| Basic Probability Definitions |  |
| :---: | :---: |
| Sample <br> Space ( $\Omega$ ) | Set of all possible outcomes of a random experiment. |
| Event | Outcome of a random experiment (inside $\Omega$ ) |
| $\sigma$-field | The allowable events constitute a family of sets F, usually referred to as $\sigma$-field. Each set in F is a subset of the sample space $\Omega$. |
| Probability measure (P) | A probability measure on $(\Omega$, <br> $F)$ is a function $P: F \rightarrow[0,1]$ <br> that satisfies the following two properties: <br> 1. $P[\Omega]=1$ <br> 2. The probability of the union of a collection of disjoint members is the sum of its probabilities |
| Probability space | ( $\Omega$, F, P) |
| Basic <br> properties <br> of probab- <br> ility <br> measures | $\begin{aligned} & P[\varnothing]=0 \\ & P\left[A^{-}\right]=1-P[A] \\ & \text { If } A \subset B \text {, then } P[B]=P[A]+ \\ & P[B \backslash A] \geq P[A] \\ & P[A \cup B]=P[A]+P[B]-P[A \\ & \cap B] \end{aligned}$ |
| Inclusion <br> Exclusion <br> Principle (comes from last basic property of probability measures) | Given sets A1, A2... <br> P[union(Ai)] $\leq \operatorname{sum}(P[A i])$ <br> When the two events are disjoint, the inequality is = as they don't share any common space: $P[A \cap B]=0$ |



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| Basic Probability Definitions (cont) |  |
| :---: | :---: |
| Sampling strategy | Choose repeatedly a random number in $\Omega$ |
| Sampling <br> with <br> replac- <br> ement | Select random numbers in $\Omega$, without taking into account which ones you've already tested. Therefore, there will be some numbers tested multiple times |
| Sampling <br> without replacement | Select random numbers in $\Omega$ taking into account which ones you've already used. Therefore, you won't run the algorithm with the same number more than once |
| Indepe- <br> ndent <br> (events <br> or family) | Two events are independent if: $P[A \cap B]=P[A] P[B]$ <br> It also applies to families $\{\mathrm{Ai}, \mathrm{i} \in$ I\} |
| Pairwise | To form all possible pairs (two items at a time) from a set |
| Pairwise <br> indepe- <br> ndent <br> (family or events) | A family or events are pairwise independent if: <br> $P[A i \cap A j]=P[A i] P[A j]$ for all $i$ ! $=$ j <br> In english terms, a family or events is pairwise independent if any of its possible pairs is independent of each other. For example: $\begin{aligned} & P(A \cap B)=P(A) P(B) \\ & P(A \cap C)=P(A) P(C) \\ & P(B \cap C)=P(B) P(C) \end{aligned}$ |

Basic Probability Definitions (cont)
Mutually More than two events (i.e. indepe- $A, B, C)$ are mutually independent ndent if:
(events) 1. They are pairwise independent
2. They meet the condition: $P(A \cap B \cap C)=P(A) \times P(B) \times$ $P(C)$
In plain english, events are mutually independent if any event is independent to the other events

Condit- If $\mathrm{P}[\mathrm{B}]>0$, the conditional ional probability that A occurs give Probab- that $B$ occurs is: $P[A \mid B]=P[-$ ility $\quad A \cap B] / P[B]$
Condit- If $A$ and $B$ are independent ional events, then:
Probab- $\quad P[A \mid B]=P[A \cap B] / P[B]=$
ility $\quad\left(P[A]^{*} P[B]\right) / P[B]=P[A]$
(indep-
endent
events)
Law of Let e1...en be partitions of $\Omega$
Total
Probab-
ility (a collection of ALL the sets in $\Omega$ which are independent of each other). Also assuming $\mathrm{P}[\mathrm{ei}]>0$ for all i. The probability of A can be written as:
$P[A]=\operatorname{sum}(i=1, n)\left(P[A \mid e i]^{*} P[e i]\right)$ In english, it's the sum of all the possible scenarios in which A can occur

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## MVE137 - Chalmers University Cheat Sheet

by Delegado FM (Learningbizz) via cheatography.com/73767/cs/34116/

Basic Probability Definitions (cont)
Bayes
Assuming e1...en be partitions
Theorem of $\Omega$ :
$P[e j \mid B]=P[E j \cap B] / P[B]=$ $(\mathrm{P}[\mathrm{B} \mid \mathrm{Ej}] P[E j]) /(\operatorname{sum}(i=1, n)(P[-$ B/eiP[ei])
It's basically using conditional theory and then applying conditional theory again for the top part and law of total probability in the lower part

## Discrete Random Variables and Expect- <br> ation

Random A random variable $X$ on a
Variable sample space $\Omega$ is a real-valued (measurable) function on $\Omega$; that is $X: \Omega \rightarrow R$.
Denoted as upper case in this course and real numbers as lower case

Discrete A discrete random variable is a Random random variable that outputs Variable only a finite or countably infinite number of values (i.e. number of kids in a family, range between 1 and $x$ )

Probab- $\quad$ Sum of all the events $w$ in $\Omega$
ility that which $\mathrm{X}(\mathrm{w})=\mathrm{x}$
$X=a$


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Discrete Random Variables and Expect-
ation (cont)
Indepe- Two random variables X and Y
ndence are independent if and only if:
of $\quad P[(X=x) \cap(Y=y)]=P[X=x]$ -
random *P[Y=y]
variables for all values $x$ and $y$
Mutually Like mutually independent
indepe- events
ndent
random
variables
Expect- It is a weighted average of the
ation values assumed by the random
(mean) variable, taking into account the probability of getting that value.
The expectation of a discrete random variable X , denoted by $E[X]$ is given by $E[X]=\operatorname{sum}(i=x, X)\left(x^{*} P[X=x]\right)$

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