

CS 5602 Introduction to Cryptography

Lecture 20

Information Theoretic Security

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Introduction

- Another practical approach, related to computational security, is to reduce breaking the system to solving some well-studied hard problem.
- For example, we can try to show that a given system is secure if a given integer N cannot be factored.
- Systems of this form are often called provably secure.
- However, we only have a proof relative to some hard problem, hence this does not provide an absolute proof.

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Goals

- To introduce the concept of perfect secrecy.
- To discuss the security of the one-time pad.
- To introduce the concept of entropy.
- To explain the notion of key equivocation, spurious keys and unicity distance.
- To use these tools to understand why the prior historical encryption algorithms are weak.
- Slides are taken right from the book

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Introduction

- Essentially, a computationally secure scheme, or one which is provably secure, is only secure when we consider an adversary whose computational resources are bounded.
- Even if the adversary has large, but limited, resources she still will not break the system.
- We need to be careful about the key sizes etc. If the key size is small then our adversary may have enough computational resources to break the system.
- We need to keep abreast of current algorithmic developments and developments in computer hardware.

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Introduction

- We first need to overview the difference between information theoretic security and computational security.
- Informally, a cryptographic system is called computationally secure if the best possible algorithm for breaking it requires N operations, where N is such a large number that it is infeasible to carry out this many operations.
- With current computing power we assume that 2^{80} operations is an infeasible number of operations to carry out.
- Hence, a value of N larger than 2^{80} would imply that the system is computationally secure.
- No actual system can be proved secure under this definition, since we never know whether there is a better algorithm than the one known.
- In practice we say a system is computationally secure if the best known algorithm for breaking it requires an unreasonably large amount of computational resources.

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Introduction

- At some point in the future we should expect our system to become broken, either through an improvement in computing power or an algorithmic breakthrough.
- Quantum computing?
- It turns out that most schemes in use today are computationally secure, and so every chapter in this book (except this one) will solely be interested in computationally secure systems.

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Unconditional Security

- On the other hand, a system is said to be unconditionally secure when we place no limit on the computational power of the adversary.
- In other words a system is unconditionally secure if it cannot be broken even with infinite computing power.
- Hence, no matter what algorithmic improvements are made or what improvements in computing technology occur, an unconditionally secure scheme will never be broken.
- Other names for unconditional security you find in the literature are perfect security or information theoretic security.

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Encryption & Decryption Functions

- We write e_k to represent encryption using the key k
- We write d_k to represent decryption using the key k
- We assume that P and K are independent variables
- The user does not select a key based on the content of a message (very bad)
- The set of ciphertexts under a specific key k is defined by
$$C(k) = \{e_k(x) \mid x \in P\}$$
- We then have the relationship
$$p(C = c) = \sum_{k: c \in C(k)} p(K = k) \cdot p(P = d_k(c)),$$

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Cryptographic Hierarchy

- You have already seen that the following systems are not computationally secure, since we already know how to break them with very limited computing resources:
 - Shift cipher
 - Substitution cipher
 - Ballaso (Vigenere) cipher
- Of the systems we shall meet later, the following are computationally secure but are not unconditionally secure:
 - DES and Rijndael,
 - RSA,
 - ElGamal encryption
- However, the one-time pad which we shall meet in this chapter is unconditionally secure, but only if it is used correctly.

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Key Example

- $P = \{a, b, c, d\}$
- $p(P = a) = 1/4$
- $p(P = b) = 3/10$
- $p(P = c) = 3/20$
- $p(P = d) = 3/10$
- $K = \{k_1, k_2, k_3\}$
- $p(K = k_1) = 1/4$
- $p(K = k_2) = 1/2$
- $p(K = k_3) = 1/4$

| | a | b | c | d |
|----------------|---|---|---|---|
| k ₁ | 3 | 4 | 2 | 1 |
| k ₂ | 3 | 1 | 4 | 2 |
| k ₃ | 4 | 3 | 1 | 2 |

- Time for some calculations

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Probability and Ciphers

- Let P denote the set of possible plaintexts.
- Let K denote the set of possible keys.
- Let C denote the set of ciphertexts.
- Each of these can be thought of as a probability distribution, where we denote the probabilities by $p(P = m)$, $p(K = k)$, $p(C = c)$.
- We will often write P, K , and C for random variables associated with P, K , and C
- So for example, if our message space is $P = \{\alpha, \beta, \gamma\}$ and the message α occurs with probability $1/4$ then we write $p(P = \alpha) = 1/4$.

