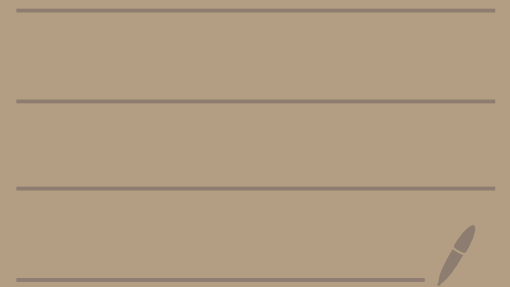


PTS - Chapter 1



1. Introduction to Probability Theory.

- P.T. provides
- concepts to speak about uncertainty (games of chance, recurring events with some pattern or variation, e.g., measurements, balls in a Galton board, etc.)
 - methods to quantify uncertainty

Example: Rolling a die D

What is $P(D=6)$? $\frac{1}{6}$

$P(D \geq 4)$? $\frac{3}{6} = \frac{1}{2}$ $P(D < 4)$? $\frac{1}{2}$

Meaning:

1) In the long run, $1/6$ of the throws results in 6.

2) There is a chance of $1/6$ that this throw results in 6.

frequentist

subjective (Bayesian)

view

No consequences for
mathematical theory

1.1 Events

Experiments:

\mathcal{S} sample space (= set of possible outcomes)

\mathcal{S} is known, but not a specific outcome

Rolling a die: $\mathcal{S} = \{1, 2, 3, \dots, 6\}$

$$\#\mathcal{S} = 6$$

Rolling two dice: $\mathcal{S} = \{(1, 1), (1, 2), \dots, (1, 6),$
 $\dots, (3, 4), \dots,$
 $(6, 1), \dots, (6, 6)\}$

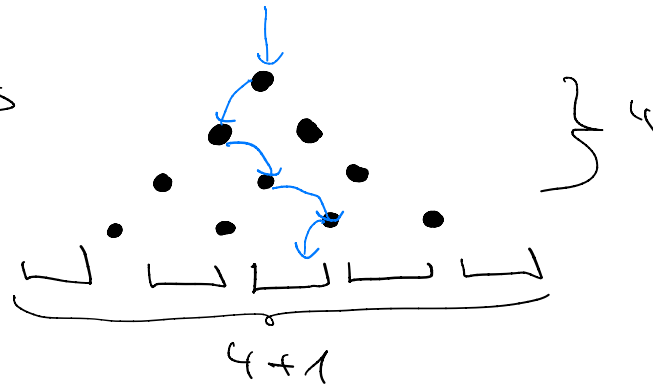
$$\begin{aligned} \#\mathcal{S} &= 6 \cdot 6 \\ &= 36 \end{aligned}$$

Galton board:

with u levels
has outcomes

$$\{1, \dots, u+1\}$$

$$\{0, \dots, u\}$$



\mathcal{I} can also be infinite!

- measuring distances: any number $m > 0$ is possible outcome
— : — $m \in [0.5, 3.00]$

infinitely many
reals in the
interval

- How many times do we have to throw
a die until a 6 appears
 \Rightarrow consider arbitrarily long sequences of throws,
even infinite one

Events:

throw an even number $\mathcal{E}_{\text{even}} = \{2, 4, 6\}$

throw two equal numbers $\mathcal{E}_{\text{equal}} = \{(1,1), (2,2), \dots, (6,6)\}$

In general: $\mathcal{E} \subseteq \mathcal{S}$ ~~is~~ events are sets

\mathcal{E} occurs if outcome $\in \mathcal{E}$

\uparrow ~~is~~ the outcome of our experiment
in an experiment

Set operations on events

$\mathcal{E} \cup \mathcal{F}$ union ("or")

$\mathcal{E} \cap \mathcal{F}$, $\mathcal{E} \bar{\mathcal{F}}$ intersection ("and")

$\bar{\mathcal{E}}$ complement ("not"), $\mathcal{S} \setminus \mathcal{E} = \bar{\mathcal{E}}$

$\mathcal{E} \setminus \mathcal{F}$ \mathcal{E} and not \mathcal{F}

$= \mathcal{E} \cap \bar{\mathcal{F}}$

Example: $E_{\text{prime}} := \{2, 3, 5\}$

$$E_{\text{even}} \cup E_{\text{prime}} = \{2, 3, 4, 5, 6\} = \overline{\{1\}}$$

$$E_{\text{even}} \cap E_{\text{prime}} = \{2\}$$

$$\overline{E_{\text{prime}}} = \{1, 4, 6\}$$

Special case $\emptyset \in \mathcal{S}$, impossible event

Terminology

$$E \cap F = \emptyset \quad \text{disjoint}$$

$$\overline{\emptyset} = \mathcal{S}, \quad \overline{\mathcal{S}} = \emptyset$$

$$\overline{\overline{E}} = E$$

$$\overline{E \cup F} = \overline{E} \cap \overline{F}, \quad \overline{E \cap F} = \overline{E} \cup \overline{F}$$

$$E \equiv F \Leftrightarrow E \subseteq F \text{ and } F \subseteq E$$

E, F are equivalent

Pronunciation



(De Morgan's Rule)

1.2 Axioms of Probability

$P(E)$ probability of E , is a real number

A1: $0 \leq P(E) \leq 1$

A2: $P(S) = 1$

A3: If E_1, E_2, \dots are *pairwise* mutually disjoint (i.e., $E_i \cap E_j = \emptyset$ for $i \neq j$) *any two are disjoint*
Then $P(E_1 \cup \dots \cup E_n) = P(E_1) + \dots + P(E_n) = \sum_{i=1}^n P(E_i)$, finite

(also holds for infinite sums)

What is $P(\emptyset) = 0$ because

S, \emptyset are disjoint

$$\underline{P(S)} = P(S \cup \emptyset) = \underline{P(S)} + \underline{P(\emptyset)}$$

$$\Rightarrow 0 = P(\emptyset)$$

$$P(\text{Even} \cup \{1\}) = P(\text{Even}) + P(\{1\})$$

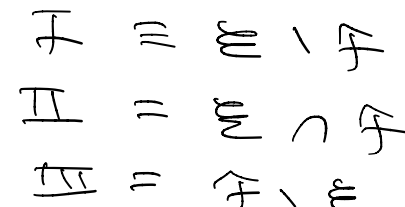
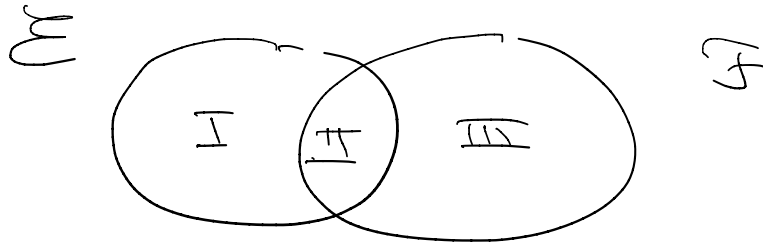
Proposition 1: $P(\bar{E}) = 1 - P(E)$

Proof: $E \cap \bar{E} = \emptyset, E \cup \bar{E} = \mathcal{S}$

$1 = P(\mathcal{S}) = P(E \cup \bar{E}) = P(E) + P(\bar{E})$
Ax2 Ax3

$\Rightarrow 1 - P(E) = P(\bar{E})$

What ^{about} $P(E \cup F)$ in general?



