



2008 - A66-811 - Secure Similar Document Detection - mmuruges@cs.purdue.edu - IAE

# 

the center for education and research in information assurance and security

# Secure Similar Document Detection

Wei Jiang, Mummoorthy Murugesan, Chris Clifton, Luo Si Computer Sciences, Purdue University

**Real World Scenario** 

document collection 1



document collection 2

Main Steps in SSDD

**Step 1: Find Common Vector Space** 



- $\succ$  Two conferences need to find the double submissions, or plagiarized papers.
- > Two companies want to find their common interests for a possible merger: e.g., # of similar or dissimilar projects > Parties don't want to exchange the actual documents for several reasons:
  - Documents are confidential
  - Not to lose the competitive edge
  - Due to laws, and regulations

# **Ideal Solution**



- Parties need to agree on a particular term vector space.  $T_1 \cap T_2 = \{ drugs, TB \}$
- Use a Secure Set Intersection (SSI) protocol to find the common vector space.

Only  $T_1 \cap T_2$  is revealed, neither  $T_1$  nor  $T_2$ .

# **Step 2: Create Normalized vectors**

• Each party computes the normalized term vectors of documents in their collections for  $T_1 \cap T_2$  $N_1' = (.5, 0.5), N_2' = (0.5, 0.5)$ 

# **Step 3: Compute Dot Product Securely**

- Use any one type of the Secure Dot Product (SDP) protocol: 1) Random Matrix (Vaidya & Clifton, 2002)
  - 2) Homomorphic encryption scheme (Goethals, et al., 2004)
- Compute pair-wise document cosine similarity using the SDP protocol on the normalized document vectors.

 $N_1' \cdot N_2' = 0.5$  (similarity score between Doc<sub>1</sub> and Doc<sub>2</sub>) • Actual values in the normalized vectors are not revealed.

- 1. Two parties send their document collections to a Trusted Third Party (TTP).
- 2. The TTP performs the similarity test on the document collections 3. The TTP sends back only the similarity scores to the parties.

#### **Do we really need a TTP?** No

# **Cosine Similarity Score**

- Let  $Doc_1 =$  "develop drugs for AIDS, and TB"
  - Doc<sub>2</sub> = "invent drugs for TB, and cancer"
  - Term vector space={develop, invent, drugs, AIDS, TB, cancer}
- Documents are represented as vectors of term frequencies
  - $D_1 = (1, 0, 1, 1, 1, 0)$  $D_2 = (0, 1, 1, 0, 1, 1)$

1. SSI - Common Vector space 2 3. SDP - Cosine Similarity ....

# **Experiments**

- > Tested on a document collection consisting of 100 papers from a major DB conference (# of terms is ~15k)
- > Tested on randomly generated vectors > The plot shows the running time of SSDD H
  - Number of documents vary from 100 to 500
  - Number of terms vary from 5K to 20K



• Normalize the vectors:  $N_i = D_i / |D_i|$  $|D_1|=2, |D_2|=2$  $N_1 = (.5, 0, .5, .5, .5, 0), N_2 = (0, .5, .5, 0, .5, .5)$ • Cosine Similarity Score  $(D_i, D_j) = D_i \cdot D_j / |D_i| |D_j| = N_i \cdot N_j$  $sim(D_1, D_2) = 0.5$ 

#### Conclusion

- Experimental results show that SSDD is practical for reasonably large collection of documents
- Can be used as a preliminary step to get candidates for more manual/comprehensive checks

# PURDUE UNIVERSITY