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Some Calculations

$$p(C = 1) = p(K = k_1)p(P = d) + p(K = k_2)p(P = b) + p(K = k_3)p(P = c) = 0.2625,$$

$$p(C = 2) = p(K = k_1)p(P = c) + p(K = k_2)p(P = d) + p(K = k_3)p(P = d) = 0.2625,$$

$$p(C = 3) = p(K = k_1)p(P = a) + p(K = k_2)p(P = a) + p(K = k_3)p(P = b) = 0.2625,$$

$$p(C = 4) = p(K = k_1)p(P = b) + p(K = k_2)p(P = c) + p(K = k_3)p(P = a) = 0.2125.$$

| | a | b | c | d |
|----------------|---|---|---|---|
| k ₁ | 3 | 4 | 2 | 1 |
| k ₂ | 3 | 1 | 4 | 2 |
| k ₃ | 4 | 3 | 1 | 2 |

| | a | b | c | d |
|---|--------|--------|--------|--------|
| P | 0.2500 | 0.3000 | 0.1500 | 0.3000 |
| K | 0.2500 | 0.5000 | 0.2500 | 0.0000 |
| C | 0.2625 | 0.2625 | 0.2125 | 0.2625 |

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Some Calculations

| | a | b | c | d |
|-------|---|---|---|---|
| k_1 | 3 | 4 | 2 | 1 |
| k_2 | 3 | 1 | 4 | 2 |
| k_3 | 4 | 3 | 1 | 2 |

$$p(C = c|P = m) = \sum_{k:m=d_k(c)} p(K = k)$$

$P = \{a, b, c, d\}$
 $p(P = a) = 1/4$
 $p(P = b) = 3/10$
 $p(P = c) = 3/20$
 $p(P = d) = 3/10$
 $K = \{k_1, k_2, k_3\}$
 $p(K = k_1) = 1/4$
 $p(K = k_2) = 1/2$
 $p(K = k_3) = 1/4$

$p(C = 1|P = a) = 0, \quad p(C = 2|P = a) = 0,$
 $p(C = 3|P = a) = 0.75, \quad p(C = 4|P = a) = 0.25,$
 $p(C = 1|P = b) = 0.5, \quad p(C = 2|P = b) = 0,$
 $p(C = 3|P = b) = 0.25, \quad p(C = 4|P = b) = 0.25,$
 $p(C = 1|P = c) = 0.25, \quad p(C = 2|P = c) = 0.25,$
 $p(C = 3|P = c) = 0, \quad p(C = 4|P = c) = 0.5,$
 $p(C = 1|P = d) = 0.25, \quad p(C = 2|P = d) = 0.75,$
 $p(C = 3|P = d) = 0, \quad p(C = 4|P = d) = 0.$

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Perfect Security (Secrecy)

- So in our previous example the ciphertext does reveal a lot of information about the plaintext. But this is exactly what we wish to avoid, we want the ciphertext to give no information about the plaintext.
- A system with this property, that the ciphertext reveals nothing about the plaintext, is said to be perfectly secure.
- Definition 5.1 (Perfect Secrecy). A cryptosystem has perfect secrecy if
$$p(P = m | C = c) = p(P = m)$$
- for all plaintexts m and all ciphertexts c.

