall -- $p(x, y) = p(x|y)p(y)$

Consider $p(x, y, z) = p(x, a)$ where a is the event (y, z)

$$\begin{aligned}\Rightarrow p(x, y, z) &= p(x, a) = p(x|a)p(a) \\ &= p(x|a)p(y, z) \\ &= p(x|a)p(y|z)p(z) \\ &= p(x|y, z)p(y|z)p(z) \\ &= p(z)p(y|z)p(x|y, z)\end{aligned}$$

- In general,

- $P(x^{(1)}, x^{(2)}, \dots, x^{(n)}) = P(x^{(1)})P(x^{(2)}|x^{(1)}) \dots P(x^{(n)}|x^{(1)}, \dots, x^{(n-1)})$

i.e.

$$P(x^{(1)}, x^{(2)}, \dots, x^{(n)}) = P(x^{(1)}) \prod_{i=2}^n P(x^{(i)}|x^{(1)}, \dots, x^{(i-1)})$$

Chain rule of conditional probability

One context for conditional probabilities



- Images may be thought of as a collection of pixels $x^{(1)}, x^{(2)}, \dots, x^{(n)}$
- The probability of a particular image may be thought of as joint probability

$$P(x^{(1)}, x^{(2)}, \dots, x^{(n)})$$

- Chain rule along with conditional independence can be used to estimate probabilities of the occurrence of images