

Probability Theory Refresher (§2)

Will be slightly informal (but in a way that can be made completely rigorous)

Axiomatic approach \rightarrow text book / after class. When in doubt: ASK!

Probability distribution on \mathcal{A} (finite set): $P: \mathcal{A} \rightarrow \mathbb{R}_{\geq 0}, \sum_{a \in \mathcal{A}} P(a) = 1$

e.g. **Bernoulli**(f): $\mathcal{A} = \{0, 1\}, P(1) = f, P(0) = 1 - f$

Uniform(\mathcal{A}): $P(a) = \frac{1}{\#\mathcal{A}} \forall a \in \mathcal{A}$

Random variable (RV) $X \hat{=} \text{prob. dist. } P_X \text{ on set } \mathcal{A}_X$

NOTATION: $X \sim P$ for $P_X = P$

UNLIKE THE
BOOK, I ALWAYS
DISTINGUISH
 X and x

$\Pr(X = x) = P_X(x) \hat{=} P(x)$ we leave out subscript if clear!

$\Pr(X \in S) = \sum_{x \in S} P(x)$

$\Pr(\text{condition on } X) = \sum_{x \text{ cond. holds}} P(x) = \Pr(X \in \{x \text{ s.t. condition holds}\})$

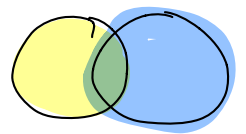
e.g. if X random variable on $\{1, \dots, 6\}$:

$\Pr(X \text{ even and } X \neq 2) = \Pr(X \in \{4, 6\}) = P(4) + P(6)$

* $\Pr(A \text{ or } B) = \Pr(A) + \Pr(B) - \Pr(A \text{ and } B)$

$\hat{=} \Pr(A) + \Pr(B)$
"union bound"

$= 0$ if mutually exclusive



* X RV, f function $\Rightarrow Y = f(X)$ RV

$$\Pr(Y=y) = \sum_{x: f(x)=y} \Pr(X=x)$$

or simply

$$P(y) = \sum_{f(x)=y} P(x)$$

More than one random variable

How to describe "pair of RVs" (X, Y) ? "Joint prob. dist.":

$$\Pr(X=x, Y=y) = P_{(X,Y)}(x,y) = P_{X,Y}(x,y) = P(x,y)$$

i.e. (X, Y) is RV on $\mathcal{A}_{X,Y} = \mathcal{A}_X \times \mathcal{A}_Y$. Similar for tuples.

* Can visualize by "probability table" or "contingency table":

$Y \backslash X$	SUN	RAIN	TOTAL
SUMMER	30%	10%	40%
WINTER	20%	40%	60%
	50%	50%	

* Marginal distributions of X & Y:

$$P(X) = \sum_Y P(x,y) \quad \& \quad P(Y) = \sum_X P(x,y)$$

i.e. $Pr(X=x) = \sum_Y Pr(X=x, Y=y)$ etc.

* X, Y are called independent if $P(x,y) = P(x) \cdot P(y)$

NOT independent!
 $P(SUN, SUMMER) \neq P(SUN) \cdot P(SUMMER)$

How about:

15%	60%
5%	20%

Independent!

Conditional prob. dist. of Y given X:

$$Pr(Y=y | X=x) := \frac{Pr(X=x, Y=y)}{Pr(X=x)}$$

NOTATION: $P_{Y|X=x}(y)$, $P_{Y|X}(y|x)$, $P(y|x)$, ...

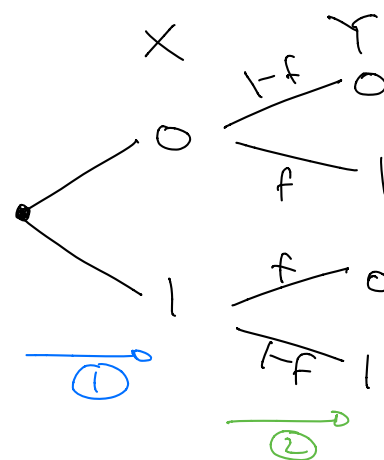
i.e. $P(y|x) = \frac{P(x,y)}{P(x)}$ and $P(x|y) = \frac{P(x,y)}{P(y)}$

* $P(y|x)$ is prob. dist in y for each fixed x!

