Encryption and Decryption with Python

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Motivation



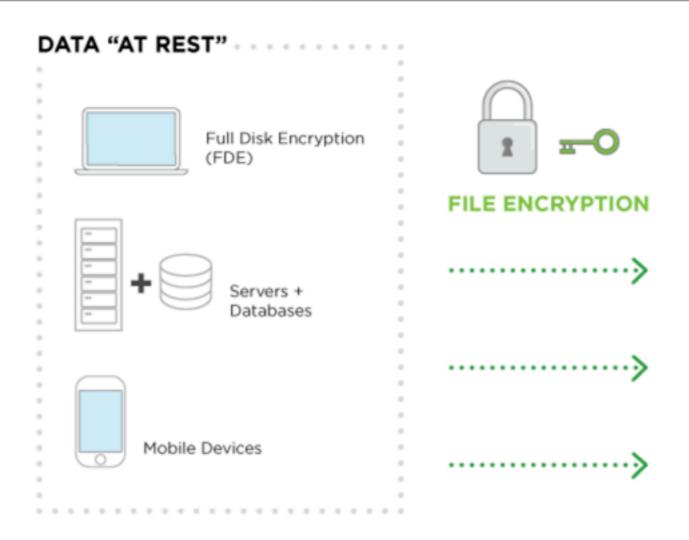




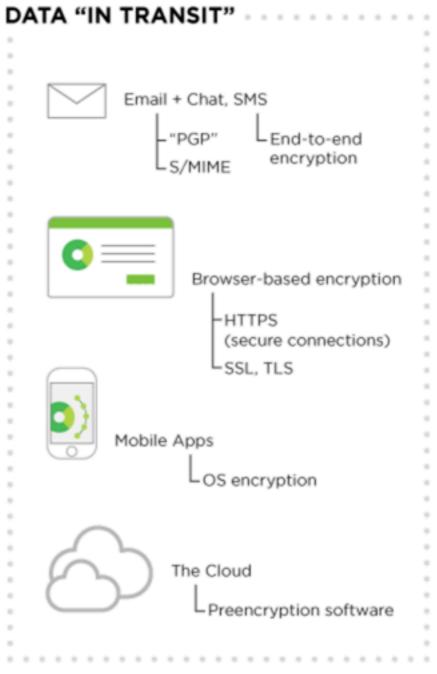




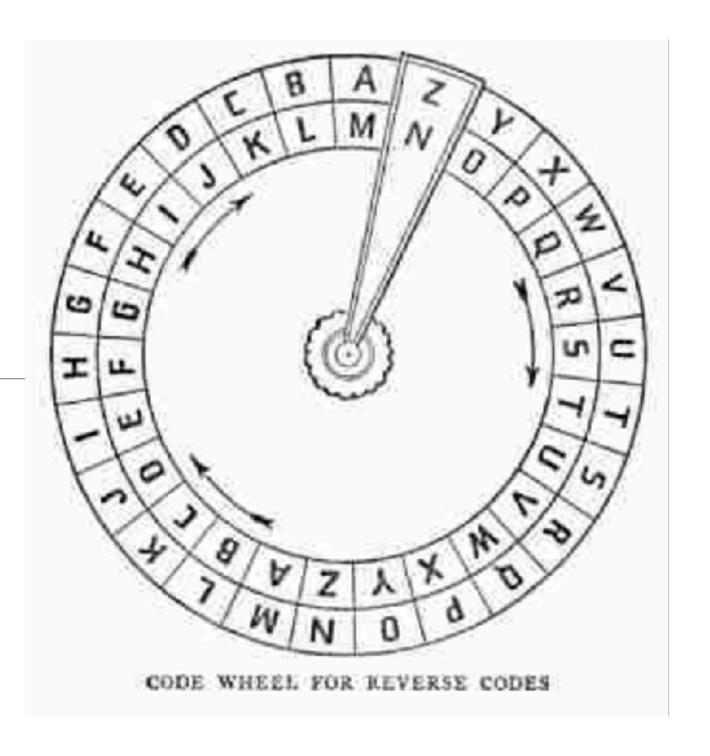
Motivation



- Plain text is encrypted into jumbled, unreadable cipher text by an algorithm.
- Cipher text is decrypted back into plain text with a key, a long string of numbers the algorithm uses to unscramble the data.



