An evolutionary computation attack on one-round TEA

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Agenda

Problem Overview

- Tiny Encryption Algorithm
- Evolutionary Computation
- Experimental Design
- Results
- Conclusions

Problem Overview

 In this project we attack <u>one-round</u> TEA with evolutionary computation.

Problem Overview



TEA is an elegant, light weight Feistel block cipher. (Wheeler & Needham, 1994)



We use an evolutionary computation algorithm to derive the encryption key.

Problem Overview

• This is called a *known-plaintext* attack.

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• TEA is a Feistel block cipher.

- <u>rounds</u> of operation on the <u>bits</u> of plaintext
- operations are <u>substitutions</u>, <u>permutations</u>
- want: unique sequence of operations for each key



The key is 128-bits long (4 × 32-bit words) Text is operated in blocks of 64-bits (2 × 32-bit words)



delta is a constant sum is as a key scheduler



round one



round two

...





round one



round 32

...





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EC Evolutionary Computation

a whole family of optimization techniques
all somewhat inspired by biological evolution
we combine two techniques in this project

genetic algorithm and harmony search

GA Genetic Algorithm

search space represented by <u>chromosomes</u>
optimize a <u>population</u> of chromosomes
called *evolution*

• Let's discuss this in context of this project.

(Fraser, 1957)

GA Genetic Algorithm • A Population



A population of seventy chromosomes (keys).

GA Genetic Algorithm

operations performed on population

get better and better keys at each generation

GA Genetic Algorithm • Point Mutation



Operation 1: We define a point mutation as a bit-flip. (2% uniform pseudorandom probability for each bit) 23

Genetic Algorithm • Crossover



Operation 2: Two parents yield two children in crossover. (pairs of parents are randomly selected)

GA Genetic Algorithm • Fitness Function

select the fittest keys to continue

- unfit keys removed from population
- want: fitness should improve over time
- our fitness function is a hamming distance

GA Genetic Algorithm • Fitness Function



Hamming distance calculated for each key's ciphertext against the correct key's ciphertext.

Harmony Search

• Let's discuss this as a modification to GA.

two more operators

Harmony Search is inspired by *musical improvisation* (Geem & Kim, 2001)

HS Harmony Search • Improvise Harmony



Operation 3: Entire population used to improvise harmony. (contributing parents randomly selected) 28

Harmony Search • Adjust Pitch



Operation 4: We define adjust pitch as byte or word swaps. (pairs of bytes or words randomly selected within a key) 29

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Experimental Design

• We will now explain ...

- the arrangement of *operators* in the EC
- the *selection* of chromosomes in the EC
- selection of known plaintexts and keys

EC Operators & Selection

Operating a population of keys in a generation ...



Each box is a reservoir of 70 keys. After all operations are complete, the best ten from each box are advanced to next generation. 32

Let's talk about how we selected plaintexts

Plaintexts and Keys

- Plaintexts were chosen at random with a pseudorandom uniform number generator.
- Ciphertexts were calculated using one-round TEA.
- 100 plaintext message blocks per trial
- 30 trials per experiment

Plaintexts and Keys

two experiments run given two keyschemes
 scheme 1: create keys with the words ...
 {0x00000000, 0xFFFFFFF, 0xXXXXXXX}⁺

scheme 2: create keys with the words ...
 {0xFF000000, 0x00FF0000, 0x0000FF00, 0x00000FF}

[†]0xXXXXXXXXX is four random bytes.

Plaintexts and Keys

• Number of keys in keyscheme 1:

- 3⁴ = 81 schemes
- Number of keys in keyscheme 2:
 - 4⁴ = 256 schemes
- Every single scheme was committed to 30 trials of randomly generated plaintexts.

