## Mathematics for AI

## Lecture 21

Joint distribution, Marginal Distribution, Conditional Distribution, Probabilistic Modeling, Generative vs Discriminative Models

Adding two random variables (general case)
X: How many Mehran sells
Y: How "Mild sells
$Z$. How many both sell
$Z=X+Y=$ ? (what if they are not independent?
$Z=f(X, Y)=$ ? $\quad p(X) \& p(Y)$ won't help!

## The joint probability distribution

- Question
- $\operatorname{Pr}(X=a)=1 / 2$ $\operatorname{Pr}(Y=b)=1 / 4$
- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?


## The joint probability distribution

K. N. Toosi

- Question
- $\operatorname{Pr}(X=a)=1 / 2$

$$
\operatorname{Pr}(Y=b)=1 / 4
$$

- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?

First Scenario: Roll a (fair) dice twice

- $X$ : first number, $Y$ : second number
- $\operatorname{Pr}(X=6$ AND $Y=1)=$ ?


## The joint probability distribution

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\operatorname{Pr}(Y=b)=1 / 4
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- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?

First Scenario: Roll a (fair) dice twice

- $X$ : first number, $Y$ : second number
- $\operatorname{Pr}(X=6$ AND $Y=1)=\operatorname{Pr}(X=6) P(Y=1)=1 / 6 * / 6=1 / 36$



## The joint probability distribution

- Question
- $\operatorname{Pr}(X=a)=1 / 2$

$$
\operatorname{Pr}(Y=b)=1 / 4
$$

- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?

Second Scenario (missing exam => failing course)

- $\operatorname{Pr}($ miss exam session $)=0.01$

$$
\operatorname{Pr}(\text { fail course })=0.08
$$

- $\operatorname{Pr}$ (miss exam session AND fail course) $=$ ?


## The joint probability distribution

- Question
- $\operatorname{Pr}(X=a)=1 / 2$

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\operatorname{Pr}(Y=b)=1 / 4
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- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?

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- $\operatorname{Pr}($ miss exam session AND fail course $)=0.01$


## The joint probability distribution

- Question
- $\operatorname{Pr}(X=a)=1 / 2$

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\operatorname{Pr}(Y=b)=1 / 4
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- what is $\operatorname{Pr}(X=a$ AND $Y=b)$ ?

Second Scenario (missing exam => failing course)

- $\operatorname{Pr}($ miss exam session $)=0.01$

$$
\operatorname{Pr}(\text { fail course })=0.08
$$

How the first and second scenarios differ?

- $\operatorname{Pr}($ miss exam session AND fail course $)=0.01$


## The joint probability distribution

- The probability of co-occurrence.
- If we have two random variables $X, Y$ we cannot model the system using $\operatorname{Pr}(X=x)$ and $\operatorname{Pr}(Y=y)$.
- Probability mass function (Discrete Variables)
- $p(x, y)=\operatorname{Pr}(X=x$ AND $Y=y)=\operatorname{Pr}(X=x, Y=y)$


## The joint probability distribution

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- Probability mass function (Discrete Variables)
- $p(x, y)=\operatorname{Pr}(X=x$ AND $Y=y)=\operatorname{Pr}(X=x, Y=y)$
- Probability Density function (Continuous Variables)
- $p(x, y)$



## The joint probability distribution

- Probability Density function (Continuous Variables)

$$
\operatorname{Pr}((x, y) \in S)=\int_{S} p(x, y) d x d y
$$



## Generalize Relations



## Probabilistic Modelling

- System variables $X_{1}, X_{2}, \ldots, X_{N}$
- Generative Model: Joint distribution p $\left(x_{1}, x_{2}, \ldots, x_{N}\right)$
- If you have the joint distribution, you have everything
- Prediction:
- Having $p(x, y, z)=\operatorname{Pr}(X=x, y=y, Z=z)$, predict $x, y, z$


## Remember: The joint probability distribution

- The probability of co-occurrence.
- Probability mass function (Discrete Variables)
- $p(x, y)=\operatorname{Pr}(X=x$ AND $Y=y)=\operatorname{Pr}(X=x, Y=y)$
- Probability Density function (Continuous Variables) - $p(x, y)$