Encryption

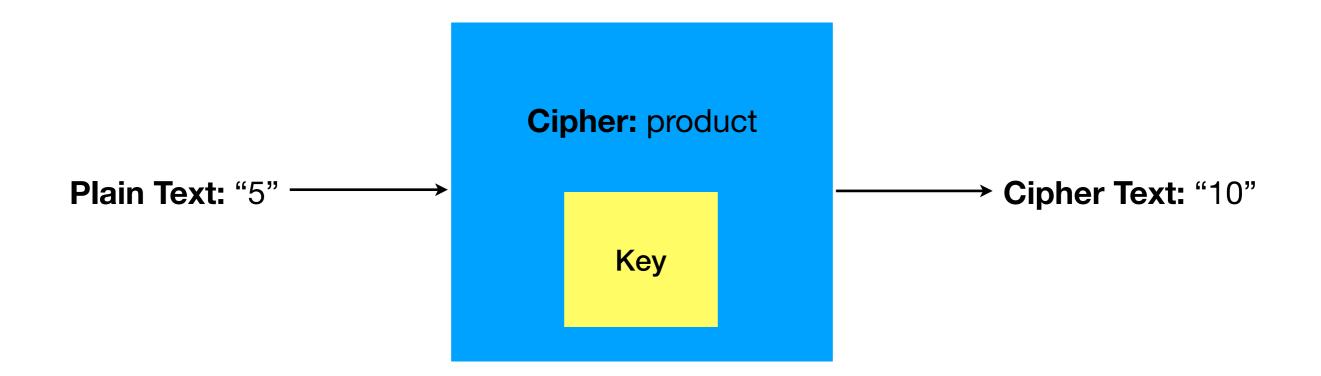


Definitions

- **Encryption**: transforming intelligible numbers or text, sounds and images into a stream of nonsense
- Plain text: the original text
- Cipher text: the encrypted text
- · Ciphers: the algorithms that perform the encryption and decryption
- Keys: a piece of auxiliary information that is used by the cipher to encode/decode the text, usually a cipher contains one or two keys
- Decryption: the process of taking encoded or encrypted text or other data and converting it back into text that you or the computer can read and understand.

Key vs. Cipher

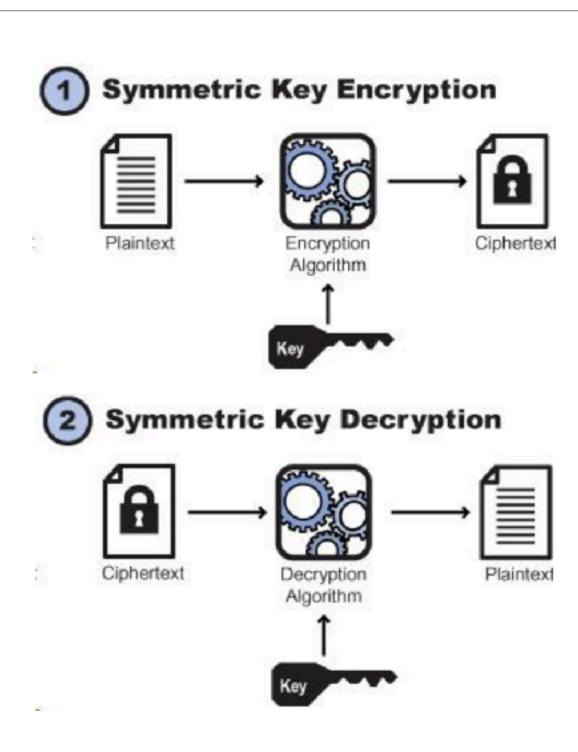




Definitions

Secret-Key Algorithm

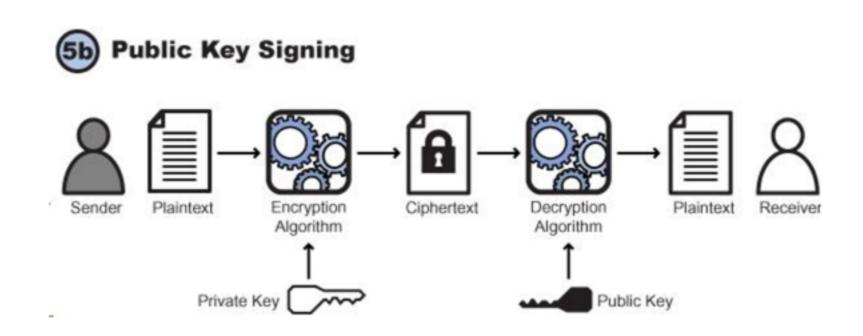
In symmetric algorithms, the sender uses a shared secret key to encrypt the message. The recipient must then use the same shared secret key to decrypt the message. Symmetric cryptography is very fast, but requires every possible pair of users to have a unique key. In a large system, the number of keys required can very quickly become unmanageable.



