



Security and Cryptography 1

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Module 3: A little reminder on discrete probability

Disclaimer: Thanks to Günter Schäfer, large parts actually taken from Dan Boneh

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You have been introduced to some ***typical actors***

You understand what ***threats*** are ... and what this means

You can tell ***security goals*** (CIA!) from ***security services***

You know ***adversary models*** and which aspects they define

You've learned the tuple (M,C,K,E,D)

You can tell the difference of

- Cryptology
- Cryptography
- Cryptanalysis

You know basic cryptanalysis (statistics/combinatorics)

You know transposition and substitution

You have learned the shift cipher

... Mono- and Polyalphabetic substitution ciphers

... Product ciphers

You know CT only attacks

And you have heard of the legends of Enigma and Alan Turing...

A few words on space and time

Probability distributions

Random variables

Independence

Randomized algorithms

Some characteristics of XOR

Recall shift cipher (Caesar)

- How many combinations of possible keys?

$$2^5$$

- random guessing, how long do you need on average?

$$2^4$$

Recall random permutation and Vigenère

- How large were the key spaces?

$$2^{88}$$

$$2^{122}$$

- How many random trials?

$$2^{87}$$

$$2^{121}$$

How large is large? (Some context)

Reference Numbers Comparing Relative Magnitudes

Reference

Magnitude

Seconds in a year	$\approx 3 \quad * 10^7$
Seconds since creation of solar system	$\approx 2 \quad * 10^{17} \approx 4.6 * 10^9 \text{ y}$
Clock cycles per year (50 MHz computer)	$\approx 1.6 \quad * 10^{15}$
Instructions per year (i7 @ 3.0 GHz)	$\approx 2^{63} \approx 7.15 \quad * 10^{18}$
Binary strings of length 64	$2^{64} \approx 1.8 \quad * 10^{19}$
Binary strings of length 128	$2^{128} \approx 3.4 \quad * 10^{38}$
Binary strings of length 256	$2^{256} \approx 1.2 \quad * 10^{77}$
Number of 75-digit prime numbers	$\approx 5.2 \quad * 10^{72}$
Number of 80-digit prime numbers	$\approx 5.4 \quad * 10^{77}$
Electrons in the universe	$\approx 8.37 \quad * 10^{77}$

Brute force attack, try all keys until intelligible plaintext found:

- Every crypto algorithm (save: OTP) can be attacked by brute force
- On average, half of all possible keys will have to be tried

Average Time Required for Exhaustive Key Search

<i>Key Size [bit]</i>	<i>Number of keys</i>	<i>Time required at 1 encryption / μs</i>	<i>Time required at 10^6 encryption / μs</i>
32	$2^{32} = 4.3 * 10^9$	$2^{31} \mu$ s = 35.8 minutes	2.15 milliseconds
56	$2^{56} = 7.2 * 10^{16}$	$2^{55} \mu$ s = 1142 years	10.01 hours
128	$2^{128} = 3.4 * 10^{38}$	$2^{127} \mu$ s = $5.4 * 10^{24}$ years	$5.4 * 10^{18}$ years
(Vigenère 26 chars	$26! = 4 * 10^{26}$	$2^{88} \mu$ s = $6.4 * 10^{12}$ years	$6.4 * 10^6$ years)

i7 could get to around 10^4 encryptions/ μ s
GPU can perform around 7×10^5 hashes

Time since human/chimpanzee lines diverged:
 5×10^6 years,
Homo sapiens: 5×10^4 years

U : finite set (e.g. $U = \{0,1\}^n$)

Def: **Probability distribution** P over U is a function $P: U \rightarrow \mathbb{R}^+$

more specifically $P(x) \in [0,1]$

such that $\sum P(x) = 1$

Examples:

1. Uniform distribution: for all $x \in U$: $P(x) = 1/|U|$
2. Point distribution at x_0 : $P(x_0) = 1$, $\forall x \neq x_0$: $P(x) = 0$

