



# A Theoretic Framework For Evaluating Similarity Digesting Tools

*By*

**Liwei Ren**

*Presented At*

The Digital Forensic Research Conference

**DFRWS 2015 EU** Dublin, Ireland (Mar 23<sup>rd</sup>- 26<sup>th</sup>)

DFRWS is dedicated to the sharing of knowledge and ideas about digital forensics research. Ever since it organized the first open workshop devoted to digital forensics in 2001, DFRWS continues to bring academics and practitioners together in an informal environment. As a non-profit, volunteer organization, DFRWS sponsors technical working groups, annual conferences and challenges to help drive the direction of research and development.

**<http://dfrws.org>**



# A Theoretic Framework for Evaluating Similarity Digesting Tools

---

**LIWEI REN, PH.D.** Trend Micro™

DFRWS EU 2015, Dublin, Ireland, March, 2015

# Agenda

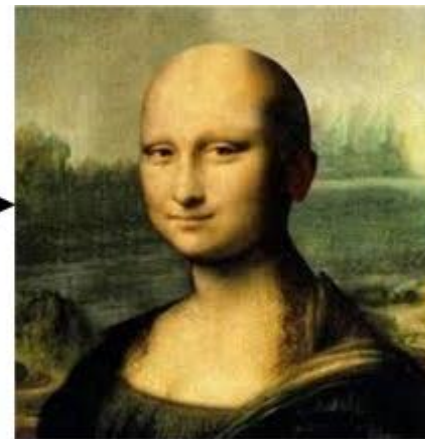
- Byte-wise Approximate Matching
- Similarity Digesting Tools
- Mathematical Models for Byte-wise Similarity
- Tool Evaluation with Theoretic Analysis
- Tool Evaluation with Data Experiment
- Further Research for Approximate Matching

# Byte-wise Approximate Matching

- Byte-wise similarity & approximate matching.
  - What is byte-wise similarity ?
- 4 Use Cases specified by *NIST*:



