

# Pattern Discovery for Wide-Window Open Information Extraction in Biomedical Literature

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# Outline

- Introduction
- Framework
- Evaluation
- Conclusion

# Characteristics of Language in Bio-Literature

- Long and complicated
  - Some sentences can be as long as a paragraph
  - Need parsing to understand the structure of the sentences
- Formal language
  - Good news for pattern extraction

# Example

- **Pre-treatment of ATRA can decrease the overexpression of cyclin\_D1 and E2F-1 induced by B(a)P.**

# Example

- Pre-treatment of [ATRA]<sub>CHEMICAL</sub> can decrease the overexpression of [cyclin\_D1]<sub>GENE</sub> and [E2F-1]<sub>GENE</sub> induced by [B(a)P]<sub>CHEMICAL</sub>.
- Task:
  - Find relationships among the entities

# Relation Extraction of Existing Studies

- Supervised methods
  - relying on annotated corpora to discover certain relation types between entities
- Distantly supervised methods
  - Using existing knowledge-bases or databases to annotate corpora
- $\langle \text{ATRA, cyclin\_D1, decrease?} \rangle \rightarrow \langle \text{ATRA, cyclin\_D1, decrease} \rangle$
- Limitations
  - Pre-defined relation types
  - Relation is pair-wise
  - The context is ignored

# Relation Extraction of Existing Studies

- OpenIE
  - Using linguistic features to discover all types of relations
- <Pre-treatment of [ATRA]<sub>CHEMICAL</sub>, **can decrease**, the overexpression of [cyclin\_D1]<sub>GENE</sub> and [E2F-1]<sub>GENE</sub> induced by [B(a)P]<sub>CHEMICAL</sub>>
- Pros
  - No pre-defined types
  - The context is kept
- Limitations
  - The extraction structure can be further improved

# How Human Structure the Information

- Pre-treatment of [ATRA]<sub>CHEMICAL</sub> **can decrease** the overexpression of [cyclin\_D1]<sub>GENE</sub> and [E2F-1]<sub>GENE</sub> induced by [B(a)P]<sub>CHEMICAL</sub>
- Pre-treatment of [ATRA]<sub>CHEMICAL</sub>, **can decrease**, the overexpression of ([cyclin\_D1]<sub>GENE</sub>, [E2F-1]<sub>GENE</sub>), where ([cyclin\_D1]<sub>GENE</sub>, [E2F-1]<sub>GENE</sub>), **induced by**, [B(a)P]<sub>CHEMICAL</sub>>
- < [ATRA]<sub>CHEMICAL</sub>, **decrease**, ([cyclin\_D1]<sub>GENE</sub>, [E2F-1]<sub>GENE</sub>) < ([cyclin\_D1]<sub>GENE</sub>, [E2F-1]<sub>GENE</sub>), induced by, [B(a)P]<sub>CHEMICAL</sub>>>
- Hierarchical structure



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# Meta-Pattern Extraction

- What are meta-patterns?
- A mixed sequence of entity types and non-type words in the corpus

- E.g., pattern: **CHEMICAL** decrease **GENE**  
instance: **CHEMICAL** = *B(a)P, ATRA, ...*  
**GENE** = *cyclin\_D1, E2F-1, ...*

- New innovation: Hierarchical pattern grouping

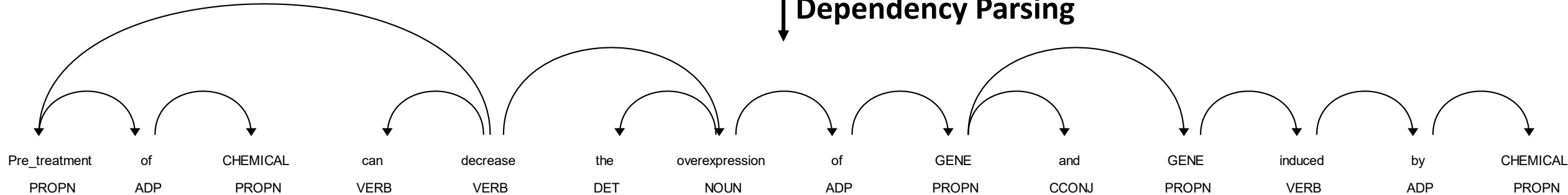
- pattern: {**CHEMICAL**} decrease {**GENE**}  
sub-patterns: {**CHEMICAL**} = *Pretreatment of CHEMICAL, ...*  
{**GENE**} = *the overexpression of GENE, GENE induced by CHEMICAL ...*