$P(E \cup F) = P(I \cup II \cup III)$
 $= P(I) \cup P(II) \cup P(III)$ (Ax3)

$P(E) = P(I) + P(II)$

$P(F) = P(II) + P(III)$

$\Rightarrow P(E \cup F) = P(E) + P(F) - P(E \cap F)$

Proposition 2 : $P(E \cup F) = P(E) + P(F) - P(E \cap F)$

Example 3 :

Drinks Quiz

B drinks beer, W drinks wine

$$\begin{aligned} P(B \cup W) &= P(B) + P(W) - P(B \cap W) \\ &= .68 + .49 - .35 = 1.17 - .35 = .82 \end{aligned}$$

$$P(\overline{B \cup W}) = 1 - .82 = .18$$

Definition 4: The odds of \mathcal{E} is

$$\frac{P(\mathcal{E})}{P(\overline{\mathcal{E}})} = \frac{P(\mathcal{E})}{1 - P(\mathcal{E})}$$

Says how much more likely \mathcal{E} is than $\overline{\mathcal{E}}$

Quiz: Odds of throwing a 4

$$\mathcal{E} = \{4\}, \quad P(\mathcal{E}) = \frac{1}{6}$$

$$\text{odds}(\mathcal{E}) = \frac{P(\mathcal{E})}{1 - P(\mathcal{E})} = \frac{\frac{1}{6}}{\frac{5}{6}} = \frac{1}{5}$$

Toump example: $P(\underbrace{\text{"T will be el."}}_{\mathcal{E}}) = .4$

$$\text{odds}(\mathcal{E}) = \frac{.4}{.6} = \frac{2}{3}$$

1.3 Uniformity

Often: all outcomes are equally likely (only possible if \mathcal{S} finite)
and have prob. > 0

$\# \mathcal{S} = u$, say $\mathcal{S} = \{1, \dots, u\}$

$$\Rightarrow P(\{1\}) = P(\{2\}) = \dots = P(\{u\}) = p$$

$$\Rightarrow 1 = P(\mathcal{S}) = \underbrace{P(\{1, \dots, u\})}_{A2} = \underbrace{P(\{1\}) + \dots + P(\{u\})}_{A3} = u \cdot p$$

$$\Rightarrow 1 = u \cdot p \Rightarrow p = \frac{1}{u} = P(\{i\}), \quad 1 \leq i \leq u$$

Generalize: $\boxed{\Sigma \subseteq \mathcal{S}} \Rightarrow P(\Sigma) = \boxed{\frac{\#\Sigma}{u}}$ $\# \Sigma$ is the
card. of Σ

Counting Principle

outcomes : throw 2 or 3 dice

first throw a die, determines a stack of cards
then pick a card (of 32)

Combinations of experiments, sequential executions

E_1 has u outcomes, E_2 has u outcomes

outcomes of " E_1 then E_2 " = $u \cdot u$

Outcomes correspond to matrix

$(1, 1), \dots, (1, u)$

$(2, 1)$

\vdots

$(u, 1) \dots (u, u)$

In general : E_i has u_i outcomes

$\Rightarrow E_1$ then $E_2 \dots$ then E_r has $u_1 \cdot \dots \cdot u_r$ outcomes

Quiz 3: Socks in a Box

$$P(\underbrace{2 \text{ socks have diff. colour}}_{\mathcal{E}}) = \frac{\# \mathcal{E}}{\# \mathcal{S}}$$

8 black, 7 white

\mathcal{S} = Pairs of socks picked, first, then second $\Rightarrow \# \mathcal{S} = 15 \cdot 14$

\mathcal{E} = Pairs with (1st = w, 2nd = b) \cup (1st = b, 2nd = w)

$$\# \mathcal{E} = 7 \times 8 + 8 \times 7$$

$$\frac{\# \mathcal{E}}{\# \mathcal{S}} = \frac{7 \cdot 8 + 8 \cdot 7}{15 \cdot 14} = \frac{\cancel{2} \cdot 7 \cdot 8}{15 \cdot \cancel{2} \cdot 7} = \frac{8}{15}$$

Quiz 4: How many words?

Words = # possibilities 1st bit

X # — a — 2nd — a —

...

X # — a — 32nd bit

2
x
2
x
i
x
2

$$= 2^{32} \quad || \quad 2^{10} = 1024 \approx 1000 = 10^3$$

$$= 2^{30} \cdot 2^2 = (2^{10})^3 \cdot 2^2$$

$$\approx (10^3)^3 \cdot 2^2 = 4,000,000,000$$

Quiz 5: People on an Elevator

$$P(\underbrace{\text{"all off at same floor"}}_E)$$

$$E_i = \text{"all off at floor } i \text{"} \quad (i = 1, \dots, 4)$$

$$\mathcal{S} = \{ (f_1, f_2, f_3, f_4) \mid f_i \in \{1, 2, 3, 4\}, i = 1, \dots, 4 \}$$

$$= \{ (1, 1, 1, 1), (1, 1, 1, 2), \dots \}$$

$$\# \mathcal{S} = 4^4 \quad // \text{ in general: } f = \# \text{ floors, } p = \# \text{ persons}$$

$$E_i = \{ (i, i, i, i) \} \Rightarrow \# E_i = 1$$

$$\mathcal{E} = \sum_{i=1}^f E_i \Rightarrow \# \mathcal{E} = f$$

$$\frac{\# \mathcal{E}}{\# \mathcal{S}} = \frac{f^1 \cdot 1}{f^p \cdot p^{-1}} = \frac{1}{f^{p-1}} \quad \text{Here: } \frac{\# \mathcal{E}}{\# \mathcal{S}} = \frac{1}{4^3} = \frac{1}{2^6}$$

disjoint union,
i.e., the E_i are mut. disjoint

$$= \frac{1}{64}$$

Example 6 10 books: 4 CS, 3 Math, 2 stat, 1 Hist

Organize so that books of same subject are together:

E.g., SS. CCC. H. MMM

How many possibilities?

1) # permutations of subjects $4 \cdot 3 \cdot 2 \cdot 1 = 4!$

2) permutation within subjects, e.g.

CS	:	4!
M	:	3!
S	:	2!
H	:	1!