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Some Calculations

Anybody recognize this?

| | a | b | c | d |
|-------|---|---|---|---|
| k_1 | 3 | 4 | 2 | 1 |
| k_2 | 3 | 1 | 4 | 2 |
| k_3 | 4 | 3 | 1 | 2 |

$$p(P = m|C = c) = \frac{p(P = m)p(C = c|P = m)}{p(C = c)}$$

$P = \{a, b, c, d\}$
 $p(P = a) = 1/4$
 $p(P = b) = 3/10$
 $p(P = c) = 3/20$
 $p(P = d) = 3/10$
 $K = \{k_1, k_2, k_3\}$
 $p(K = k_1) = 1/4$
 $p(K = k_2) = 1/2$
 $p(K = k_3) = 1/4$

$p(P = a|C = 1) = 0, \quad p(P = b|C = 1) = 0.571,$
 $p(P = c|C = 1) = 0.143, \quad p(P = d|C = 1) = 0.286,$
 $p(P = a|C = 2) = 0, \quad p(P = b|C = 2) = 0,$
 $p(P = c|C = 2) = 0.143, \quad p(P = d|C = 2) = 0.857,$
 $p(P = a|C = 3) = 0.714, \quad p(P = b|C = 3) = 0.286,$
 $p(P = c|C = 3) = 0, \quad p(P = d|C = 3) = 0,$
 $p(P = a|C = 4) = 0.294, \quad p(P = b|C = 4) = 0.352,$
 $p(P = c|C = 4) = 0.352, \quad p(P = d|C = 4) = 0.$

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Lemma 5.2+

- Lemma 5.2+. A cryptosystem has perfect secrecy iff $p(C = c|P = m) = p(C = c)$ for all m and c.
- Use Bayes's Theorem

PROOF. This trivially follows from the definition

$$p(P = m|C = c) = \frac{p(P = m)p(C = c|P = m)}{p(C = c)}$$

and the fact that perfect secrecy means $p(P = m|C = c) = p(P = m)$.

- Note that if $p(P=m|C=c) = p(P=m)$ then $p(C=c|P=m) = p(C=c)p(P=m|C=c)/p(P=m) = p(C=c)$

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Some Conclusions

- If we see the ciphertext 1 then we know the message is not equal to a. We also can guess that it is more likely to be b rather than c or d.
- If we see the ciphertext 2 then we know the message is not equal to a or b. We also can be pretty certain that the message is equal to d.
- If we see the ciphertext 3 then we know the message is not equal to c or d and have a good chance that it is equal to a.
- If we see the ciphertext 4 then we know the message is not equal to d, but cannot really guess with certainty as to whether the message is a, b or c.

$p(P = a|C = 1) = 0, \quad p(P = b|C = 1) = 0.571,$
 $p(P = c|C = 1) = 0.143, \quad p(P = d|C = 1) = 0.286,$
 $p(P = a|C = 2) = 0, \quad p(P = b|C = 2) = 0,$
 $p(P = c|C = 2) = 0.143, \quad p(P = d|C = 2) = 0.857,$
 $p(P = a|C = 3) = 0.714, \quad p(P = b|C = 3) = 0.286,$
 $p(P = c|C = 3) = 0, \quad p(P = d|C = 3) = 0,$
 $p(P = a|C = 4) = 0.294, \quad p(P = b|C = 4) = 0.352,$
 $p(P = c|C = 4) = 0.352, \quad p(P = d|C = 4) = 0.$

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Lemma 5.3

- Lemma 5.3. Assume the cryptosystem is perfectly secure, then $|K| \geq |C| \geq |P|$ (Smart uses #K, etc., instead of |K|.)
- Proof: For any cryptosystem you must have $|C| \geq |P|$ since you must be able to recover the plaintext from the ciphertext
- We assume that every ciphertext can occur, i.e. $p(C = c) > 0$ for all $c \in C$, since if this does not hold then we can alter our definition of C.
- Then for any message m and any ciphertext c we have from Lemma 5.2+ that $p(C = c|P = m) = p(C = c) > 0$.
- This means for each m, that for all c there must be a key k such that $e_k(m) = c$.
- Thus, $|K| \geq |C|$