# Input: Biomedical Corpus

Pre-treatment of ATRA can decrease the overexpression of cyclin\_D1 and E2F-1 induced by B(a)P

...

## 1. BioNER Dependency Parsing



## 2. Sentence Break-Down

treatment can decrease overexpression  
pre-treatment of CHEMICAL  
the overexpression of GENE  
GENE and GENE  
GENE induced by CHEMICAL

## 3. Pattern mining

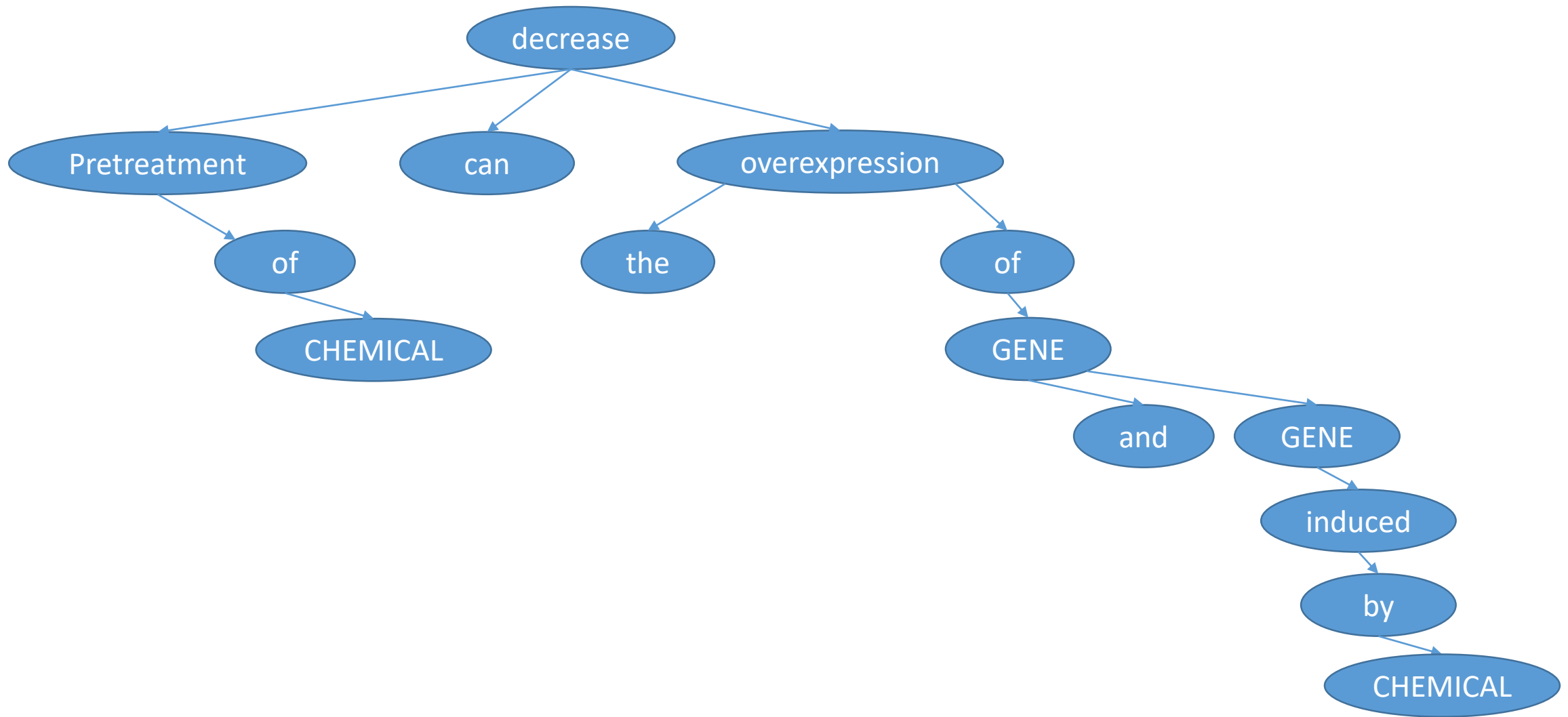
{CHEMICAL phrase} can decrease {GENE phrase}  
pre-treatment of CHEMICAL  
the overexpression of GENE  
GENE and GENE  
GENE induce by CHEMICAL