Arrangements = $4! \cdot 4! \cdot 3! \cdot 2! \cdot 1!$

$$= (2^3 \cdot 3) \cdot (2^3 \cdot 3) \cdot (2^1 \cdot 3) \cdot 2^1$$

$$= 2^8 \cdot 3^3 = 256 \cdot 27$$

$$\approx 250 \cdot 28$$

$$= 250 \cdot 4 \cdot 7 = 7 \cdot 1000 = 7000$$

Arrangement, w/ books mixed up arbitrarily

$$10 \cdot 9 \cdot 8 \cdots 1 = 10!$$

Random array of books
easy to subjects
together
 $\frac{7000}{10!}$

Example 7 Course with 5 male, 3 female students,

We had an exam: all students got different marks

$$P(\text{"all female students got the top marks"}) =$$

S = all possible rankings, $\#S = 8!$ \mathcal{E}

Females Top Rankings:

F	M
3!	5!

$$P(\mathcal{E}) = \frac{\#\mathcal{E}}{\#S} = \frac{3! \cdot 5!}{8!} = \frac{3! \cdot \cancel{5!}}{8 \cdot 7 \cdot 6 \cdot \cancel{5!}} = \frac{\cancel{3!}}{8 \cdot 7 \cdot 6}$$

$$= \frac{1}{8 \cdot 7} = \frac{1}{56}$$

Ex 2 Revised: The top 3 students are a randomly chosen set of 3 out of 8.

We asked: What is the probability that a specific set is chosen?

How many choices of 3 out of 8 are possible?

1.) Choose 3 out of 8 in sequence: How many possibilities?

$$8 \cdot 7 \cdot 6$$

2.) A set of 3 can be obtained in $3!$ ways

a, b, c b, c, a
a, c, b c, a, b
b, a, c c, b, a

$$\begin{aligned} \text{In total: } \frac{8 \cdot 7 \cdot 6}{3 \cdot 2 \cdot 1} &= \frac{(8 \cdot 7 \cdot 6) \cdot (5 \cdot 4 \cdot \dots \cdot 1)}{(3 \cdot 2 \cdot 1) \cdot (5 \cdot 4 \cdot \dots \cdot 1)} \\ &= \frac{8!}{3! 5!} = \binom{8}{3} = \binom{8}{5} \end{aligned}$$

Generalize: Choose r out of n

$$\frac{n!}{r! (n-r)!}$$

$$= \binom{n}{r}$$

" n choose r "

"binomial coefficients"

1.) Select r elements out of n :

$$n (n-1) \dots (n-r+1)$$

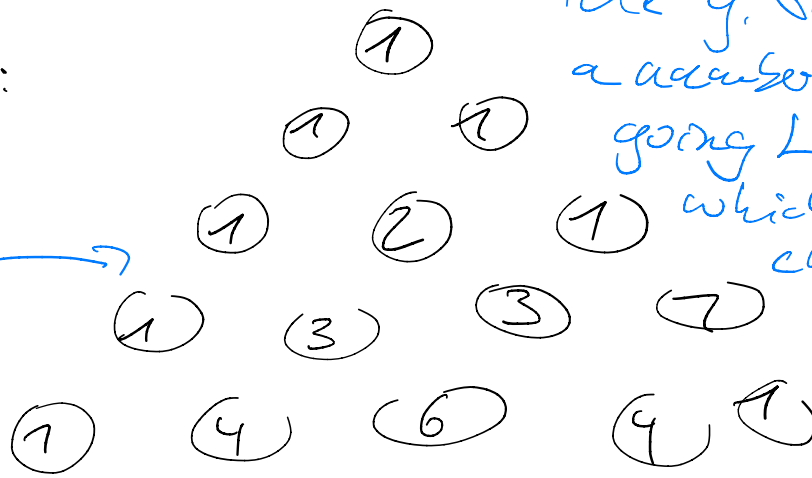
$$= \binom{n}{r}$$

2.) We get the same set $r!$ times

$$(x+y)^n = \binom{n}{0} x^n y^0 + \binom{n}{1} x^{n-1} y^1 + \binom{n}{2} x^{n-2} y^2 + \dots$$

Pascal's triangle:

only one path leading to a pin on the sides



We reach a node in the G board by making a series of choices of going L or R, which correspond to choices of x and y

each node has the # of paths leading to it in the G board

Example 9: Given 5 men, 8 women. Randomly select 5 persons.

$$P(\text{"2 men, 3 women are selected"}) =$$

S = all possible choices of 5 out of 13

$$\# S = \binom{13}{5}$$

E = selection 2 men out of 5,
3 women out of 8

$$\# E = \binom{5}{2} \cdot \binom{8}{3}$$

$$P(E) = \frac{\binom{5}{2} \cdot \binom{8}{3}}{\binom{13}{5}} = \frac{560}{1287}$$

on handw.
notes

Example 10: Given objects $1, \dots, u$. Select subset of size k .

$$P(\underbrace{1 \text{ is in the selection}}_{\Sigma}) =$$

$\mathcal{S} =$ all subsets of size k of $\{1, \dots, u\}$

$$\#\mathcal{S} = \binom{u}{k}$$

$\Sigma =$ all subsets of size k containing 1

$$\#\Sigma = \binom{u-1}{k-1}$$

i.e., only choose remaining $k-1$ out of $u-1$

$$P(\Sigma) = \frac{\binom{u-1}{k-1}}{\binom{u}{k}} = \frac{\cancel{(u-1)!}}{\cancel{(k-1)!} \cdot \underbrace{\cancel{(u-1-(k-1))!}}_{(u-k)!}} = \frac{k \cdot \cancel{(u-k)!}}{n!} = \frac{k}{n}$$

1.4 Conditional Probabilities

Throw two dice

$$P(D_1 + D_2 = 8) = \frac{\#\{(2,6), (3,5), (4,4), (5,3), (6,2)\}}{\#\mathcal{S}}$$
$$= \frac{5}{36}$$

Suppose, we know $D_1 = 3$. (What if $D_1 = 12$)
 $\Rightarrow P(\cdot) = 0$

$$P(D_1 + D_2 = 8 \mid D_1 = 3) = ?$$

condition

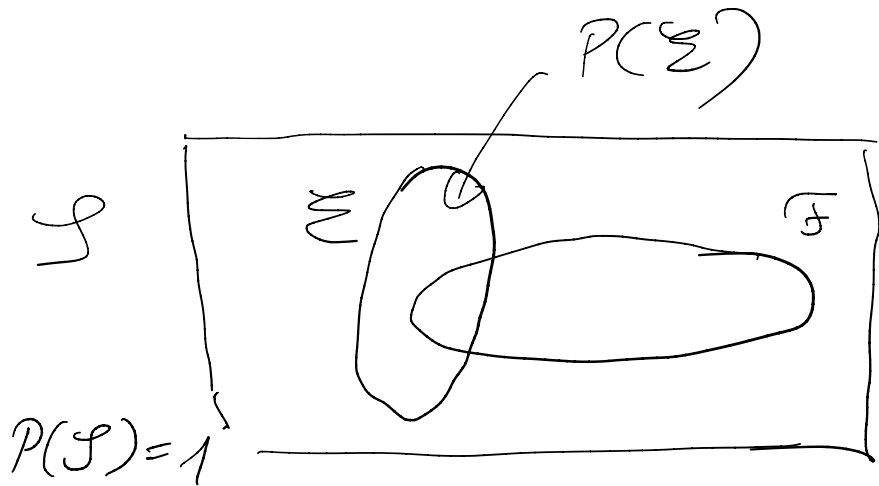
New sample space $\mathcal{S}' = \{(3,1), (3,2), \dots, (3,6)\}$
 $\#\mathcal{S}' = 6$

$$\mathcal{E}' = \{(3,5)\} \Rightarrow \#\mathcal{E}' = 1$$

$$P(D_1 + D_2 = 8 \mid D_1 = 3) = \frac{\#\mathcal{E}'}{\#\mathcal{S}'} = \frac{1}{6}$$