Definitions

Public-Key Algorithm

In asymmetric algorithms (also known a public-key cryptography), the sender and recipient
use related but different keys to encrypt and decrypt the message. Each participant has a
pair of keys (one public and one private), and everyone has access to everyone else's public
key. When a sender wants to encrypt a message, he or she does so using the recipient's
public key. The recipient can then decrypt that message using his or her private key.
 Asymmetric cryptography is slower than symmetric cryptography, but key management is
greatly simplified.



Ciphers

- Simple Substitution Ciphers: can be demonstrated by writing the alphabet in some order to represent the substitution
 - 25! = 1.44e+25
- Homophonic Substitution Cipher: in an attempt to increase the difficulty of deciphering substitution ciphers, in homophonic ciphers, plaintext letters map to more than one ciphertext symbol
- Transposition Cipher: the units of plaintext are rearranged in a different and quite complex manner.

Substitution Cipher

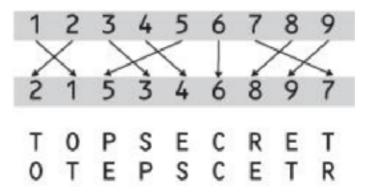
ABCDEFGHIJKLMNOPQRSTUVWXYZ

QWERTYUIOPASDFGHJKLZXCVBNM

GRAY FOX HAS ARRIVED
UKQN YGB IQL QKKOCTR

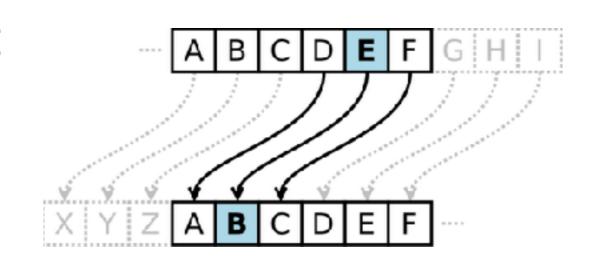


Transposition Cipher



Ciphers - Examples

- Caesar Cipher: shifted alphabet
 - · Example: a left shift of three



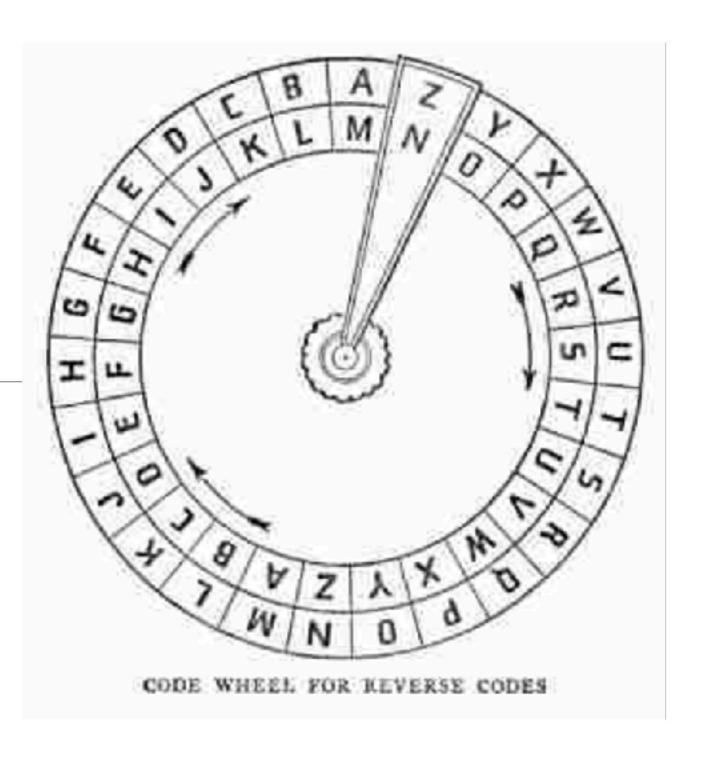
Plain: ABCDEFGHIJKLMNOPQRSTUVWXYZ Cipher: XYZABCDEFGHIJKLMNOPQRSTUVW -