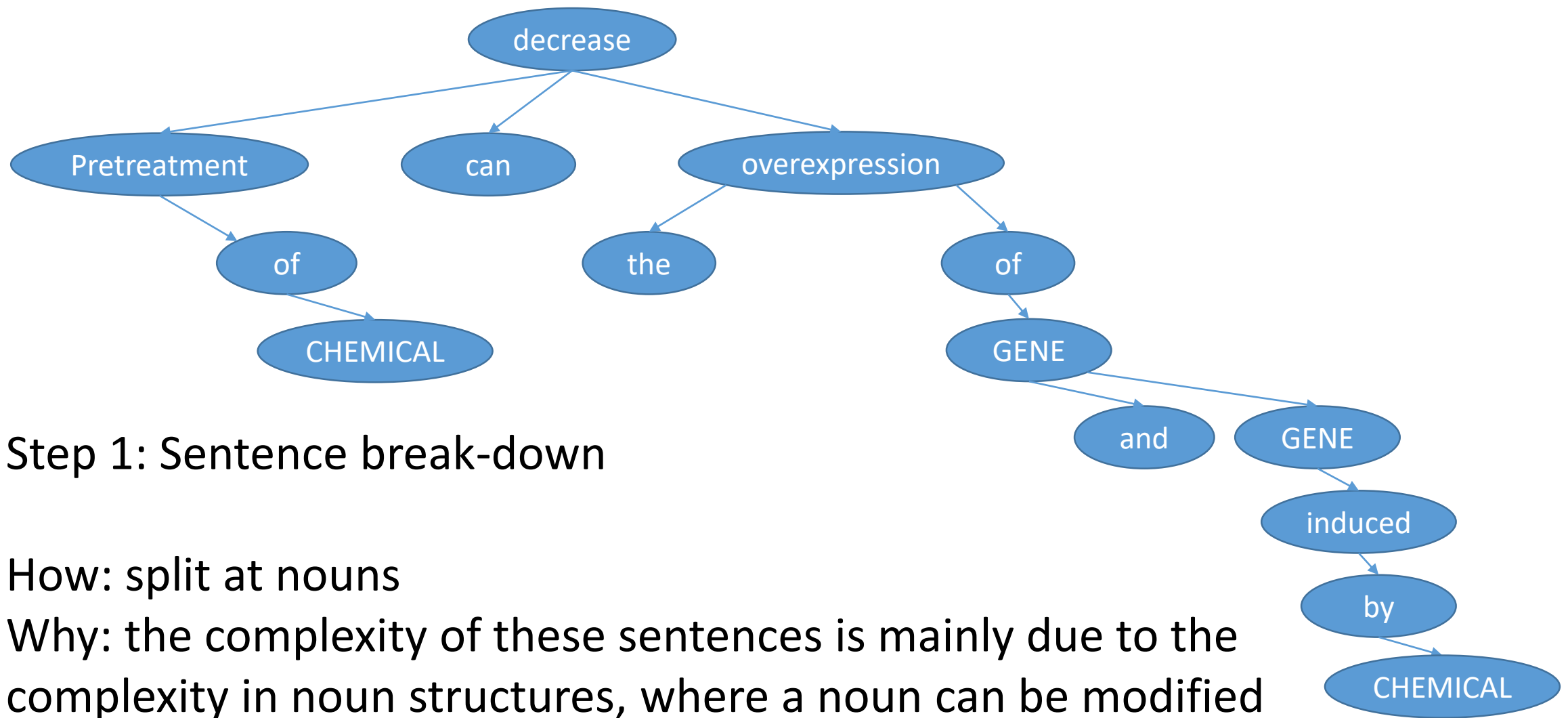
## 4. output

Patterns
{CHEMICAL phrase} can decrease {GENE phrase}
GENE induce by CHEMICAL
...

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Extractions
<ATRA, decrease, cyclin_D1:<(cyclin_D1, E2F-1), induced by, B(a)P>>