Definition 11: Events Σ, \mathcal{F} , $P(\mathcal{F}) > 0$

$$P(\Sigma | \mathcal{F}) = \frac{P(\Sigma \cap \mathcal{F})}{P(\mathcal{F})}$$



Note: $P(\cdot | \mathcal{F})$ is again a probability on S

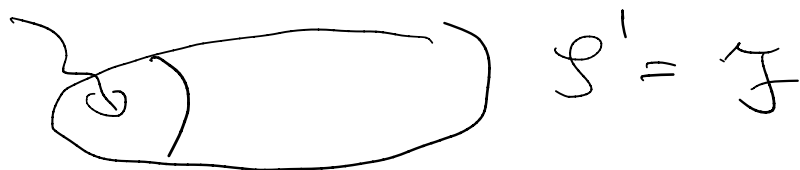
E.g.: $P(E_1 + \dots + E_k | \mathcal{F})$

$= P(E_1 | \mathcal{F}) + \dots + P(E_k | \mathcal{F})$ if

\mathcal{F}_i are mutually disjoint

Now, only outcomes in \mathcal{F} are considered possible

$$\Sigma' = \Sigma \cap \mathcal{F}$$



Normalize P to $P' = P(\cdot | \mathcal{F})$, such that $P'(S') = 1$

$$P'(\Sigma) = P(\Sigma | \mathcal{F}) = \frac{P(\Sigma \cap \mathcal{F})}{P(\mathcal{F})}$$

Subjective (= Bayesian) view of probabilities

We have seen the subj. view of cond. prob.

Frequentist view: n experiments

$\mathcal{F} \approx n \cdot P(\mathcal{F})$ many times

$\mathcal{E}\mathcal{F} \approx n \cdot P(\mathcal{E}\mathcal{F})$ many times

Ignore outcomes not \mathcal{F} .

Among the $n \cdot P(\mathcal{F})$ many \mathcal{F} -outcomes, there are $n \cdot P(\mathcal{E}\mathcal{F})$ many

$\mathcal{E}\mathcal{F}$ -outcomes

$$\Rightarrow P(\mathcal{E} | \mathcal{F}) = \frac{n \cdot P(\mathcal{E}\mathcal{F})}{n \cdot P(\mathcal{F})} = \frac{P(\mathcal{E}\mathcal{F})}{P(\mathcal{F})}$$

Example 11 Box with 32 transistors:

20 working, 8 partly working, 4 deficient

Exp: Choose 1 transistor.
Suppose it does not fail. What is the prob. that it is working?

Three events: W, P, D (pick a working, ... transistor)

$$P(W|\bar{D}) = \frac{P(W\bar{D})}{P(\bar{D})} = \frac{P(W)}{P(W \cup P)} = \frac{\frac{20}{32}}{\frac{28}{32}} = \frac{20}{28} = \frac{5}{7}$$

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$

Quiz 7: Tossing Coins

$$P(2 \text{ heads} \mid \geq 1 \text{ head}) = \frac{P(\mathcal{E} \cap \mathcal{F})}{P(\mathcal{F})} = \frac{P(2 \text{ heads} \wedge \geq 1 \text{ head})}{P(\geq 1 \text{ head})}$$

$$= \frac{P(\{(H, H)\})}{P(\{(H, H), (T, H), (H, T)\})} = \frac{\frac{1}{4}}{\frac{3}{4}} = \frac{1}{3}$$

$$\mathcal{S} = \left\{ \begin{array}{l} (H, H) \\ (H, T) \\ (T, H) \\ (T, T) \end{array} \right\}$$

\mathcal{E} (circled in red)
 \mathcal{F} (circled in blue)

Quiz 8: Champions

reaching the final



winning the final

$$P(\text{Champions}) = P(FW) = P(W|F)P(F) \\ = .5 \times .2 = \underline{\underline{0.1}}$$

$$P(F) = .2$$

$$P(W|F) = .5$$

$$P(W|F) = \frac{P(WF)}{P(F)}$$

$$\Rightarrow P(W|F)P(F) = P(WF)$$

1.5 Bayes' Formula

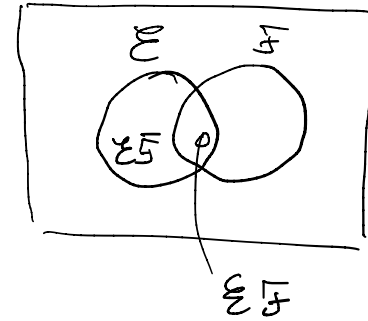
\mathcal{E}, \mathcal{F} events $\Rightarrow \mathcal{E} = \mathcal{E}\mathcal{F} \cup \mathcal{E}\bar{\mathcal{F}}$

$$\Rightarrow P(\mathcal{E}) = P(\mathcal{E}\mathcal{F}) + P(\mathcal{E}\bar{\mathcal{F}})$$

$$= P(\mathcal{E}|\mathcal{F})P(\mathcal{F}) + P(\mathcal{E}|\bar{\mathcal{F}})P(\bar{\mathcal{F}})$$

Here: $P(\mathcal{E})$ can be computed by conditioning on some \mathcal{F}

special case of the law of total probability (LOTP)



Example 15 Insurance company: People are risk takers (30%) or not. Every year, 40% of risk takers have an accident, only 20% of non-risk takers. What is $P(A)$?

$$P(A) = P(A|R)P(R) + P(A|\bar{R})P(\bar{R})$$

$$= .4 \times .3 + .2 \times .7 = .12 + .14 = 0.26$$

Updating beliefs in the presence of new information

Example 16: Suppose, a client has an accident.