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Shannon's Theorem

- Theorem 5.4 (Shannon). Let $(P, C, K, e_k(\cdot), d_k(\cdot))$ denote a cryptosystem with $|P| = |C| = |K|$. Then the cryptosystem provides perfect secrecy iff
 - every key is used with equal probability $1/|K|$,
 - for each $m \in P$ and $c \in C$ there is a unique key k such that $e_k(m) = c$.
- Proof: (perfect secrecy \rightarrow two conditions)
- Suppose the system gives perfect secrecy. Then we have already seen for all $m \in P$ and $c \in C$ there is a key k such that $e_k(m) = c$. Now, since we have assumed $|C| = |K|$ we have $|\{e_k(m) \mid k \in K\}| = |K|$

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Example 1

- Modified Shift Cipher. Recall the shift cipher is one in which we "add" a given letter (the key) onto each letter of the plaintext to obtain the ciphertext.
- We now modify this cipher by using a different key for each plaintext letter.
- For example, to encrypt the message HELLO, we choose five random keys, say FUIAT.
- We then add the key onto the plaintext, modulo 26, to obtain the ciphertext MYTLH.
- Notice, how the plaintext letter L encrypts to different letters in the ciphertext.
- When we use the shift cipher with a different random key for each letter, we obtain a perfectly secure system.
- To see why this is so, consider the situation of encrypting a message of length n .
- Then the total number of keys, ciphertexts and plaintexts are all equal, namely: $\#K = \#C = \#P = 26^n$.
- In addition each key will occur with equal probability: $p(K = k) = 1/26^n$, and for each m and c there is a unique k such that $ek(m) = c$. Hence, by Shannon's Theorem this modified shift cipher is perfectly secure.
- This is a lame example in my opinion, because this is the one-time pad in a convoluted way

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Shannon's Theorem

We need to show that every key is used with equal probability, i.e.

$$p(K = k) = 1/\#K \text{ for all } k \in K.$$

Let $n = \#K$ and $P = \{m_i : 1 \leq i \leq n\}$, fix $c \in C$ and label the keys k_1, \dots, k_n such that $e_{k_i}(m_i) = c$ for $1 \leq i \leq n$.

We then have, noting that due to perfect secrecy $p(P = m_i | C = c) = p(P = m_i)$,

$$\begin{aligned} p(P = m_i) &= p(P = m_i | C = c) \\ &= \frac{p(C = c | P = m_i)p(P = m_i)}{p(C = c)} \\ &= \frac{p(K = k_i)p(P = m_i)}{p(C = c)}. \end{aligned}$$

Hence we obtain, for all $1 \leq i \leq n$,

$$p(C = c) = p(K = k_i).$$

This says that the keys are used with equal probability and hence

$$p(K = k) = 1/\#K \text{ for all } k \in K.$$

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Example 2 (one-Time Pad)

- The above modified shift cipher basically uses addition modulo 26.
- One problem with this is that in a computer, or any electrical device, mod 26 arithmetic is hard, ut binary arithmetic is easy.
- We are particularly interested in the addition operation, which is denoted by \oplus and is equal to the logical exclusive-or, or XOR, operation:

| | | |
|----------|---|---|
| \oplus | 0 | 1 |
| 0 | 0 | 1 |
| 1 | 1 | 0 |

- In 1917 Gilbert Vernam patented a cipher which used these principles, called the Vernam cipher or one-time pad.

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Shannon's Theorem (\leftarrow Part)

Now we need to prove the result in the other direction. Namely, if

- $\#K = \#C = \#P$,
- every key is used with equal probability $1/\#K$,
- for each $m \in P$ and $c \in C$ there is a unique key k such that $e_k(m) = c$,

then we need to show the system is perfectly secure, i.e.

$$p(P = m | C = c) = p(P = m).$$

We have, since each key is used with equal probability,

$$\begin{aligned} p(C = c) &= \sum_k p(K = k)p(P = d_k(c)) \\ &= \frac{1}{\#K} \sum_k p(P = d_k(c)). \end{aligned}$$