← (1) Similarity →



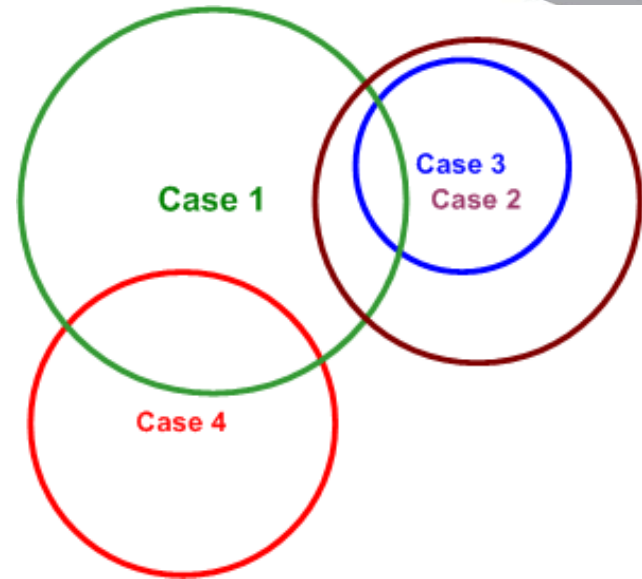
← (2) Cross sharing →



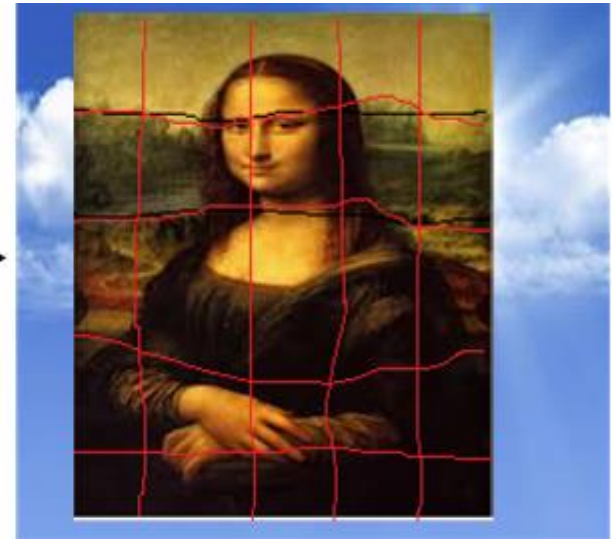
# Byte-wise Approximate Matching



← (3) Containment →



← (4) Fragmentation →



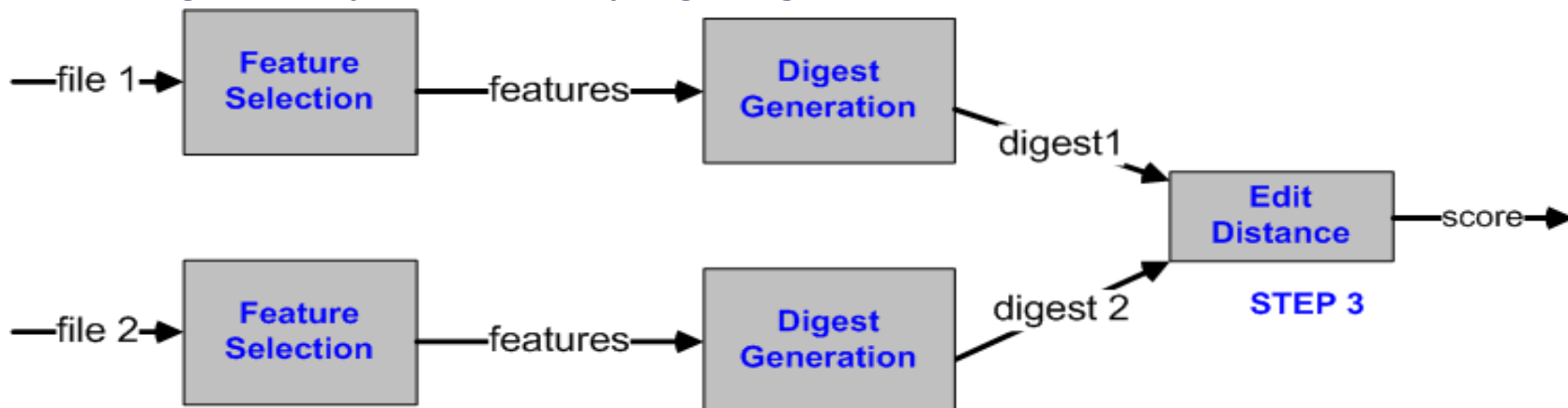
# Similarity Digesting Tools

- **Similarity digesting:**