What is the prob. this was a risk taker?

$$P(R|A) = \frac{P(RA)}{P(A)} = \frac{P(A|R)P(R)}{P(A)}$$

Bayes' Law

Remember:

$$P(A|R) = \frac{P(AR)}{P(R)} = \frac{P(RA)}{P(R)}$$

$$\Rightarrow P(RA) = P(A|R)P(R)$$

$$\begin{aligned} &= \frac{.4 \times .3}{0.26} \\ &= \frac{.12}{.26} = \frac{12}{26} = \frac{6}{13} \end{aligned}$$

$$= 0.46$$

Bayes' Formula

$$P(\bar{F} | E) = \frac{P(E | \bar{F}) P(\bar{F})}{P(E)}$$

Quiz 9: Testing for a Disease

D person has disease

J test is positive

$$P(J | D) = \frac{99}{100}$$

$$P(J | \bar{D}) = \frac{1}{100}$$

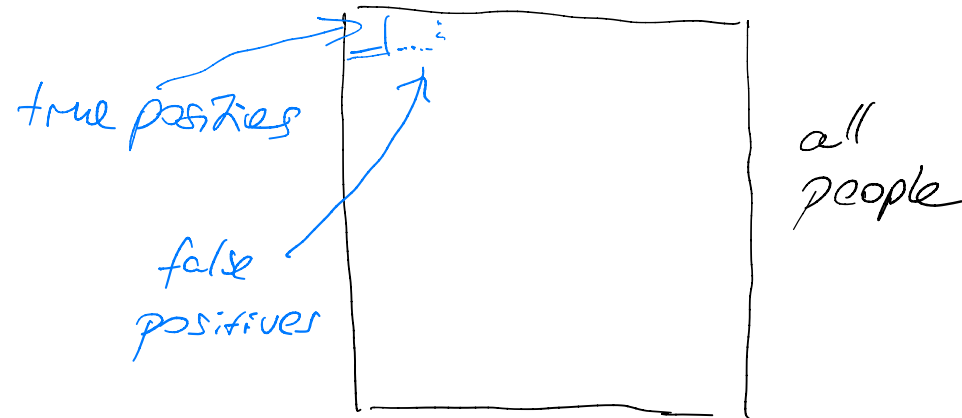
$$P(D) = \frac{1}{100}$$

$$P(D | J) = \frac{P(J | D) P(D)}{P(J)} = \frac{\frac{99 \cdot 1}{100 \cdot 100}}{\frac{2 \cdot 99}{100 \cdot 100}} = \frac{1}{2}$$

$$P(J) = P(J | D) P(D) + P(J | \bar{D}) P(\bar{D})$$

$$= \frac{99}{100} \cdot \frac{1}{100} + \frac{1}{100} \cdot \frac{99}{100}$$

LOTP



General Law of Total Probability (LOTP)

Let $\mathcal{F}_1, \dots, \mathcal{F}_n$ be a partition of \mathcal{S} , i.e.

$$\bullet \mathcal{F}_i \cap \mathcal{F}_j = \emptyset, \quad i \neq j$$

$$\bullet \bigcup_{i=1}^n \mathcal{F}_i = \mathcal{F}_1 \cup \dots \cup \mathcal{F}_n = \mathcal{S}$$

Note:

$$\sum_{i=1}^n P(\mathcal{F}_i) = 1$$

$$\mathcal{E} \subseteq \mathcal{S} \text{ event} \Rightarrow \mathcal{E} = \bigcup_{i=1}^n \mathcal{E} \mathcal{F}_i$$

$$\Rightarrow P(\mathcal{E}) = \sum_{i=1}^n P(\mathcal{E} \mathcal{F}_i) = \sum_{i=1}^n P(\mathcal{E} | \mathcal{F}_i) P(\mathcal{F}_i)$$

Quiz 11: Marbles

LOTP: What are E and the F_i ?

E = Marble drawn is red

F_i = The i -th bag is chosen

$$P(E) = P(E|F_1)P(F_1) + P(E|F_2)P(F_2) + P(E|F_3)P(F_3)$$

$$P(F_1) = P(F_2) = P(F_3) = \frac{1}{3}$$

$$P(E|F_1) = \frac{80}{100} \quad P(E|F_2) = \frac{55}{100} \quad P(E|F_3) = \frac{45}{100}$$

$$\Rightarrow P(E) = \frac{1}{3} \frac{80 + 55 + 45}{100} = \frac{1}{3} \frac{180}{100} = \frac{60}{100}$$

$(F_i)_{i=1}^n$ partition of S , E event

$$P(F_i | E) = \dots$$

$$\begin{aligned} P(F_i | E) &= \frac{P(F_i \cap E)}{P(E)} = \frac{P(E \cap F_i)}{P(E)} \\ &= \frac{P(E | F_i) P(F_i)}{P(E)} \\ &= \frac{P(E | F_i) P(F_i)}{\sum_{i=1}^n P(E | F_i) P(F_i)} \end{aligned}$$

Generalized Bayes' Formula

Quiz 10: Covid-20 Diagnosis

B person has Covid

T test is positive

Interesting: $P(B|T) = ?$

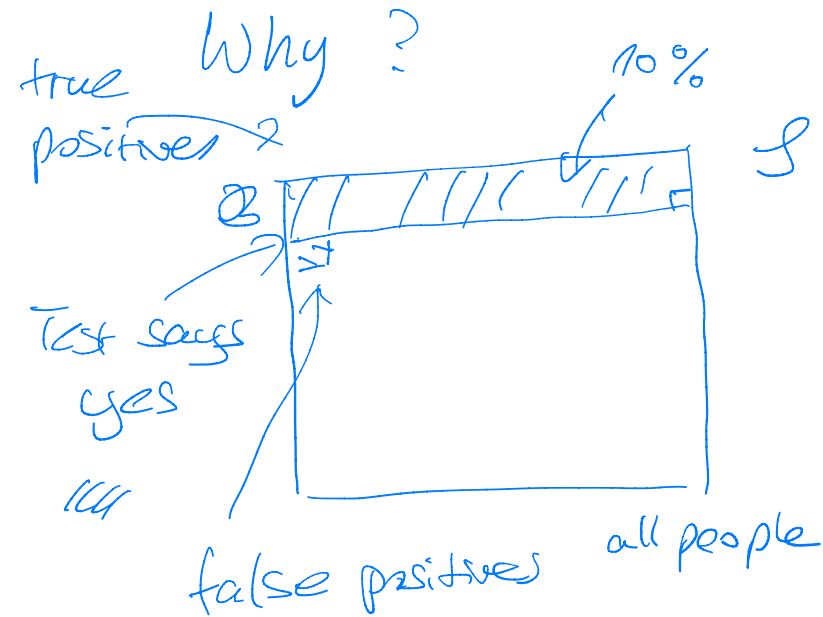
$$P(T|B) = \frac{99}{100}$$

$$P(\bar{T}|\bar{B}) = \frac{99}{100} \Rightarrow P(T|\bar{B}) = \frac{1}{100}$$

$$P(B) = \frac{10}{100}$$

$$P(B|T) = \frac{P(T|B)P(B)}{P(T)} = \frac{\frac{99}{100} \cdot \frac{10}{100}}{\frac{99}{100} \cdot \frac{10}{100} + \frac{1}{100} \cdot \frac{90}{100}}$$

$$P(T) = P(T|B)P(B) + P(T|\bar{B})P(\bar{B}) = \frac{99}{100} \cdot \frac{10}{100} + \frac{1}{100} \cdot \frac{90}{100} = \frac{99}{1000} + \frac{9}{1000} = \frac{108}{1000} = 10.8\% \approx 11\%$$