Fair coin:

$$U = \{H, T\}, P(H) = P(T) = \frac{1}{2}$$

Dice:

$$U = \{1, \dots, 6\}, P(x) = 1/6, 1 \leq x \leq 6$$

Two dice (combined):

$$U = \{1, \dots, 6\}^2 = \{(x_1, x_2) : 1 \leq x_1, x_2 \leq 6\}$$

$$P(x_1, x_2) = 1/36 \text{ for all } x_1, x_2$$

Uniform distributions for all of the above

Two dice (sum):

$$U = \{2, 3, \dots, 12\}$$

$$P(2) = 1/36, P(3) = 2/36, \dots, P(12) = 1/36$$

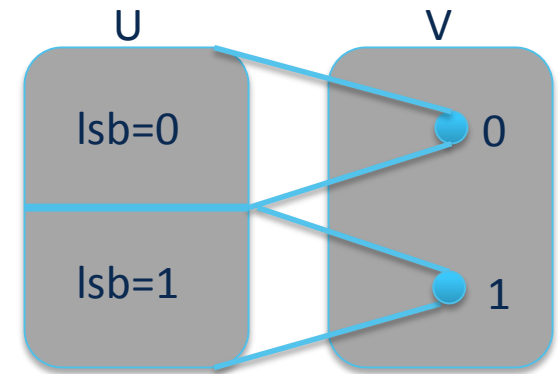


Def: a random variable X is a function $X:U \rightarrow V$

Example: $X: \{0,1\}^n \rightarrow \{0,1\}$; $X(y) = \text{lsb}(y) \in \{0,1\}$

For the uniform distribution on U :

$$\Pr[X=0] = 1/2, \quad \Pr[X=1] = 1/2$$



More generally:

rand. var. X induces a distribution on V : $\Pr[X=v] := \Pr[X^{-1}(v)]$


Let U be some set, e.g. $U = \{0,1\}^n$

We write $r \stackrel{R}{\leftarrow} U$ to denote a uniform random variable over U

$$\text{for all } a \in U: \Pr[r = a] = 1/|U|$$

Let r be a uniform random variable on $\{0,1\}^2$

Define the random variable $X = r_1 + r_2$

Then $\Pr[X=2] =$ 

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Let r be a uniform random variable on $\{0,1\}^2$

Define the random variable $X = r_1 + r_2$

Then $\Pr[X=2] = \frac{1}{4}$

For a set $A \subseteq U$: $\Pr[A] = \sum_{x \in A} P(x) \in [0,1]$

note: $\Pr[U]=1$

The set A is called an **event**

Example: $U = \{0,1\}^8$

$A = \{ \text{all } x \text{ in } U \text{ such that } \text{lsb}_2(x)=11 \} \subseteq U$

for the uniform distribution on $\{0,1\}^8$: $\Pr[A]$



What about $U = \{0,1\}^n$ (with $n > 2$)?

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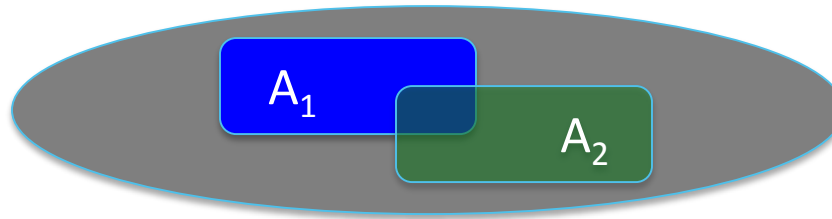
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for the uniform distribution on $\{0,1\}^8$: $\Pr[A] = 1/4$

What about $U = \{0,1\}^n$ (with $n > 2$)?

For events A_1 and A_2

$$\Pr[A_1 \cup A_2] \leq \Pr[A_1] + \Pr[A_2]$$



$$A_1 \cap A_2 = \emptyset \rightarrow \Pr[A_1 \cup A_2] = \Pr[A_1] + \Pr[A_2]$$

Example:

$$A_1 = \{ \text{all } x \text{ in } \{0,1\}^n \text{ s.t. } \text{lsb}_2(x)=11 \} ;$$

$$A_2 = \{ \text{all } x \text{ in } \{0,1\}^n \text{ s.t. } \text{msb}_2(x)=11 \}$$

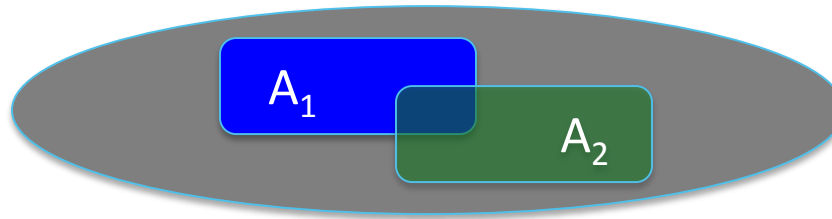
$$\Pr[\text{lsb}_2(x)=11 \text{ or } \text{msb}_2(x)=11] = \Pr[A_1 \cup A_2] \leq$$



When is it less than 1/2?