- A class of hash techniques or tools that preserve similarity.
- Typical steps for digest generation:



- Detecting similarity with similarity digesting:



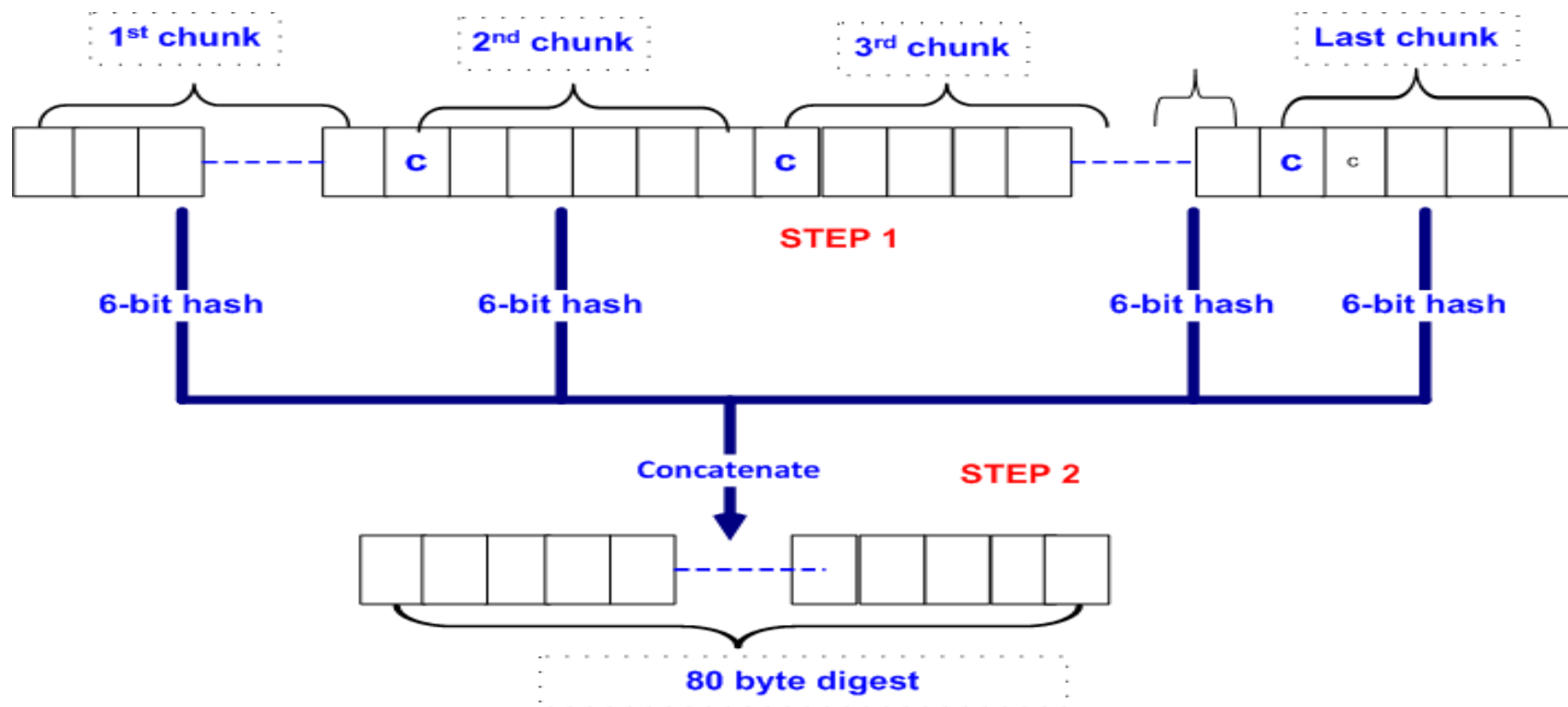
- **Three similarity digesting algorithms and tools:**