pre-treatment of CHEMICAL:ATRA can decrease the overexpression of GENE:cyclin_D1 and GENE:E2F-1 induced by CHEMICAL: B(a)P

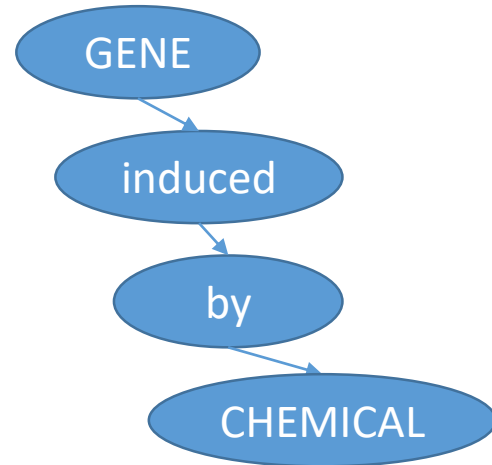
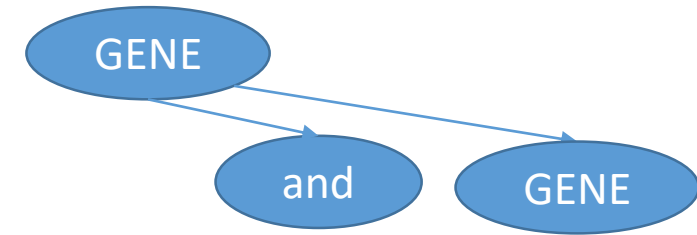
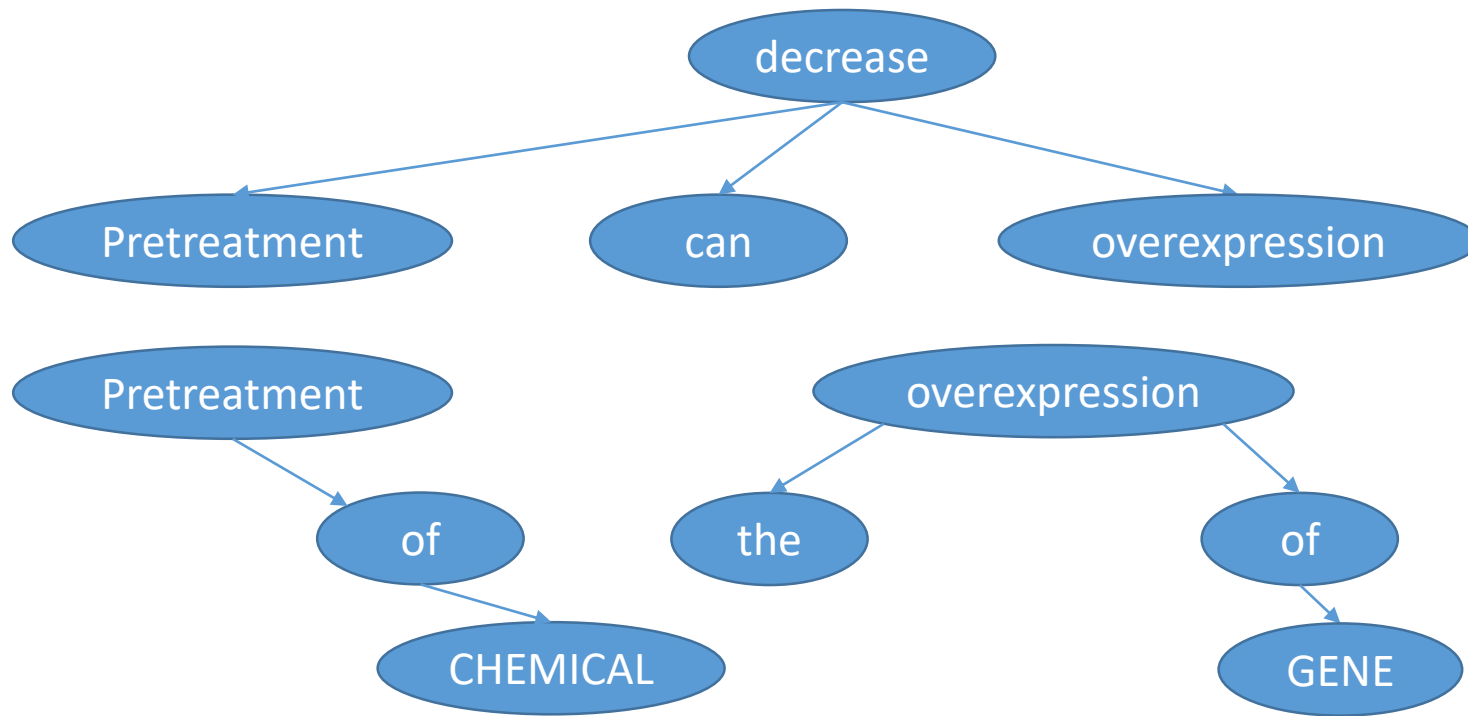