Plaintext: THE QUICK BROWN FOX JUMPS OVER THE LAZY DOG Ciphertext: QEB NRFZH YOLTK CLU GRJMP LSBO QEB IXWV ALD

Albasıı Cipiler. reversed aipilabet

Plain	Α	В	С	D	Е	F	G	Н	1	J	K	L	М	N	0	Р	Q	R	s	Т	U	٧	W	X	Υ	Z
Cipher	Z	Υ	X	W	٧	U	Т	s	R	Q	Р	0	N	М	L	K	J	T	Н	G	F	Е	D	С	В	Α

Decryption





Deciphering Steps

- 1. Create a Transition Probability Matrix on Plain Text
- 2. Define the Log-Likelihood loss function
- 3. Decipher using the Metropolis Algorithm

Step 1 - Transition Probability Matrix

Definition:

 A square matrix used to describe the transitions of a Markov chain

In our case:

- Which letter is likely to follow which letter in English
- 26x26 (25 English Letter, '_' for non-letter(?!+-))

• Training:

Very long English Text (War and Peace by Tolstoy)

Step 1 - Transition Probability Matrix - Example

English Beta:

 Imagine we have a simplified version of the English alphabet consisting of just 4 letters: d, o, g, and f

• Example phrase in English Beta:

'Good Dog of God.'

Step 1 - Transition Probability Matrix - Example

- Constructing Transition Matrix for the phrase:
 - 'Good Dog of God.'
- Loop through each two consecutive characters: 'g'<'o', 'o'<'o', 'o'<'o', 'd'<'', 'd'<'o', 'o'<'g', 'g'<' ', ''<'o', 'o'<'f', 'f'<' ', ''<'g', 'g'<'o', 'o'<'d', 'd'<'.'

	d	О	g	f	_	row_sum
d	0	1	0	0	1 + 1 = 2	3
o	1 + 1 = 2	1	1	1	0	5
g	0	1 + 1 = 2	0	0	1	3
f	0	0	0	0	1	1
_	1	1	1	0	0	3

Step 1 - Transition Probability Matrix - Example

- Normalize by Row Sum to make probability
- Example:
 - 'o' is likely to be followed by letter 'd' (40%)
- 5 times 'o' occurred in ex. 2 times 'o' followed by 'd'

row_sum	_		g	0	d	
3	0.66667	0	0	0.33333	0	d
5	0	0.2	0.2	0.2	0.4	0
3	0.33333	0	0	0.66667	0	g
1	1	0	0	0	0	f
3	0	0	0.3333	0.33333	0.33333	_
	0.33333 1 0	0	0.3333	0	0.33333	f _

Step 2 - Defining Log Likelihood

- Quantity of a text making sense in English
- If a text makes, each consecutive 2 letters of the word should correspond to the transition prob matrix

Log-like =
$$\sum_{i=first\ letter}^{last\ letter} \log P(i, i_next_letter)$$

 Where i and j correspond to the labels of a row and column of Transition Probability Matrix, P.

Step 2 - Defining Log Likelihood - Example

- loglikehood(good) =
 - log(P(g,o)) + log(P(o,o)) + log(P(o,d))
 - $\log(0.66) + \log(0.2) + \log(0.4) = -2.94$
- loglikehood(og+g) = ?

	d	0	g	f	_	row_sum
d	0	0.33333	0	0	0.66667	3
0	0.4	0.2	0.2	0.2	0	5
g	0	0.66667	0	0	0.33333	3
f	0	0	0	0	1	1
_	0.33333	0.33333	0.3333	0	0	3

Step 2 - Defining Log Likelihood - Example

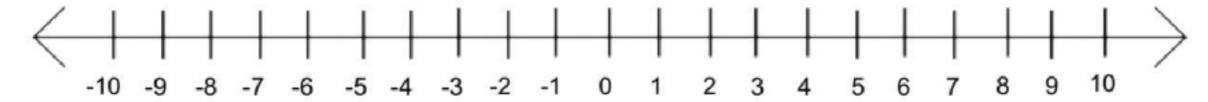
- loglikehood(good) =
 - log(P(g,o)) + log(P(o,o)) + log(P(o,d))
 - $\log(0.66) + \log(0.2) + \log(0.4) = -2.94$
- loglikehood(og+g) =
 - $\log(0.2) + \log(0.333) + \log(0.333) = -3.81$

	d	0	g	f	_	row_sum
d	0	0.33333	0	0	0.66667	3
o	0.4	0.2	0.2	0.2	0	5
g	0	0.66667	0	0	0.33333	3
f	0	0	0	0	1	1
_	0.33333	0.33333	0.3333	0	0	3