Also, since for each m and c there is a unique key k with $e_k(m) = c$, we must have

$$\sum_m p(P = d_k(c)) = \sum_m p(P = m) = 1.$$

Hence, $p(C = c) = 1/\#K$. In addition, if $c = e_k(m)$ then $p(C = c | P = m) = p(K = k) = 1/\#K$. So using Bayes' Theorem we have

$$\begin{aligned} p(P = m | C = c) &= \frac{p(P = m)p(C = c | P = m)}{p(C = c)} \\ &= \frac{p(P = m) \frac{1}{\#K}}{\frac{1}{\#K}} \\ &= p(P = m). \end{aligned}$$

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Example 2 (one-Time Pad)

- To send a binary string you need a key, which is a binary string as long as the message.
- To encrypt a message we XOR each bit of the plaintext with each bit of the key to produce the ciphertext.
- Each key is only allowed to be used once, hence the term one-time pad.
- This means that key distribution is a pain, a problem which we shall come back to again and again.
- To see why we cannot get away with using a key twice, consider the following chosen plaintext attack.

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Example 2 (one-Time Pad)

- We assume that Alice always uses the same key k to encrypt a message to Bob.
- Eve wishes to determine this key and so carries out the following attack:
 - Eve generates m and asks Alice to encrypt it.
 - Eve obtains $c = m \oplus k$.
 - Eve now computes $k = c \oplus m$.
- You may object to this attack since it requires Alice to be particularly stupid, in that she encrypts a message for Eve.
- *I disagree that Alice needs to be particularly stupid*
- But in designing our cryptosystems we should try and make systems which are secure even against stupid users.

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Entropy (Introduction)

- To simplify the key distribution problem we need to turn from perfectly secure encryption algorithms to ones which are, hopefully, computationally secure.
- This is the goal of modern cryptographers, where one aims to build systems such that
 - one key can be used many times,
 - one small key can encrypt a long message.
- Such systems will not be unconditionally secure, by Shannon's Theorem, and so must be at best only computationally secure.

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Example 2 (one-Time Pad)

- Another problem with using the same key twice is the following.
- Suppose Eve can intercept two messages encrypted with the same key
- $c_1 = m_1 \oplus k$,
- $c_2 = m_2 \oplus k$.
- Eve can now determine some partial information about the pair of messages m_1 and m_2 since she can compute $c_1 \oplus c_2 = (m_1 \oplus k) \oplus (m_2 \oplus k) = m_1 \oplus m_2$.
- Despite the problems associated with key distribution, the one-time pad has been used in the past in military and diplomatic contexts.

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Entropy (Introduction)

- The word entropy is another name for uncertainty, and the basic tenet of information theory is that uncertainty and information are essentially the same thing.
- This takes some getting used to, but consider that if you are uncertain what something means then revealing the meaning gives you information.
- As a cryptographic application suppose you want to determine the information in a ciphertext, in other words you want to know what its true meaning is,
 - you are uncertain what the ciphertext means,
 - you could guess the plaintext,
 - the level of uncertainty you have about the plaintext is the amount of information contained
- in the ciphertext.

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Entropy (Introduction)

- If every message we send requires a key as long as the message, and we never encrypt two messages with the same key, then encryption will not be very useful in everyday applications such as Internet transactions.
- This is because getting the key from one person to another will be an impossible task.
- After all one cannot encrypt it since that would require another key.
- This problem is called the key distribution problem.

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Entropy (Introduction)

- If X is a random variable, the amount of entropy (in bits) associated with X is denoted by $H(X)$, we shall define this quantity formally in a second. First let us look at a simple example to help clarify ideas.
- Suppose X is the answer to some question, i.e. Yes or No.
- If you know I will always say Yes then my answer gives you no information.
- So the information contained in X should be zero, i.e. $H(X) = 0$.
- There is no uncertainty about what I will say, hence no information is given by me saying it, hence there is zero entropy.
- If you have no idea what I will say and I reply Yes with equal probability to replying No then I am revealing one bit of information.
- Hence, we should have $H(X) = 1$.
- Note that entropy does not depend on the length of the actual message: in the above case we have a message of length at most three letters but the amount of information is at most one bit.