- ssdeep, sdhash & TLSH

# Similarity Digesting Tools

- ssdeep

- Two steps for digesting:

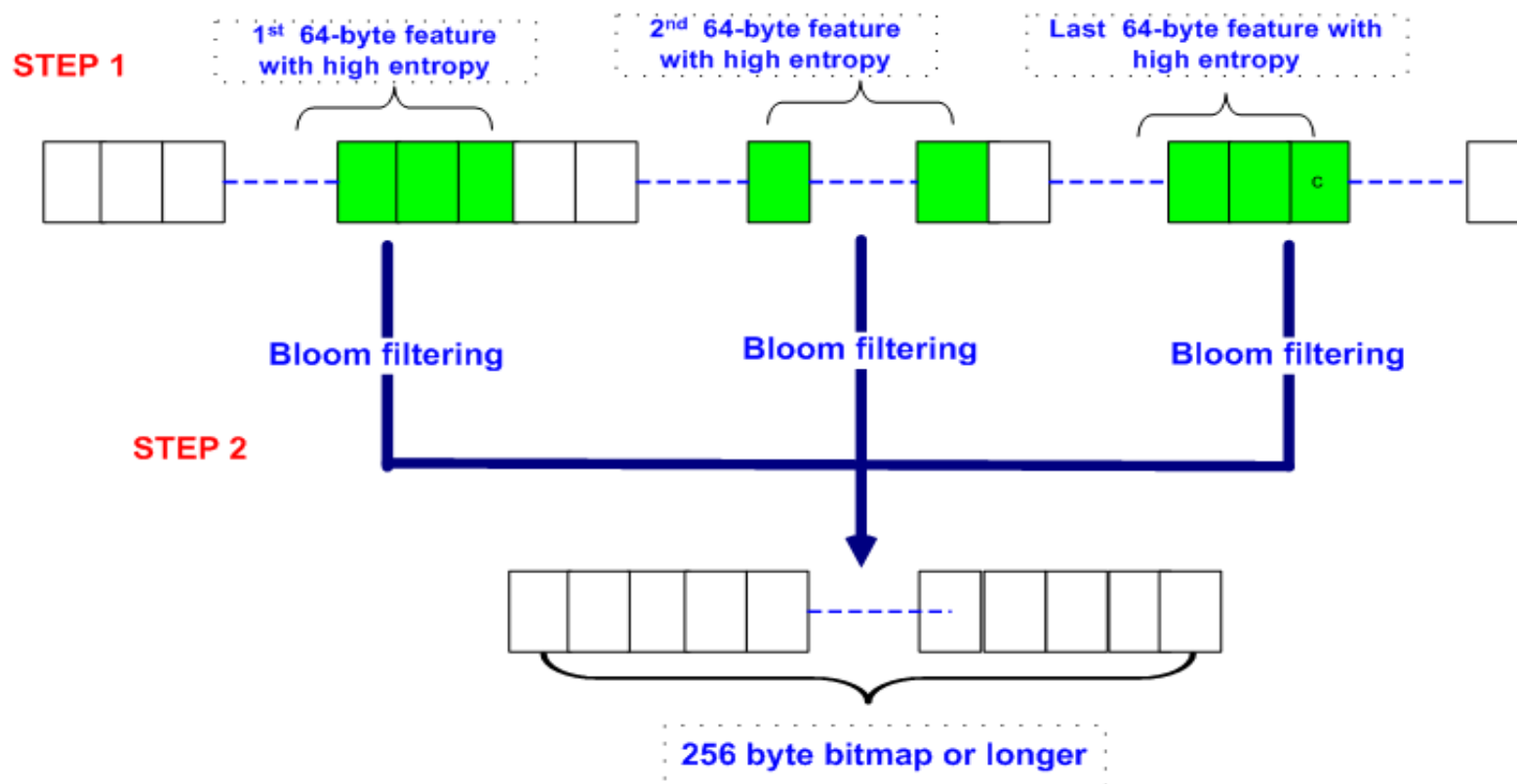


- Edit Distance: Levenshtein distance

# Similarity Digesting Tools

- sdhash

- Two steps for digesting:



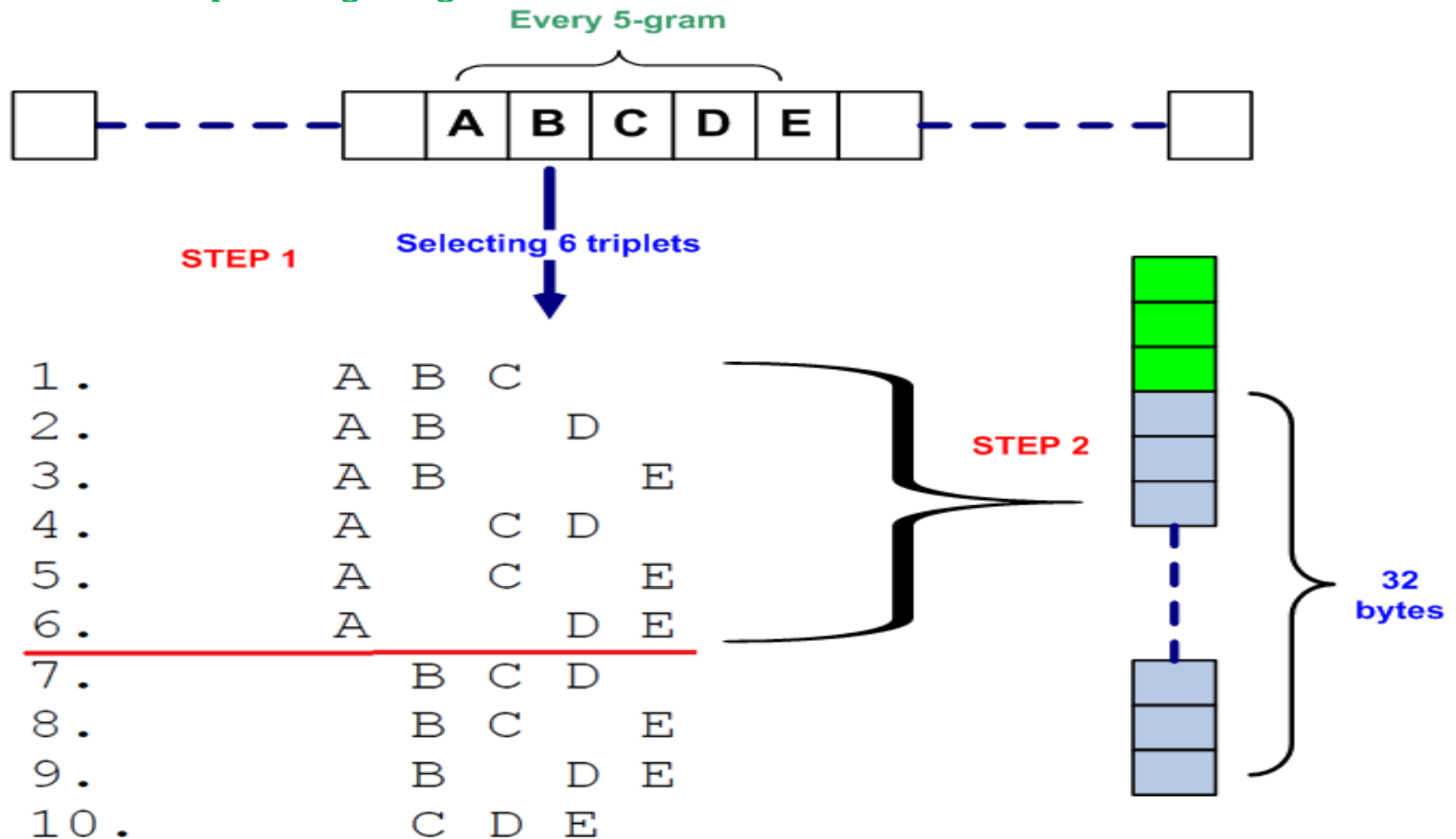
- **Edit Distance:** Hamming distance



# Similarity Digesting Tools

- TLSH

- Two steps for digesting :

