# BILL <br> RECORD THIS LECTURE 

September 16, 2020

## Revisit GCD and Math Notation

September 16, 2020

## Revisit GCD Briefly

Two things about GCD I want to clarify.

- Why is $\operatorname{GCD}(x, 0)=x$ for $x \geq 1$ ?
- When does the algorithm stop?


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To make our formula $\operatorname{GCD}(x, y)=\operatorname{GCD}(x-k y, x)$ work all the way to 0 , we define $\operatorname{GCD}(0, x)=x$.

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Need to prove that all choices of sequences yield the same result. We won't do that here

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# Gen Sub Cipher: How to Really Crack 

September 16, 2020

## General Substitution Cipher

Def Gen Sub Cipher with perm $f$ on $\{0, \ldots, 25\}$.

1. Encrypt via $x \rightarrow f(x)$.
2. Decrypt via $x \rightarrow f^{-1}(x)$.

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4. One usually talks about the freq of $n$-grams.

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Lets try this with gen sub, ignoring the issue of 26 ! perms.
To crack gen sub shift went through all 26! perm $\sigma$ :

1. If $f_{\sigma(T)} \cdot f_{E}$ is large then $\sigma$ is correct perm. Large $\sim 0.065$.
2. If $f_{\sigma(T)} \cdot f_{E}$ is small then $\sigma$ is incorrect perm. Small. Hmmm?
3. We have a problem. If $\sigma$ only changed a few letters around, then likely $f_{E} \cdot f_{\sigma(T)}$ will be large. We do not have a gap!
What to do?

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1. Use $n$-grams instead of 1 -grams. This does not close the Gap but will help anyway.
2. Rather than view the Is-English program as a YES-NO, view it as comparative:
$T_{1}$ looks more like English than $T_{2}$.

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Pick the $\sigma_{r}$ with $\min$ good $_{r}$ or have human look at all $\sigma_{r}(T)$

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An old question:
What came first, the chicken or the egg?

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Same here.
We find the parameters for texts where we know the answers.

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5. Keep track of how how many iterations suffice and how many redos suffice.

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For each text he generated 1 random perm (will rerun with more later).

## Parameters for 1-Grams

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Nothing worked.

## Parameters for 2-Grams

Nothing worked.

Parameters for 3-Grams

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1. The average time to get within 1-2 swaps was 1 minute.

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3. Seems hard to get those 1-2 swaps right.
4. The average number of iterations was 900 . The MAX number of iterations was 1902. TAKE $\mathrm{I}=2000$.
5. The average number of redos the program needed to get within 2 swaps was 1.14. The max number of times was 3 . TAKE $\mathrm{R}=4$.

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## 4-Grams

1. The average time to get it perfect was 6 minutes.
2. The min time was 4 minutes, the max time was 30 minutes.
3. The average number of iterations was 1000. The MAX number of iterations was 1966. TAKE $\mathrm{I}=2000$.
4. The average number of REDOS to get it perfects was 1.3. The max number of times was 7 . TAKE $\mathrm{R}=8$

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6. Zan found that our algorithm was already known, which did not surprise me. We discuss this on the next slide.
7. Does ML really help crypto? Not sure.

## Our Algorithm Already Known

A Fast Method for Cryptanalysis of Substitution Ciphers by Jakobsen, (1995)
has our approach with the following caveats:

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3. Since they can use bigrams rather than trigrams (I assume) their algorithm is faster.
4. So why did I present ours? (1) Educationally mine and theirs are the same, and (2) I knew all of the parameters of our algorithm and how we got them.

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Since the text was in blocks of five and we want to totally mechanize, need a method to find word breaks.

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We leave this topic for now.

# BILL <br> STOP RECORDING THIS LECTURE 

September 16, 2020