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Entropy

DEFINITION 5.5 (Entropy). Let X be a random variable which takes on a finite set of values x_i , with $1 \leq i \leq n$, and has probability distribution $p_i = p(X = x_i)$. The entropy of X is defined to be

$$H(X) = - \sum_{i=1}^n p_i \log_2 p_i.$$

Why - ?

We make the convention that if $p_i = 0$ then $p_i \log_2 p_i = 0$.

Let us return to our *Yes* or *No* question above and show that this definition of entropy coincides with our intuition. Recall, X is the answer to some question with responses *Yes* or *No*. If you know I will always say *Yes* then

$$p_1 = 1 \text{ and } p_2 = 0.$$

We compute

$$H(X) = -1 \cdot \log_2 1 - 0 \cdot \log_2 0 = 0.$$

Hence, my answer reveals no information to you.

If you have no idea what I will say and I reply *Yes* with equal probability to replying *No* then

$$p_1 = p_2 = 1/2.$$

We now compute

$$H(X) = -\frac{\log_2 \frac{1}{2}}{2} - \frac{\log_2 \frac{1}{2}}{2} = 1.$$

Hence, my answer reveals one bit of information to you.

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Upper Bound for Entropy

THEOREM 5.6 (Jensen's Inequality). Suppose

$$\sum_{i=1}^n a_i = 1$$

with $a_i > 0$ for $1 \leq i \leq n$. Then, for $x_i > 0$,

$$\sum_{i=1}^n a_i \log_2 x_i \leq \log_2 \left(\sum_{i=1}^n a_i x_i \right).$$

With equality occurring if and only if $x_1 = x_2 = \dots = x_n$.

Using this we can now prove the following theorem:

THEOREM 5.7. If X is a random variable which takes n possible values then

$$0 \leq H(X) \leq \log_2 n.$$

The lower bound is obtained if one value occurs with probability one, the upper bound is obtained if all values are equally likely.

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Entropy

- We always have $H(X) \geq 0$.
- The only way to obtain $H(X) = 0$ is if for some i we have $p_i = 1$ and $p_j = 0$ when $i \neq j$.
- If $p_i = 1/n$ for all i then $H(X) = \log_2 n$.
- Another way of looking at entropy is that it measures by how much one can compress the information.
- If I send a single ASCII character to signal *Yes* or *No*, for example I could simply send *Y* or *N*. I am actually sending 8 bits of data, but I am only sending one bit of information.
- If I wanted to I could compress the data down to 1/8th of its original size.
- Hence, naively if a message of length n can be compressed to a proportion ϵ of its original size then it contains $\epsilon \cdot n$ bits of information in it.

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Proof of Upper Bound

$$\begin{aligned} H(X) &= - \sum_{i=1}^n p_i \log_2 p_i \\ &= \sum_{i=1}^n p_i \log_2 \frac{1}{p_i} \\ &\leq \log_2 \left(\sum_{i=1}^n \left(p_i \times \frac{1}{p_i} \right) \right) \text{ by Jensen's inequality} \\ &= \log_2 n. \end{aligned}$$

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Baby Cryptosystem

Let us return to our baby cryptosystem considered in the previous section. Recall we had the probability spaces

$$\mathbb{P} = \{a, b, c, d\}, \mathbb{K} = \{k_1, k_2, k_3\} \text{ and } \mathbb{C} = \{1, 2, 3, 4\},$$

with the associated probabilities:

- $p(P = a) = 0.25$, $p(P = b) = p(P = d) = 0.3$ and $p(P = c) = 0.15$,
- $p(K = k_1) = p(K = k_3) = 0.25$ and $p(K = k_2) = 0.5$,
- $p(C = 1) = p(C = 2) = p(C = 3) = 0.2625$ and $p(C = 4) = 0.2125$.

We can then calculate the relevant entropies as:

$$\begin{aligned} H(P) &\approx 1.9527, \\ H(K) &\approx 1.5, \\ H(C) &\approx 1.9944. \end{aligned}$$

Hence the ciphertext 'leaks' about two bits of information about the key and plaintext, since that is how much information is contained in a single ciphertext. Later we will calculate how much of this information is about the key and how much about the plaintext.

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