1.6 Independent Events

Example: Consider a deck of French cards

\mathcal{E} = draw a red card

$$P(\mathcal{F}) = \frac{4}{52} = \frac{1}{13} \quad \begin{array}{l} \text{red} \\ \mathcal{E} \\ \text{aces} \end{array}$$

\mathcal{F} = draw an ace

$$P(\mathcal{F}|\mathcal{E}) = \frac{2}{26} \quad \begin{array}{l} \rightarrow \\ \text{\# red cards} \end{array}$$

$$P(\mathcal{E}) = \frac{1}{2}, \quad P(\mathcal{E}|\mathcal{F}) = \frac{1}{2}$$

Intuition: Knowing \mathcal{F} doesn't tell us anything about \mathcal{E}

\mathcal{E}, \mathcal{F} are independent

In general, $P(\mathcal{E}|\mathcal{F}) \neq P(\mathcal{E})$

\rightarrow posterior prob. of \mathcal{E}

\leftarrow prior probability of \mathcal{E}

Definition: \mathcal{E}, \mathcal{F} independent

$$P(\mathcal{E}|\mathcal{F}) = P(\mathcal{E})$$

Note: the definition is symmetric



$$P(\Sigma | \mathcal{F}) = P(\Sigma) \Leftrightarrow P(\Sigma) = \frac{P(\Sigma \mathcal{F})}{P(\mathcal{F})}$$

$$\Leftrightarrow \boxed{P(\Sigma)P(\mathcal{F}) = P(\Sigma \mathcal{F})}$$

if $P(\mathcal{F}) \neq 0$,
 $P(\Sigma) \neq 0$

The first definition assumes that $P(\mathcal{F}) \neq 0$

Alternative:

$$\Sigma, \mathcal{F} \text{ indep. iff } P(\Sigma \mathcal{F}) = P(\Sigma) \cdot P(\mathcal{F})$$

$$\begin{aligned} P(\Sigma)P(\mathcal{F}) = P(\Sigma \mathcal{F}) &\Rightarrow P(\mathcal{F}) = \frac{P(\Sigma \mathcal{F})}{P(\Sigma)} \\ &= \frac{P(\mathcal{F} \Sigma)}{P(\Sigma)} = P(\mathcal{F} | \Sigma) \end{aligned}$$

Quiz: Dice and independence

$$\Sigma = \{D_1 + D_2 = 7\}$$

$$\mathcal{F} = \{D_1 + D_2 = 8\}$$

$$\mathcal{G} = \{D_1 = 5\}$$

$$\mathcal{F} = \{(2,6), (3,5), (4,4), (5,3), (6,2)\}$$

$$\Rightarrow \#\mathcal{F} = 5$$

$$\mathcal{G} = \{(3,1), \dots, (3,6)\}$$

$$\Rightarrow \#\mathcal{G} = 6$$

$$\Sigma \text{ ind } \mathcal{F} : P(\Sigma \cap \mathcal{F}) = 0 \neq \frac{1}{6} \cdot \frac{5}{36} = P(\Sigma)P(\mathcal{F})$$

No!

$$\Sigma \text{ ind } \mathcal{G} : P(\Sigma \cap \mathcal{G}) = \frac{1}{36} = \frac{1}{6} \cdot \frac{1}{6} = P(\Sigma)P(\mathcal{G})$$

Yes!

$$\mathcal{F} \text{ ind } \mathcal{G} : P(\mathcal{F} \cap \mathcal{G}) = \frac{1}{36} \neq \frac{5}{36} \cdot \frac{1}{6} = P(\mathcal{F})P(\mathcal{G})$$

No!

$$P(\Sigma) = \frac{6}{36} = \frac{1}{6}$$

$$P(\Sigma \cap \mathcal{F}) = 0$$

$$P(\mathcal{F}) = \frac{5}{36}$$

$$P(\Sigma \cap \mathcal{G}) = \frac{1}{36}$$

$$P(\mathcal{G}) = \frac{6}{36} = \frac{1}{6}$$

$$P(\mathcal{F} \cap \mathcal{G}) = \frac{1}{36}$$

Example: Consider a deck of French cards

Σ = draw a red card

\mathcal{F} = draw an ace

$$P(\Sigma) = \frac{1}{2}, \quad P(\Sigma | \mathcal{F}) = \frac{1}{2}$$

$$P(\mathcal{F}) = \frac{1}{13}, \quad P(\mathcal{F} | \Sigma) = \frac{1}{13}$$

Symmetry:

$$P(\Sigma \mathcal{F}) = P(\Sigma) P(\mathcal{F})$$

$$\Leftrightarrow P(\Sigma | \mathcal{F}) = P(\Sigma)$$

$$\Leftrightarrow P(\mathcal{F} | \Sigma) = P(\mathcal{F})$$

if $P(\Sigma) \neq 0 \neq P(\mathcal{F})$

We observe as well:

$$P(\bar{\Sigma} | \mathcal{F}) = \frac{1}{2}$$

$$P(\bar{\mathcal{F}} | \Sigma) = \frac{12}{13}$$

$$P(\bar{\Sigma}) = \frac{1}{2}$$

$$P(\bar{\mathcal{F}}) = \frac{12}{13}$$

Definition of independence of events

Proposition 20 Σ, \mathcal{F} independent $\Rightarrow \Sigma, \bar{\mathcal{F}}$ independent

Proof: Show $P(\Sigma \bar{\mathcal{F}}) = P(\Sigma)P(\bar{\mathcal{F}})$

$$\begin{aligned} P(\Sigma) &= P(\Sigma \mathcal{F} \cup \Sigma \bar{\mathcal{F}}) = P(\Sigma \mathcal{F}) + P(\Sigma \bar{\mathcal{F}}) \\ &= P(\Sigma) \cdot P(\mathcal{F}) + P(\Sigma \bar{\mathcal{F}}) \end{aligned}$$

$$\Rightarrow P(\Sigma) - P(\Sigma) \cdot P(\mathcal{F}) = \underline{P(\Sigma \bar{\mathcal{F}})}$$

$$(1 - P(\mathcal{F}))P(\Sigma) = \underline{P(\bar{\mathcal{F}}) \cdot P(\Sigma)}$$

Σ, \mathcal{F} ind. $\Rightarrow \Sigma, \bar{\mathcal{F}}$ ind.