Two simple rewritings:

$$* P(x,y) = P(x)P(y|x) = P(y)P(x|y)$$

e.g. X channel input, $P(y|x)$ channel
 Y channel output



FORWARD

* Bayes rule:

$$P(x|y) = \frac{P(y|x)P(x)}{P(y)} = \frac{P(y|x)P(x)}{\sum_{x'} P(y|x')P(x')}$$

INVERSE

e.g. $P(pos|sick) = P(neg|healthy) = 90\%$, $P(sick) = 1\%$

$$\Rightarrow P(sick|pos) = \frac{0.9 \times 0.01}{0.9 \times 0.01 + 0.1 \times 0.99} = \frac{1}{12} < 10\% \quad \nabla$$

e.g. decoding the repetition code R_3 : assume $S \sim \text{Uniform}(\{0,1\})$
all independent RV's

$Y_1 = S \oplus N_1, \dots, Y_3 = S \oplus N_3$ where $N_1, N_2, N_3 \sim \text{Bernoulli}(f)$
Sum modulo two (XOR)

Assume we received $y = y_1 y_2 y_3$. How should we estimate s ?

$$P(s|y) = \frac{P(y|s) P(s)}{P(y)} = \frac{1}{2}$$

fixed

$$\rightarrow \frac{P(S=0|y)}{P(S=1|y)} = \frac{P(y|S=0)}{P(y|S=1)} = \frac{P(y|X=000)}{P(y|X=111)} = \prod_{k=1}^3 \frac{P(y_k|X_k=0)}{P(y_k|X_k=1)}$$

$$= \left(\frac{1-f}{f}\right)^{\#0's - \#1's} = \begin{cases} > 1 & \text{if } \#0's > \#1's \\ < 1 & \text{if } \#1's > \#0's \end{cases}$$

= majority vote

$\frac{1-f}{f}$ if $y_k=0$, else $\frac{f}{1-f}$

Combining independent RV's: independent and identical distribution

Quiz: ① Let $X, N \stackrel{iid}{\sim} \text{Uniform}(\{0,1\})$, $Y = X \oplus N$.
 Are X and Y independent? uniform!

YES! $\Pr(X=x, Y=y) = \Pr(X=x, N=x \oplus y) = \frac{1}{4} = \Pr(X=x) \Pr(Y=y)$

② How to label two dice w/ numbers from $0, \dots, 6$ such that their sum is $\sim \text{Uniform}(\{1, 2, \dots, 12\})$?

A: 123456
 B: 000666

Binomial(n, f): Distribution of $Y = X_1 + \dots + X_n$ where $X_i \stackrel{iid}{\sim} \text{Bernoulli}(f)$

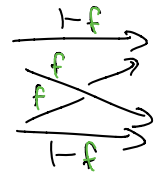
* e.g. number of bit flips when we send n bits through

* $\Pr(Y=k) = \binom{n}{k} f^k (1-f)^{n-k}$

bitstrings with k ones and $n-k$ zeros

probability of any such string

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$



binomial coefficient

"Numerical" random variables

If $X \sim P$ is RV with values in $\mathcal{X} \subseteq \mathbb{R}$:

Expectation value (mean): $EX = E[X] = \sum_{x \in \mathcal{X}} P(x) \cdot x$

* $E[f(x)] = \sum_x P(x) \cdot f(x)$ "law of the unconscious statistician"

* $E[cX] = c \cdot E[X]$ & $E[X+Y] = E[X] + E[Y]$ (A)

* If X, Y independent: $E[XY] = E[X] \cdot E[Y]$ $\sum_{x,y} p(x)p(y)xy$

↳ $X \sim \text{Uniform}(\{-1, 1\}), Y = -X \Rightarrow E[XY] = -1, E[X] = E[Y] = 0$
NST indep

Examples

Variance: $\text{Var}(X) = E[(X - EX)^2]$
 $= \sum_x P(x)(x - EX)^2 = E[X^2] - E[X]^2$

P	Bernoulli(f)	Binomial(n, f)
E	f	$n \cdot f$ (A)
Var	$f(1-f)$	$n \cdot f \cdot (1-f)$ (B)

* $\text{Var}(cX) = c^2 \text{Var}(X)$

* If X, Y independent:

$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$ (B)

↳ we that $E[XY] = E[X] \cdot E[Y]$

$E[(X - EX)^2] = E[(X - f)^2]$
 $= f(1-f)^2 + (1-f)(0-f)^2 = f(1-f)$

Interpretation?

Markov inequality: If $X \geq 0$: $\Pr(X \geq t) \leq \frac{E[X]}{t}$ ($\forall t > 0$)

With high probability

Chebyshev inequality: $\Pr(|X - EX| \geq \epsilon) \leq \frac{\text{Var}(X)}{\epsilon^2}$

↳ WHP deviation from mean is of order $\sqrt{\text{Var}(X)}$

Law of large numbers: $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} P$ w/ mean μ , variance σ^2 ,

$\bar{X} := \frac{1}{n}(X_1 + \dots + X_n)$

$\Rightarrow \Pr(|\bar{X} - \mu| \geq \epsilon) \leq \frac{\sigma^2}{n \cdot \epsilon^2}$

WHP: empirical averages \approx expectation value

We will look into these results more carefully next week

and then discuss "entropy" and compression!