## Step 1: Sentence break-down

How: split at nouns

Why: the complexity of these sentences is mainly due to the complexity in noun structures, where a noun can be modified by other nouns, adjectives, adjectival clauses, etc.



## Step 1: Sentence break-down

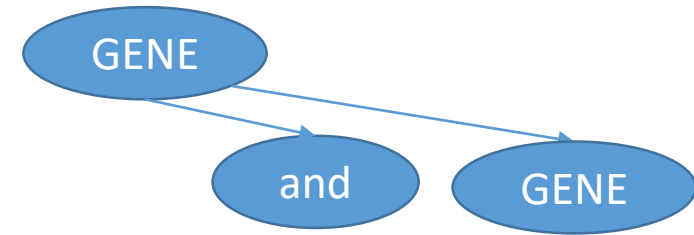
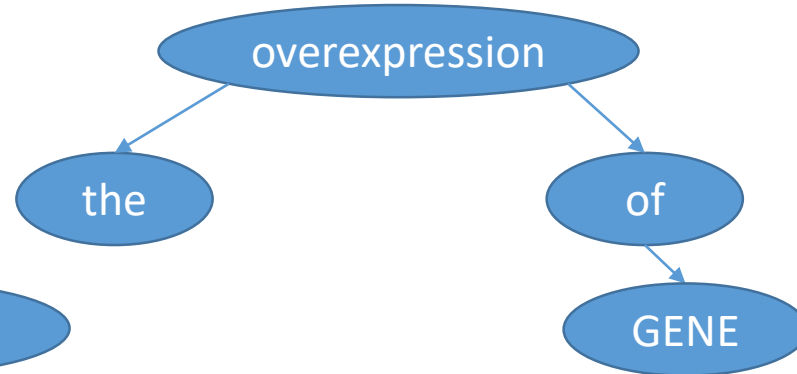
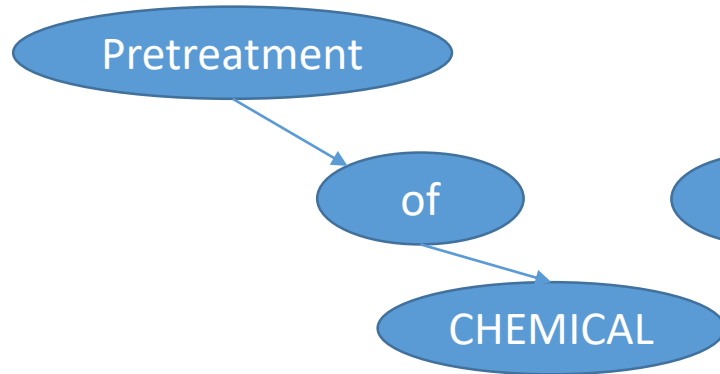