Step 2 - Defining Log Likelihood

Less sense





fi yuo cna raed tihs, yuo hvae a sgtrane mnid too. Cna yuo raed tihs? Olny smoe plepoe can. i canualt blueiee taht I cluad aulacity uesdnathrd waht I was rdanieg. The phaonmneal pweor of the hmuan mnid, accodining to a rischeearch at Cmabrigde Uinervtisy, it dseno't mtaetr in waht oerdr the ltteres in a wrod are, the olay iproamtnt tihng is taht the frsit and Isat Itteer be in the righit polae. The riset can be a taotl mises and you can sitll raed it whotuit a phoerlm. Tihs is bouseae the huamn mniddeos not raed ervey Iteter by istlef, but the wrod as a wlohe. Azanmig huh? yaeh and I awlyas tghuhot slpeling was ipmorantt! if you can raed tihs sahre it.



Aside: If you ask a Mathematician?

- Treat like Global Optimization Problem
 - Define Objective Function (Log-likelihood)
 - 2. Find a mapping (deciphering) to maximize objective fun
 - 3. **Assumptions:**
 - continuous function,
 - finite decision vectors
 - decision vectors are non-empty

Those assumptions are not always practical....

Step 3 - Metropolis Algorithm - Assumptions

- Proposal Distribution is symmetric.
- In our case, satisfied (explain later).

Step 3 - Metropolis Algorithm - Warm Up

- Probabilistic Model, random number generation
- Example Encoding: 'a' > 'a', 'b' > 'b', 'd' > 'd' ... 'z' > 'z'
- 'bad' > ?

- Scrambled letters 'a' > 'b', 'b' > 'a', 'd' > 'c', ... 'z' > 'z'
- 'bad' > ?
- Mess up letters > New mapping

Step 3 - Metropolis Algorithm

 Simply, using random number generation, start with a random scrambled mapping (current mapping)

Compute log-like of decoded text, current mapping (current log-like)

Randomly swap Two letters of current mapping to propose a new mapping

Compute log-like of proposed mapping

Step 3 - Metropolis Algorithm

 Probability to accept is Normalized log-like ratio of the new mapping and the current mapping

```
P(accept proposed) = exp(Log-like(proposed) - Log-like(current))
```

RNG to decide accept or not

If accepted, update current mapping with Proposed. Repeat

Step 3 - Metropolis Algorithm - Example

- True mapping (encoding): d>o, o>f,g>d, f>g
- 'dog' >> 'ofd' (encoded text to decipher)
- Initial mapping (random, 4 letters): d>g, o>d, g>f, f>o
- Decoded based on current mapping: 'fgo'
- log-like(current) = log(0.01) + log(0.667) = -5.01
- proposed mapping (g, d): d>d, o>g, g>f, f>o
- Decoded based on proposed mapping: 'fgd'
- log-like(proposed) = log(0.01) + log(0.01) = -9.21

Step 3 - Metropolis Algorithm - Example

Accept the proposed mapping with the probability

```
P(accept proposed) = exp(Log-like(proposed) - Log-like(current))
```

• P=exp(-9.21 - 5.01) = 0.0000007

- If proposed mapping gets accepted, update the current mapping with the proposed mapping
- Given enough time and bigger text, the current mapping will eventually converge into the true mapping (true encoding).

Fun Example

- 1. Go to: https://github.com/snooravi/meetups
- 2. Clone/download the directory to your local computer
- 3. Open jupyter notebook (run 'jupyter notebook' in CLI)
- 4. Navigate to: decryptation_example/ .ipynb

Further Reading:

- Introduction to Cryptology: <u>here</u>
- Decrypting Classical Cipher Text Using Markov Chain Monte Carlo: <u>here</u>
- Text Decryption Using MCMC: <u>here</u>
- · Substitution Ciphers: here