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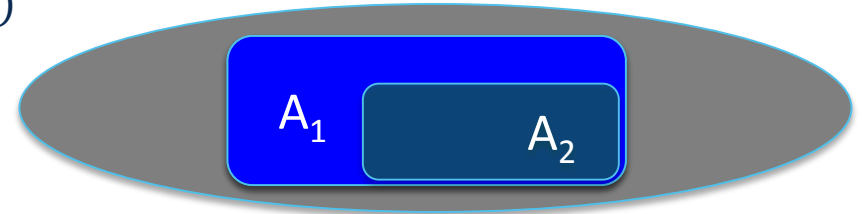
$$\Pr[\text{lsb}_2(x)=11 \text{ or } \text{msb}_2(x)=11] = \Pr[A_1 \cup A_2] \leq \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$$

When is it less than 1/2?

Suppose event $A \subseteq U$.

The *induced* probability P_A is defined as:

$$P_A(x) = \frac{P(x)}{P(A)} \text{ for } x \in A$$



For two events, $A_1, A_2 \subseteq U$ we say that $P_{A_1}(A_2)$, or $P(A_2|A_1)$ is

$$P(A_2|A_1) = P_{A_1}(A_1 \cap A_2) = \frac{P(A_1 \cap A_2)}{P(A_1)}$$

For events A_1 and A_2 :

$$A_1 = \{x \text{ in } \{0,1\}^3 : \text{msb}(x) = 1\}$$

$$A_2 = \{x \text{ in } \{0,1\}^3 : x < 110\}$$

$$P(A_1 \cap A_2) = \{x \text{ in } \{0,1\}^3 : \text{msb}(x) = 1, x < 110\}$$

0	0	0
0	0	1
0	1	0
0	1	1
1	0	0
1	0	1
1	1	0
1	1	1

What if:
 $A_1 = \{x \text{ in } \{0,1\}^3 : \text{lsb}(x) = 1\}$?

$$P(A_1) =$$

$$P(A_1 \cap A_2) =$$

$$P(A_2|A_1) =$$

...we call A_1 and A_2 **independent** if $P(A_2|A_1) = P(A_2)$

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0	0	0
0	0	1
0	1	0
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1	0	0
1	0	1
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What if:

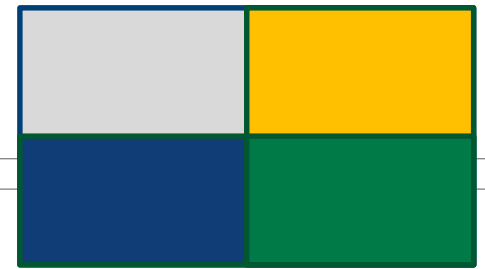
$$A_1 = \{x \text{ in } \{0,1\}^3 : lsb(x) = 1\}$$

$$P(A_1) = 1/2$$

$$P(A_1 \cap A_2) = 2/8 = 1/4$$

$$P(A_2|A_1) = \frac{1/4}{1/2} = 1/2$$

...we call A_1 and A_2 **independent** if $P(A_2|A_1) = P(A_2)$



Definition:

Events A_1 and A_2 are **independent**, if $\Pr[A_1 \text{ and } A_2] = \Pr[A_1] \times \Pr[A_2]$

Random variables X, Y , both taking values in V are **independent** if

$$\forall a, b \in V: \Pr[X=a \text{ and } Y=b] = \Pr[X=a] \times \Pr[Y=b]$$

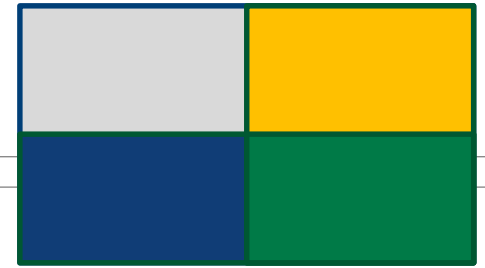
Example:

$$U = \{0,1\}^2 = \{00, 01, 10, 11\} \quad \text{and} \quad r \xleftarrow{R} U$$

Define random variables X and Y as: $X = \text{lsb}(r)$, $Y = \text{msb}(r)$

$$\Pr[X=0 \text{ and } Y=0] =$$





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$$\Pr[X=0 \text{ and } Y=0] = \Pr[r=00] = \frac{1}{4} = \Pr[X=0] \times \Pr[Y=0]$$