## Remember: Probabilistic Modelling

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- System variables $X_{1}, X_{2}, \ldots, X_{N}$
- Generative Model: Joint distribution p $\left(x_{1}, x_{2}, \ldots, x_{N}\right)$
- If you have the joint distribution, you have everything
- Prediction:
- Having $p(x, y, z)=\operatorname{Pr}(X=x, y=y, Z=z)$, predict $x, y, z$
- Find the most likely configuration of system variables

$$
x^{*}, y^{*}, z^{*}=\arg \max _{x, y, z} p(x, y, z)
$$

## Remember: Probabilistic Modelling

K. N. Toosi

- System variables $X_{1}, X_{2}, \ldots, X_{N}$
- Generative Model: Joint distribution p $\left(x_{1}, x_{2}, \ldots, x_{N}\right)$
- If you have the joint distribution, you have everything
- Prediction:
- Having $p(x, y, z)=\operatorname{Pr}(X=x, y=y, Z=z)$
- If we know $Z=Z_{0}$, predict $x, y$

$$
x, y=\arg \max _{x, y} p\left(x, y, z_{0}\right)
$$

## Generative Model



1. learning/modeling:

- find $p\left(x_{1}, x_{2}, \ldots, x_{m}, y_{1}, y_{2}, \ldots, y_{n}\right)$

2. prediction/testing
$y_{1}^{*}, y_{2}^{*}, \ldots, y_{n}^{*}=\arg \max _{y_{1}, \ldots, y_{n}} p\left(x_{1}, \ldots, x_{m}, y_{1}, \ldots, y_{n}\right)$

## Example:

The joint probability of

- having a rainfall in an hour, and
- the sky being cloudy at the moment
- $p(r, c)=\operatorname{Pr}(R=r, C=c)$

| $r$ (rain) | c (cloudy) | $\operatorname{Pr}(R=r, C=c)$ |
| :---: | :---: | :---: |
| 0 | 0 | 0.75 |
| 0 | 1 | 0.10 |
| 1 | 0 | 0.05 |
| 1 | 1 | 0.10 |

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## Question

- Having the joint distribution $\operatorname{Pr}(R=r, C=c)$, what is $\operatorname{Pr}(R=r)$ ?

$$
\operatorname{Pr}(R=r)=?
$$

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$$
\operatorname{Pr}(R=r)=\operatorname{Pr}((R=r \text { AND } C=0) O R(R=r \text { AND } C=1))
$$

## Question

- Having the joint distribution $\operatorname{Pr}(R=r, C=c)$, what is $\operatorname{Pr}(R=r)$ ?

$$
\begin{aligned}
\operatorname{Pr}(R=r) & =\operatorname{Pr}((R=r \text { AND } C=0) O R(R=r \text { AND } C=1)) \\
& =\operatorname{Pr}(R=r \text { AND } C=0)+\operatorname{Pr}(R=r \text { AND } C=1) \quad(w h y ?)
\end{aligned}
$$

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- Having the joint distribution $\operatorname{Pr}(R=r, C=c)$, what is $\operatorname{Pr}(R=r)$ ?

$$
\begin{align*}
\operatorname{Pr}(R=r) & =\operatorname{Pr}((R=r \text { AND } C=0) O R(R=r \text { AND } C=1)) \\
& =\operatorname{Pr}(R=r \text { AND } C=0)+\operatorname{Pr}(R=r \text { AND } C=1) \tag{why?}
\end{align*}
$$

$$
\begin{aligned}
& \operatorname{Pr}(R=r, C=c) \\
& \begin{array}{l|l|l} 
& \mathrm{R}=0 & \mathrm{R}=1 \\
\hline \mathrm{C}=0 & 0.75 & 0.05 \\
\mathrm{C}=1 & 0.10 & 0.10
\end{array}
\end{aligned}
$$

$$
\begin{gathered}
\operatorname{Pr}(R=r) \\
\begin{array}{c|c|c} 
\\
\mathrm{R}=0 & \mathrm{R}=1 \\
\hline 0.85 & 0.15
\end{array}
\end{gathered}
$$