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Results Experiment 1

{0x<u>00000000</u>, 0x<u>FFFFFFF</u>, 0x<u>XXXXXXX</u>}

Results Experiment 1 - Summary Plots



Proportion of Convergences

- really easy:
 - >2 zeros
- easy:
 - (KO, K1) = 0 or F
 - (K2, K3) = 0 or F

• hard:

• >1 X (random)

Results

Experiment 1 - Number of Convergences

Word	Occurrences	Affected keys	<i>x</i> / 30	σ
0	1	32	2.9	3.9
	2	24	5.0	5.1
	3	8	13.1	1.5
	4	1	24.0	_
F	1	32	2.8	4.4
	2	24	5.6	5.0
	3	8	9.0	3.3
	4	1	14.0	_
X	1	32	5.1	4.0
	2	24	0.9	1.3
	3	8	0.0	0.0
	4	1	0.0	_

Average number of convergences over 30 trials (with standard deviation)

Results Experiment 1 - Summary Plots



Speed of Convergence

- larger dot is earlier
- cross = no convergence
- same pattern as before

Results

Experiment 1 - Convergence Generation

Word	Occurrences	Converged trials	\bar{x} generation	σ
0	1	92 / 960	2069.6	1519.1
	2	121 / 720	2192.8	1399.7
	3	105 / 240	1699.0	1311.8
	4	24 / 30	801.0	1112.8
F	1	89 / 960	2118.6	1504.9
	2	135 / 720	2220.5	1370.9
	3	72 / 240	1506.2	1334.5
	4	14 / 30	1216.7	1357.5
X	1	164 / 960	2241.4	1435.5
	2	22 / 720	2931.8	1466.5

Average generation of convergence over 30 trials (with standard deviation)

Results Experiment 1 - Summary Plots



Proportion of Degenerate Keys

- TEA degenerate keys
- each key is part of an equivalent triplet
- (KO, K1) = 0 or F
- (K2, K3) = 0 or F
- large number of random breaks: {(000X), (FXFF), (FFFX), (XXFF)}

Results Experiment 2

{0xFF000000, 0x00FF0000, 0x0000FF00, 0x00000FF}

Results

Experiment 2 - Number of Convergences

Arrangement of matching words	Affected keys	<i>x</i> / 30	σ
$K0 = K1 \land K2 = K3$	16	7.0	4.4
$K0 = K1 \land K2 \neq K3$	48	6.9	3.7
$K0 \neq K1 \wedge K2 = K3$	48	6.9	4.1
$K0 \neq K1 \land K2 \neq K3$	144	7.0	3.2
$K0 = K2 \wedge K1 = K3$	16	10.8	2.8
$K0 = K2 \wedge K1 \neq K3$	48	8.3	2.8
$K0 \neq K2 \wedge K1 = K3$	48	8.3	3.6
$K0 \neq K2 \wedge K1 \neq K3$	144	5.7	3.2
exactly four matching words	4	13.0	3.4
exactly three matching words	48	9.8	3.2
exactly two matching words	180	6.5	3.1
no matching words	24	3.5	1.7

Average number of convergences over 30 trials (with standard deviation)

Results Experiment 2 - Convergence Generation

Word	Converged trials	\bar{x} generation	σ
exactly four matching words	52 / 120	1779.7	1352.0
exactly three matching words	471 / 1440	2196.9	1348.2
exactly two matching words	1175 / 5400	2408.9	1291.0
no matching words	85 / 720	2776.8	1207.8

Average generation of convergence over 30 trials (with standard deviation)

Results Experiment 2 - Summary Plots



Proportion of Degenerate Keys

• large degenerate areas

- (KO, K1) = FF00000
- (K2, K3) = FF00000

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Conclusions Experiment 1 {0x0000000, 0xFFFFFFF, 0xXXXXXX}

more random words implies more resilience

- all-on-bit words more resilient to all-off-bit
- EC method can find equivalent keys

Conclusions

Experiment 2 {0xFF000000, 0x00FF0000, 0x0000FF00, 0x00000FF}

matched words in (K0, K2) and (K1, K3) easier

more matched words even easier (due to HS)

 equivalent keys easiest to derive for words with {0xFF000000} in (K0, K1) or (K2, K3).

Conclusions

- EC methods capable of attacking <u>one-round</u>
 Feistel block ciphers.
- HS is particularly good at probing solutions with repetitions.
- Equivalent keys can also be derived.

Thanks for listening! http://eddiema.ca ema@uoguelph.ca

Results Experiment 2 - Summary Plots



Proportion of Convergences

• no obvious pattern

Results Experiment 2 - Summary Plots



• • • •

• no obvious pattern

Speed of Convergence