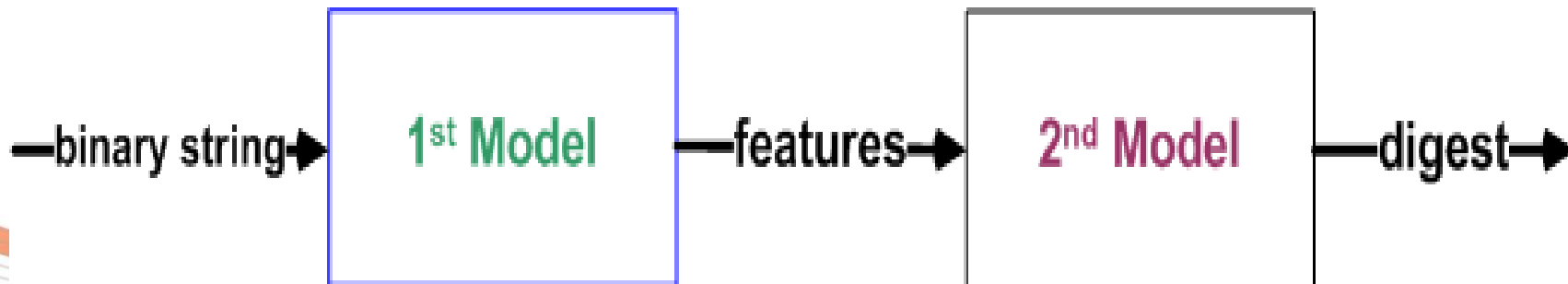


- **Edit Distance:** A diff based evaluation function

# Mathematical Models for Byte-wise Similarity

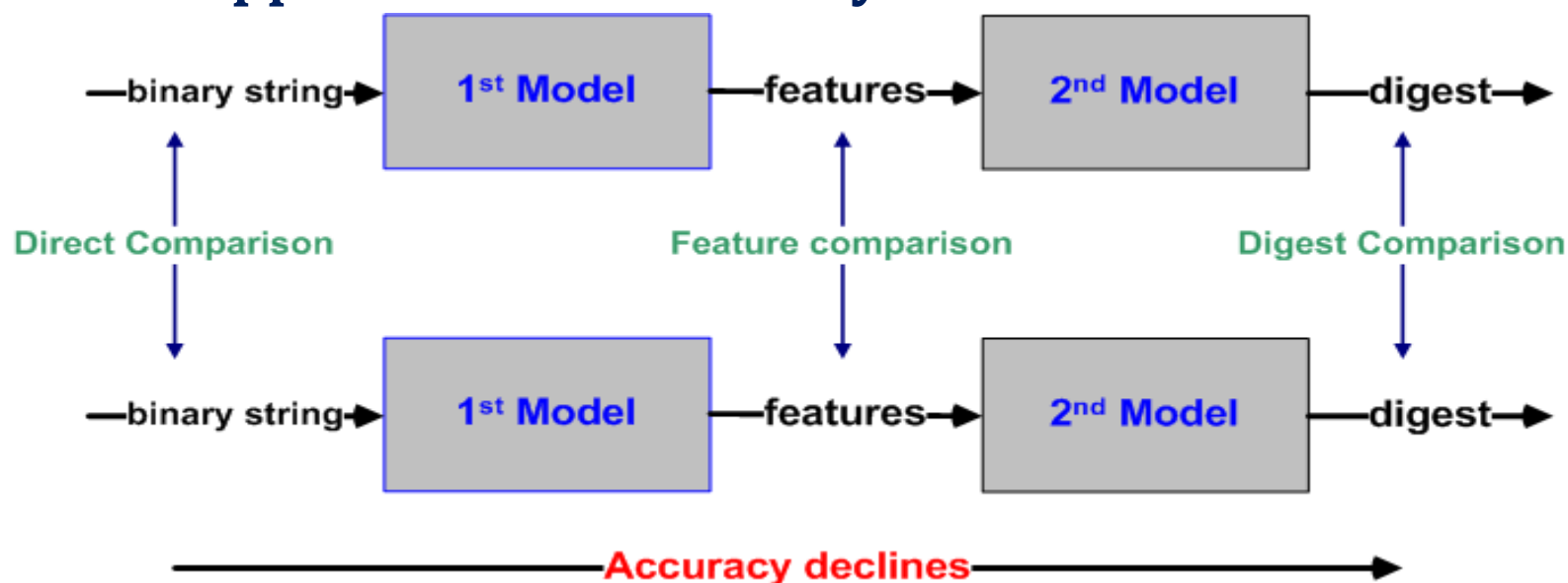
- **Summary of Three Similarity Digesting Schemes:**