Independence is inherited by complements

Consider \mathcal{E} ind of \mathcal{F} , $\not\Rightarrow$ \mathcal{E} ind $\mathcal{F} \cap \mathcal{G}$
 \mathcal{E} ind of \mathcal{G}

Example: Throw two dice

$\mathcal{E} = "D_1 + D_2 = 7"$ $\mathcal{F} = "D_1 = 1"$ $\mathcal{G} = "D_2 = 6"$

$$P(\mathcal{E}) = \frac{1}{6} \quad P(\mathcal{F}) = \frac{1}{6} \quad P(\mathcal{G}) = \frac{1}{6}$$

$$P(\mathcal{E} \cap \mathcal{F}) = \frac{1}{36} \quad P(\mathcal{E} \cap \mathcal{G}) = \frac{1}{36} \quad P(\mathcal{F} \cap \mathcal{G}) = \frac{1}{36}$$

$\Rightarrow \mathcal{E}, \mathcal{F}$ ind. \mathcal{E}, \mathcal{G} ind

What about ind. of \mathcal{E} and $\mathcal{F} \cap \mathcal{G} = \mathcal{F} \cap \mathcal{G}$?

$$P(\mathcal{E})P(\mathcal{F} \cap \mathcal{G}) = \frac{1}{6} \frac{1}{36} \neq \frac{1}{36} = P(\mathcal{E} \cap (\mathcal{F} \cap \mathcal{G})) = P(\mathcal{E} \cap (\mathcal{F} \cap \mathcal{G}))$$

$\Rightarrow \mathcal{E}$ and $\mathcal{F} \cap \mathcal{G}$ are not ind.

Let \mathcal{E} ind of \mathcal{F} , \mathcal{E} ind of \mathcal{G} . Can we conclude that
 \mathcal{E} ind of $\mathcal{F}\mathcal{G}$ ($= \mathcal{F} \cap \mathcal{G}$)?

No!

Example: Throw two dice!

$$\mathcal{E} = "D_1 + D_2 = 7" \quad \mathcal{F} = "D_1 = 1" \quad \mathcal{G} = "D_2 = 6"$$

$$P(\mathcal{E}) = \frac{1}{6} \quad P(\mathcal{F}) = \frac{1}{6} \quad P(\mathcal{G}) = \frac{1}{6}$$

$$P(\mathcal{E}\mathcal{F}) = \frac{1}{36} \quad P(\mathcal{E}\mathcal{G}) = \frac{1}{36} \quad P(\mathcal{F}\mathcal{G}) = \frac{1}{36}$$

$\Rightarrow \mathcal{E}, \mathcal{F}$ ind.

$\Rightarrow \mathcal{E}, \mathcal{G}$ ind.

What about the independence of \mathcal{E} and $\mathcal{F}\mathcal{G}$?

Compare $P(\mathcal{E}(\mathcal{F}\mathcal{G}))$ and $P(\mathcal{E}) \cdot P(\mathcal{F}\mathcal{G})$!

$$P(\mathcal{E}(\mathcal{F}\mathcal{G})) = P(\mathcal{E}\mathcal{F}\mathcal{G}) = \frac{1}{36}$$

$$P(\mathcal{E}) \cdot P(\mathcal{F}\mathcal{G}) = \frac{1}{6} \cdot \frac{1}{36}$$

Since $\frac{1}{36} \neq \frac{1}{6} \cdot \frac{1}{36}$, \mathcal{E} and $\mathcal{F}\mathcal{G}$ are not independent

Definition 22 : $\mathcal{E}, \mathcal{F}, \mathcal{G}$ are independent if

- \mathcal{E}, \mathcal{F} and \mathcal{E}, \mathcal{G} and \mathcal{F}, \mathcal{G} are ind.
(pairs. ind.)

$$\bullet P(\mathcal{E} \mathcal{F} \mathcal{G}) = P(\mathcal{E}) P(\mathcal{F}) P(\mathcal{G})$$

Now, \mathcal{E} and $\mathcal{F}\mathcal{G}$ are ind.

$$\begin{aligned} P(\mathcal{E} \cdot (\mathcal{F}\mathcal{G})) &= P(\mathcal{E} \mathcal{F} \mathcal{G}) = P(\mathcal{E}) \underbrace{P(\mathcal{F}) P(\mathcal{G})}_{\substack{\text{ind.} \\ \text{of} \\ \mathcal{F}, \mathcal{G}}} \\ &= P(\mathcal{E}) P(\mathcal{F}\mathcal{G}) \end{aligned}$$

Remark: Σ, F, G ind $\Rightarrow \Sigma$ and $(F \cup G)$ ind.

$$P(\Sigma(F \cup G)) = P(\Sigma F \cup \Sigma G)$$

$$= P(\Sigma F) + P(\Sigma G) - P(\Sigma F G) \quad \underbrace{= P(FG)}$$

$$= P(\Sigma)P(F) + P(\Sigma)P(G) - P(\Sigma)P(F)P(G)$$

$$= P(\Sigma)(P(F) + P(G) - P(FG))$$

$$= P(\Sigma)P(F \cup G)$$

Generalized
Definition

$\Sigma_1, \dots, \Sigma_n$ are independent iff
for every subset of $\Sigma_1, \dots, \Sigma_m$:

$$P(\Sigma_1 \dots \Sigma_m) = P(\Sigma_1) \dots P(\Sigma_m)$$

Examples: • Sequences of experiments; Σ_i refers to i -th
execution of experiment

E.g.: Rolling die

• Sequences of disease tests:

Intuition for one test being positive for
people w/ and w/o disease:

$$P(J|D) = .99$$

$$P(\bar{J}|\bar{D}) = .99$$

$$P(J|\bar{D}) = .01$$

prob. of false positives

- sequences of disease tests:
intuition for one test being positive for people w/ and w/o disease:

$$P(T|D) = .99$$

↗
sensitivity

$$P(\bar{T}|\bar{D}) = .99$$

$$P(T|\bar{D}) = .01$$

prob. of false positives

- let T_1, T_2 be two applications of test to same person w/ pos. outcome

$$P(T_1 T_2 | D) \quad \frac{.99}{100} \cdot \frac{.99}{100} \approx .98$$

cond.
independ-
ence

$$\stackrel{?}{\sim} P(T_1 | D) \cdot P(T_2 | D)$$

sensitivity does not
go much down

$$P(T_1 T_2 | \bar{D}) (= P((T_1 \cap T_2) | \bar{D}))$$

$$\stackrel{?!}{=} P(T_1 | \bar{D}) \cdot P(T_2 | \bar{D}) = \frac{1}{100} \cdot \frac{1}{100} = \frac{1}{10000}$$

probability for false positives is small

- Sequences of disease tests

Example probabilities for a test being positive for people w/ and w/o disease

$$P(T|D) = .99 \quad \swarrow$$

sensitivity

$$P(\bar{T}|\bar{D}) = .99 \quad \swarrow$$

specificity

$$\Rightarrow P(T|\bar{D}) = .01$$

If $P(D)$ is low (e.g., 1%), then we cannot rely on the test because there are as many true positive as false positive test results.

• What can we do?

Idea: Apply the test twice! First \mathcal{T}_1 , then \mathcal{T}_2 .

