### 11791: Milestone 1

Team 1:

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# Pipeline Structure

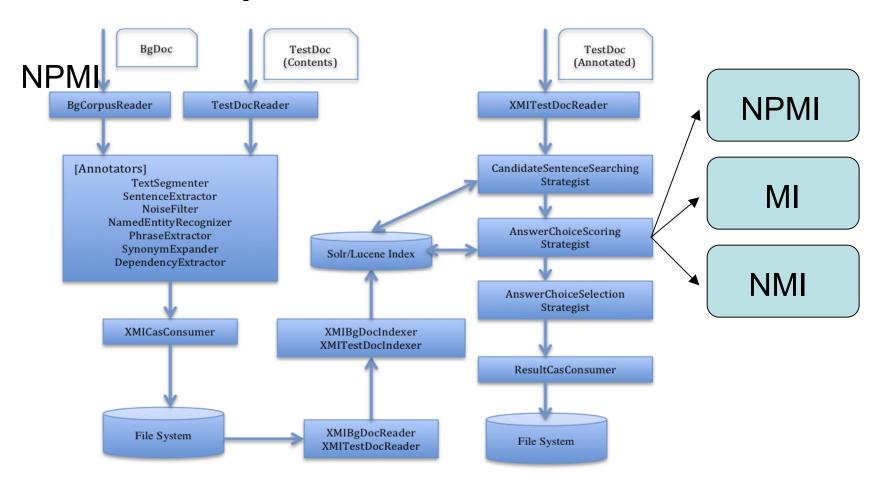


Figure 1. UIMA-based System Architecture for QA4MRE

## Type System

Maintaining the same Type System as given.

#### **Baseline Methods**

- PMI  $PMI(X,Y) = \ln \frac{p(x,y)}{p(x)p(y)}$ 
  - Captures correlation between x, y.
  - Becomes infinitely large when x, y are rare.
- NPMI

$$NPMI(X,Y) = rac{\ln rac{p(x,y)}{p(x)p(y)}}{-\ln p(x,y)}$$
 (Gerlof 2009)

- Reduce the impact of low frequency problem
- Has nice lower & upper bounds:
  - When two words only occur together: 1
  - When they only occur separately: -1
  - When they are independent: 0

#### **Baseline Methods**

$$NPMI = \frac{\log\left(\frac{P(x,y)}{P(x)P(y)}\right)}{-\log\left(P(x,y)\right)}$$

$$NPMI = \frac{\log(P(x,y)) - \log P(x) - \log P(y)}{-\log P(x,y)}$$

$$NPMI = \frac{\log C(x,y) - Z - \log C(x) + Z - \log C(y) + Z}{-\log C(x,y) + Z}$$

where  $Z = \log C(all\_doc)$ 

#### **Baseline Methods**

• MI

$$MI(X,Y) = \sum_{x,y} p(x,y) * PMI(x,y)$$

- Expected value of PMI
- Good indicator of whether two vectors are related or not (but not how much they're related)

• NMI 
$$NMI(X,Y) = \frac{\sum_{x,y} p(x,y) * PMI(x,y)}{-\sum_{x,y} p(x,y) \ln p(x,y)}$$

- Normalize by the joint entropy
- Behaves more like PMI or NPMI
- Recommended that MI and NMI be used together

#### Performance

- Original Baseline: P(xly) 0 2 5 3 3 (0.325)
- PMI:  $\ln(P(x,y)/(P(x)^*P(y))) 0 4 2 4 1 (0.275)$
- NPMI with Z: 0 4 1 2 2 (0.25) where Z = C(x)+C(y)-C(x,y)
- NPMI with Z^2: 0 2 0 1 1 (0.1)
- MI with Z: 0 4 4 2 3 (**0.325**)
- NMI with Z: 0 1 2 3 3 (0.225)
- $P(x,y)^2/(P(x)^*P(y)) 0 2 2 1 3 (0.1)$
- 0.5 MI + 0.5 P(xly): 0 4 3 1 2 (0.25)

#### **Future Plans**

- Use Machine Learning methods to estimate the best coefficients
  - E.g. SVM
  - Challenge: Not enough data??
- Other knowledge-based approaches:
  - Synonyms, Acronyms...
  - Expanding NPs and/or other POS tags

#### Reference

- Gerlof, B. (2009) Normalized (Pointwise) Mutual Information in Collocation Extraction. In: Chiarcos, Eckart de Castilho Stede (eds), Von der Form zur Bedeutung: Texte automatisch verarbeiten / From Form to Meaning: Processing Texts Automatically, Proceedings of the Biennial GSCL Conference 2009, pp3140, Tbingen, Gunter Narr Verlag.
- Evert, S. (2004/2005) The Statistics of Word Cooccurrences: Word Pairs and Collocations. PhD thesis, IMS Stuttgart.