- Using a **first model** to describe a binary string with *selected features*:
  - **ssdeep model**: a string is a *sequence* of *chunks* (split from the string).
  - **sdhash model**: a string is a *bag* of *64-byte blocks* (selected with entropy values).
  - **TLSH model**: a string is a *bag* of *triplets* (selected from all 5-grams).
- Using a **second model** to map the selected features into a digest which is able to preserve similarity to certain degree.
  - **ssdeep model**: a sequence of chunks is mapped into a 80-byte digest.
  - **sdhash model**: a bag of blocks is mapped into one or multiple 256-byte bloom filter bitmaps.
  - **TLSH model**: a bag of triplets is mapped into a 32-byte container.



# Mathematical Models for Byte-wise Similarity

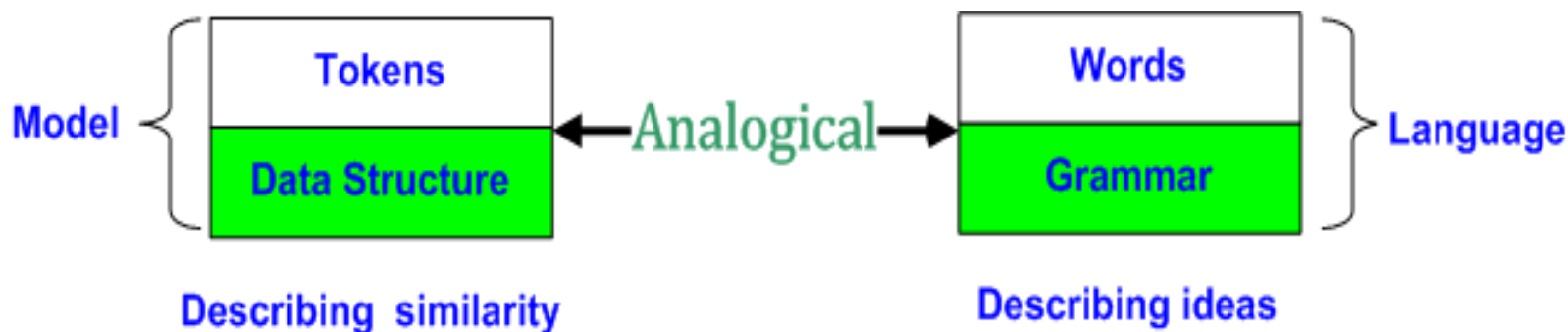
- Three approaches for similarity evaluation:



- 1<sup>st</sup> model plays critical role for similarity comparison.
  - Let focus on discussing various 1<sup>st</sup> models today.
    - Based on a unified format.
- 2<sup>nd</sup> model saves space but further reduces accuracy.

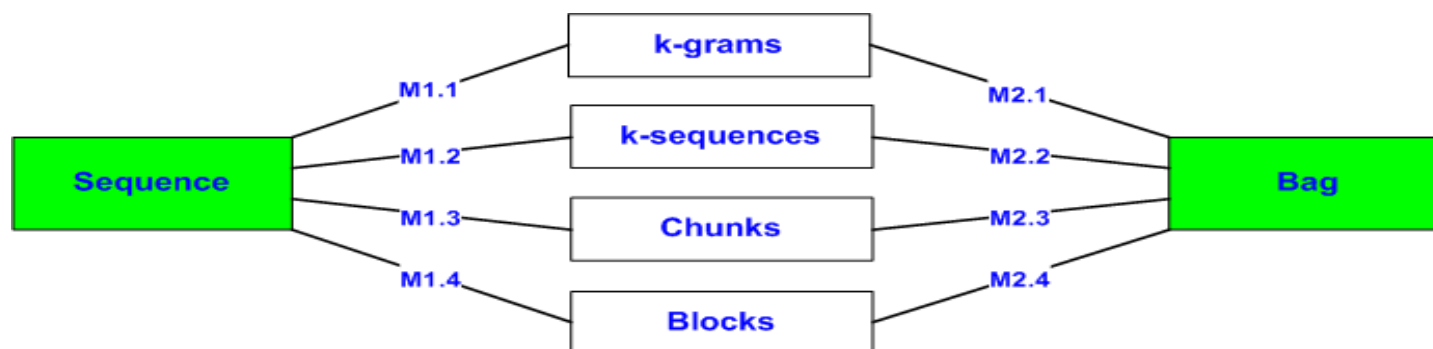
# Mathematical Models for Byte-wise Similarity