But: Need to ensure that probabilities multiply!

That is

$$\begin{aligned} P(\mathcal{T}_1, \mathcal{T}_2 | \mathcal{D}) &= P(\mathcal{T}_1 | \mathcal{D}) P(\mathcal{T}_2 | \mathcal{D}) \\ &= \frac{99}{100} \cdot \frac{99}{100} = \frac{9801}{10000} \approx 98\% \end{aligned}$$

numbers from our example
↙

This property of $\mathcal{T}_1, \mathcal{T}_2, \mathcal{D}$ is spelt out as

" \mathcal{T}_1 and \mathcal{T}_2 are conditionally independent given \mathcal{D} "

It is independence of $\mathcal{T}_1, \mathcal{T}_2$ with regard to the probability measure

$$P(\cdot | \mathcal{D})$$

remember that for every probability $P(\cdot)$ on \mathcal{S} also $P(\cdot | \mathcal{D})$ is a probability on \mathcal{S} .

Assume that T_1, T_2 are also independent given \bar{D} .

Then

$$\begin{aligned} P(T_1, T_2 | \bar{D}) &= P(T_1 | \bar{D}) \cdot P(T_2 | \bar{D}) \\ &= \frac{1}{100} \cdot \frac{1}{100} = \frac{1}{10000} \end{aligned}$$

numbers from our example
↙

This shows that the probability of false positives has been sharply reduced:

The relationship between false and true positives now is

$$\frac{1}{9801}$$

numbers from our example
↙

A Bayesian analysis of what can be concluded from the other test results ($T_1 \bar{T}_2$, $\bar{T}_1 T_2$ and $\bar{T}_1 \bar{T}_2$) will be part of the assignment.

What is $P(\mathcal{D} | \overline{T}_1 \overline{T}_2)$?

$$= P(\mathcal{D} | T_1) P(\overline{T}_2 | \mathcal{D}) = \frac{99}{100} \cdot \frac{1}{100} = \frac{99}{10^4}$$

$$P(T_1 \overline{T}_2 \cup \overline{T}_1 T_2 | \mathcal{D}) \approx \frac{200}{10,000} = \frac{2}{100}$$

$$\begin{aligned} P(\mathcal{D} | T_1 \overline{T}_2 \cup \overline{T}_1 T_2) &= \frac{P(\mathcal{D})}{P(T_1 \overline{T}_2 \cup \overline{T}_1 T_2 | \mathcal{D})} = \frac{P(\mathcal{D})}{2 P(T_1 \overline{T}_2 | \mathcal{D}) \cdot P(\mathcal{D})} \\ &= \frac{1}{2 P(T_1 \overline{T}_2)} \end{aligned}$$

$$\begin{aligned} P(T_1 \overline{T}_2) &= P(T_1 \overline{T}_2 | \mathcal{D}) \cdot P(\mathcal{D}) + P(T_1 \overline{T}_2 | \overline{\mathcal{D}}) \cdot P(\overline{\mathcal{D}}) \\ &= \frac{99}{10^4} \cdot \frac{1}{100} + \frac{99}{10^4} \cdot \frac{99}{100} \end{aligned}$$

$$\begin{aligned} &= \frac{2 \cdot \frac{99}{10^4} \cdot \frac{1}{100}}{2 \cdot \frac{99 + 99^2}{10^6}} = \frac{99}{(1+99) \cdot 99} = \boxed{\frac{1}{100}} \end{aligned}$$

What is $P(\overline{D} | \overline{S}_1, \overline{S}_2)$?

$$= P(\overline{S}_1 | \overline{D}) P(\overline{S}_2 | \overline{D}) = \frac{1}{100} \cdot \frac{99}{100} = \frac{99}{10^4}$$

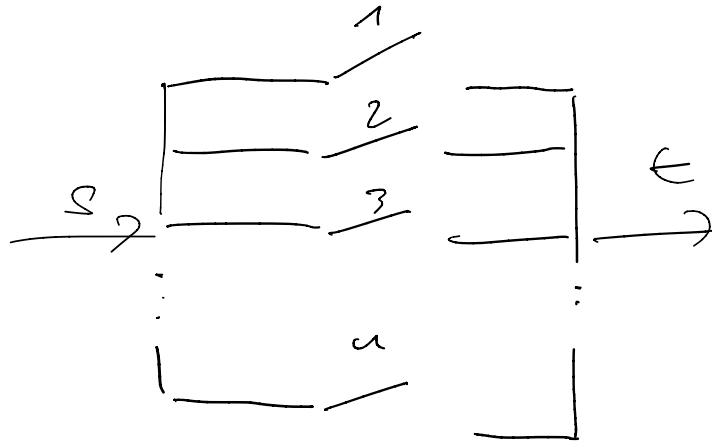
$$P(\overline{S}_1, \overline{S}_2 \cup \overline{S}_1, S_2 | \overline{D}) \approx \frac{200}{10,000} = \frac{2}{100}$$

$$P(\overline{D} | \overline{S}_1, \overline{S}_2 \cup \overline{S}_1, S_2) = \frac{P(\overline{D} | \overline{S}_1, \overline{S}_2 \cup \overline{S}_1, S_2) \cdot P(\overline{D})}{P(\overline{S}_1, \overline{S}_2 \cup \overline{S}_1, S_2)} = \frac{2 P(\overline{S}_1, \overline{S}_2 | \overline{D}) \cdot P(\overline{D})}{2 P(\overline{S}_1, \overline{S}_2)}$$

$$P(\overline{S}_1, \overline{S}_2) = P(\overline{S}_1, \overline{S}_2 | D) \cdot P(D) + P(\overline{S}_1, \overline{S}_2 | \overline{D}) \cdot P(\overline{D})$$
$$= \frac{99}{10^4} \cdot \frac{1}{100} + \frac{99}{10^4} \cdot \frac{99}{100}$$

$$= \frac{2 \cdot \frac{99}{10^4} \cdot \frac{99}{100}}{2 \cdot \frac{99 + 99^2}{10^6}} = \frac{99^2}{(1+99) \cdot 99} = \frac{99}{100}$$

Example 23



components are ind.

work with p_i^c for

comp i

System works iff ≥ 1 comp. works

E = "system works", F_i = "comp i works" $\Rightarrow P(E) = ?$

\bar{E} = "system doesn't work" iff no comp. works
iff \bar{F}_1 and $\bar{F}_2 \dots \bar{F}_n$

Note: (F_i) ind $\Rightarrow (\bar{F}_i)$ ind

$$\begin{aligned} P(E) &= 1 - P(\bar{E}) = 1 - P(\bar{F}_1 \dots \bar{F}_n) = 1 - P(\bar{F}_1) \dots P(\bar{F}_n) \\ &= 1 - \prod_{i=1}^n P(\bar{F}_i) = 1 - \prod_{i=1}^n (1 - P(F_i)) = 1 - \prod_{i=1}^n (1 - p_i^c) \end{aligned}$$