## Marginal Distribution

- Discrete: probability mass function $p(m, n)=\operatorname{Pr}(M=m, N=n)$

$$
p(m)=\operatorname{Pr}(M=m)=\sum_{n} p(m, n)
$$

- Continuous: probability density function $p(x, y)$

$$
p(x)=\int p(x, y) d y
$$

## Marginal Probability

| $P(x, y)$ | $x=0$ | $x=1$ | $x=2$ | row sum |
| :--- | ---: | ---: | ---: | ---: |
| $y=0$ | 0.32 | 0.03 | 0.01 | 0.36 |
| $y=1$ | 0.06 | 0.24 | 0.02 | $\mathbf{0 . 3 2}$ |
| $y=2$ | 0.02 | 0.03 | 0.27 | $\mathbf{0 . 3 2}$ |
| col sum | $\mathbf{0 . 4 0}$ | $\mathbf{0 . 3 0}$ | $\mathbf{0 . 3 0}$ | checksum $=1.0$ | image from http://stats.stackexchange.com

## Marginal Probability


image from www.wolfram.com

## Question

- What is the probability of having a rainfall today?

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- What is the probability of having a rainfall today?

$$
\operatorname{Pr}(R=1)=\operatorname{Pr}(R=1, C=0)+\operatorname{Pr}(R=1, C=1)
$$

$\operatorname{Pr}(R=r, C=c)$

|  | $\mathrm{R}=0$ | $\mathrm{R}=1$ |
| :--- | :--- | :--- |
| $\mathrm{C}=0$ | 0.75 | 0.05 |
| $\mathrm{C}=1$ | 0.10 | 0.10 |
|  |  |  |

## Question

- What is the probability of having a rainfall today?

$$
\operatorname{Pr}(R=1)=\operatorname{Pr}(R=1, C=0)+\operatorname{Pr}(R=1, C=1)=0.05+0.10=0.15
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$$
\begin{aligned}
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& \begin{array}{l|l||l|} 
& \mathrm{R}=0 & \mathrm{R}=1 \\
\hline \mathrm{C}=0 & 0.75 & 0.05 \\
\mathrm{C}=1 & 0.10 & 0.10 \\
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\end{aligned}
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## Question

- What is the probability of having a rainfall today?

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- If we know the sky is cloudy, what is the probability of having a rainfall today?

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$$
\operatorname{Pr}(R=1 \mid C=1)=0.10 /(0.10+0.10)=0.5
$$

$$
=\frac{\operatorname{Pr}(R=1, C=1)}{\operatorname{Pr}(R=1, C=1)+\operatorname{Pr}(R=0, C=1)}
$$

## Question

- What is the probability of having a rainfall today?

$$
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## Conditional Distribution

- Discrete: joint PMF $\quad p(m, n)=\operatorname{Pr}(M=m, N=n)$

$$
\begin{aligned}
\operatorname{Pr}\left(N=n_{0} \mid M=m\right) & =\frac{\operatorname{Pr}\left(N=n_{0}, M=m\right)}{\sum_{n} \operatorname{Pr}(N=n, M=m)} \\
& =\frac{\operatorname{Pr}\left(N=n_{0}, M=m\right)}{\operatorname{Pr}(M=m)}
\end{aligned}
$$

- Continuous: joint PDF $p(x, y)$


## Conditional Distribution

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\end{aligned}
$$

- Continuous: joint PDF $p(x, y)$

$$
p(y \mid x)=\frac{p(x, y)}{\int p(x, y) d y}=\frac{p(x, y)}{p(x)}
$$

Generalize Functions



Generalize Functions

 continuous

## Discriminative Model



Generative: $\quad p\left(x_{1}, x_{2}, \ldots, x_{m}, y_{1}, y_{2}, \ldots, y_{n}\right)$
Discreminative: $p\left(y_{1}, y_{2}, \ldots, y_{n} \mid x_{1}, x_{2}, \ldots, x_{m}\right)$

## Discriminative Model



Generative: $\quad p(X, Y)$ Discreminative: $p(Y \mid X)$

$$
P(X, Y)=P(Y \mid X) P(X)
$$

$$
P(Y \mid X)=\frac{P(X, Y)}{P(X)}=\frac{P(X, Y)}{\sum_{Y^{\prime}} P\left(X, Y^{\prime}\right)}
$$