- Unified format for 1<sup>st</sup> model:
  - A string is described as a collection of tokens (aka, features) organized by a data structure:
    - ssdeep: a sequence of chunks.
    - sdhash: a bag of 64-byte blocks with high entropy values.
    - TLSH: a bag of selected triplets.
  - Two types of data structures: sequence, bag.
  - Three types of tokens: chunks, blocks, triplets.
- Analogical comparison:



# Mathematical Models for Byte-wise Similarity

- Four general types of tokens from binary strings:
  - k-grams where k is as small as 3,4,...
  - k-subsequences: any subsequence with length k. The triplet in TLSH is an example.
  - Chunks: whole string is split into non-overlapping chunks.
  - Blocks: selected substrings of fixed length.
- Eight different models to describe a string for similarity.



- Analogical thinking:
  - we define different distances to describe a metric space.

# Tool Evaluation with Theoretic Analysis

- **Data Structure:**

- **Bag:** a bag ignores the order of tokens. It is *good at handling content swapping*.
- **Sequence:** a sequence organizes tokens in an order. This is *weak for handling content swapping*.

- **Tokens:**

- **k-grams:** Due to the small  $k$  ( 3,4,5,...), this fine granularity is *good at handling fragmentation*.
- **k-sequences:** Due to the small  $k$  ( 3,4,5,...), this fine granularity is *good at handling fragmentation*.
- **Chunks:** This approach takes account of every byte in raw granularity. It should be *OK at handling containment and cross sharing*
- **Blocks:** Depending on different selection functions, even though it does not take account of every byte, but it may present a string more efficiently and that is good for generating similarity digests. Due to the nature of fixed length blocks, it is *good at handling containment and cross sharing*.



# Tool Evaluation with Theoretic Analysis

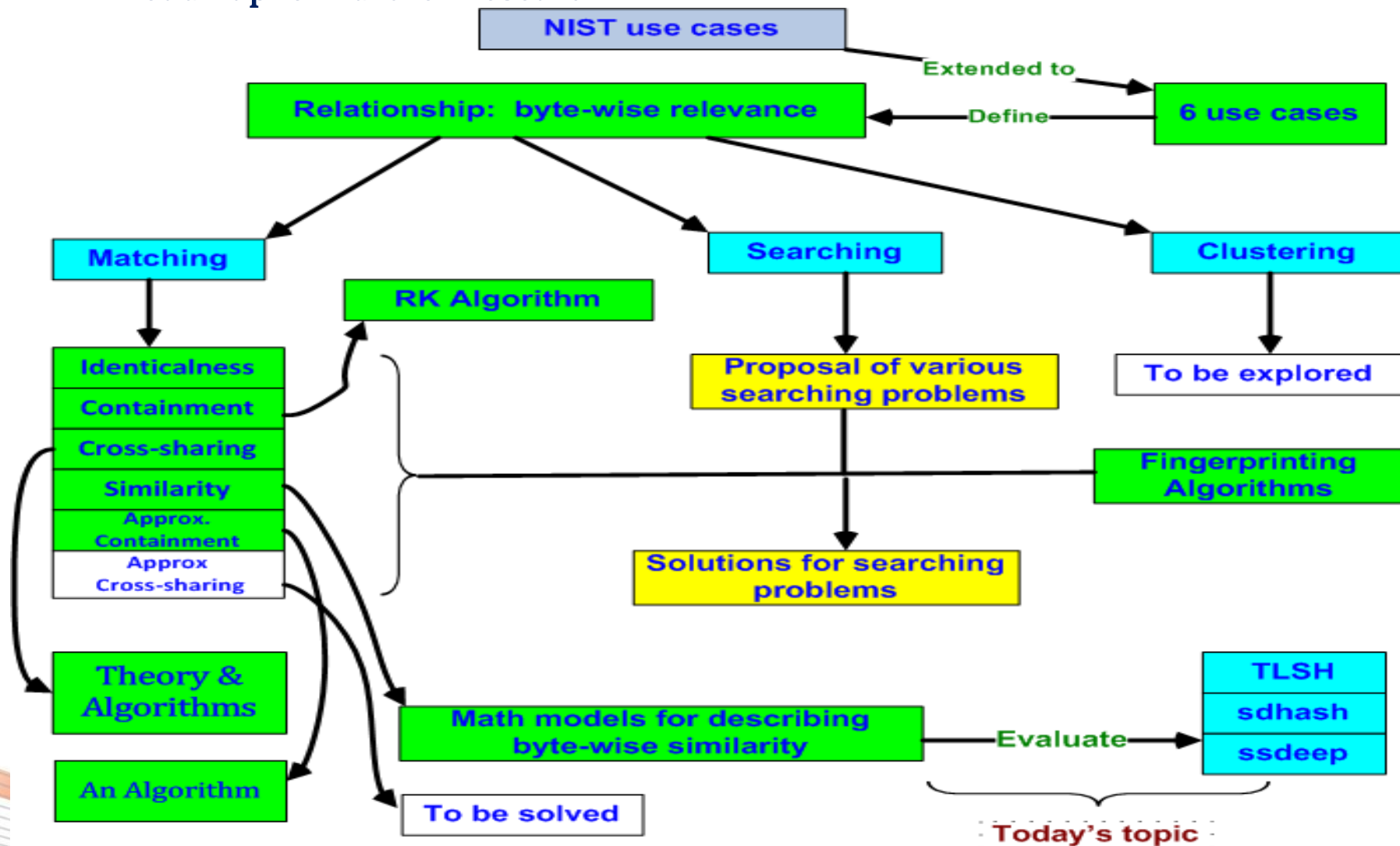
Tool	Model	Minor Changes	Containment	Cross sharing	Swap	Fragmentation
ssdeep	M1.3	<i>High</i>	<i>Medium</i>	<i>Medium</i>	<i>Medium</i>	<i>Low</i>
sdhash	M2.4	<i>High</i>	<i>High</i>	<i>High</i>	<i>High</i>	<i>Low</i>
TLSH	M2.2	<i>High</i>	<i>Low</i>	<i>Medium</i>	<i>High</i>	<i>High</i>
Sdhash + TLSH	Hybrid	<i>High</i>	<i>High</i>	<i>High</i>	<i>High</i>	<i>High</i>

# Tool Evaluation with Data Experiment

		Base File Size = 2MB			Base File Size = 64KB		
Purpose of Tests	Edit Operations	<u>ssdeep</u>	<u>sdhash</u>	TLSH	<u>ssdeep</u>	<u>sdhash</u>	TLSH
<b>Containment</b>	Cut 30% at the beginning	82	60	31	79	89	31
	Cut 60% at the end	54	100	X	58	99	X
	Cut 90% at the beginning	X	77	X	X	100	X
<b>Cross sharing</b>	Substitute 30% at the end	72	70	69	75	59	68
	Substitute 60% in the end	47	40	54	47	37	62
	Substitute 90% at the end	29	10	47	X	6	42
<b>Swap</b>	Swap with 2-1	52	71	99	54	68	98
	Swap with 4-3-2-1	36	59	98	33	54	98
	Swap with 8-7-6-5-4-3-2-1	32	62	99	X	48	96
<b>Fragmentation</b>	Modify the bytes at 64*j	X	X	58	X	X	78
	Modify the bytes at 128*j	X	X	78	X	X	83
	Modify the bytes at 256*j	X	15	86	X	33	82
<b>Minor changes</b>	Swap with 1-2-3-4-5-7-6-8-9-10-11-12-13-14-15-16. <u>Subst</u> 1% at the end. Cut 1% at the beginning.	90	88	93	83	93	84
	Swap with 1-2-3-5-4-6-7-8-9-10-11-12-13-14-15-16. Cut 2% in the beginning. <u>Subst</u> 1% at the end.	91	85	92	82	82	86

# Further Research for Approximate Matching

- A Roadmap for Further Research :



# Q&A

- Thank you for your interest.
- Any questions?
- My Contact Information:
  - Email: [liwei\\_ren@trendmicro.com](mailto:liwei_ren@trendmicro.com)
  - LinkedIn: <https://www.linkedin.com/in/drliweiren>
  - Academic Page: <https://pitt.academia.edu/LiweiRen>