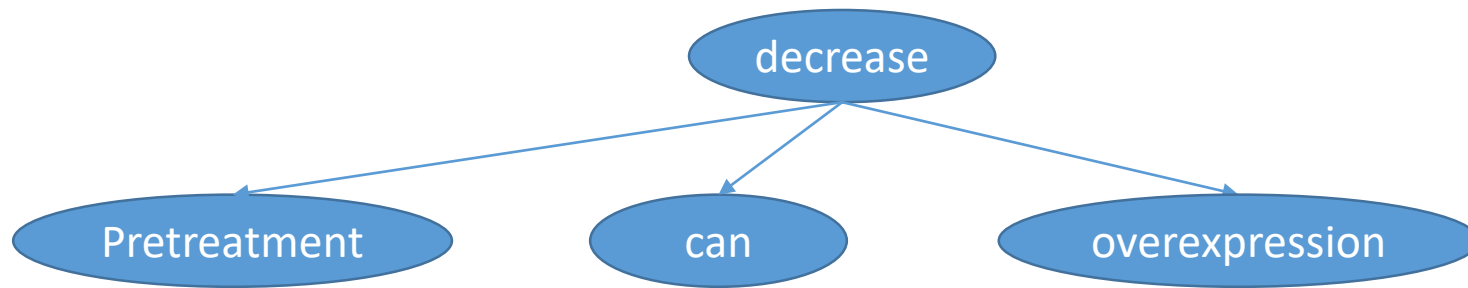
decrease: treatment can decrease overexpression

Pretreatment: Pretreatment of **CHEMICAL**

overexpression: the overexpression of **GENE**

**GENE**: **GENE** and **GENE**

**GENE**: **GENE** induced by **CHEMICAL**



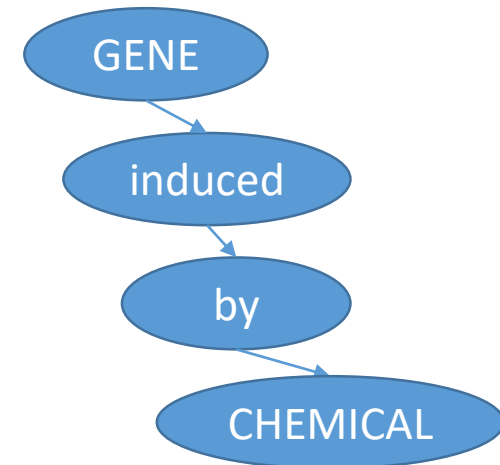
## Step 2: Pattern mining on short sentences

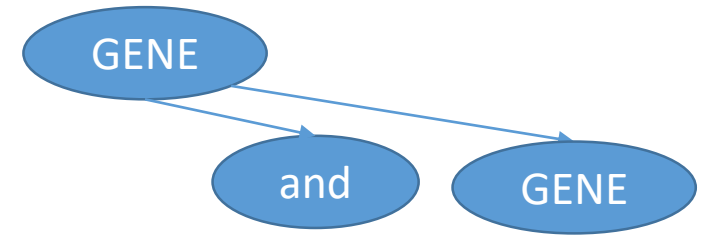
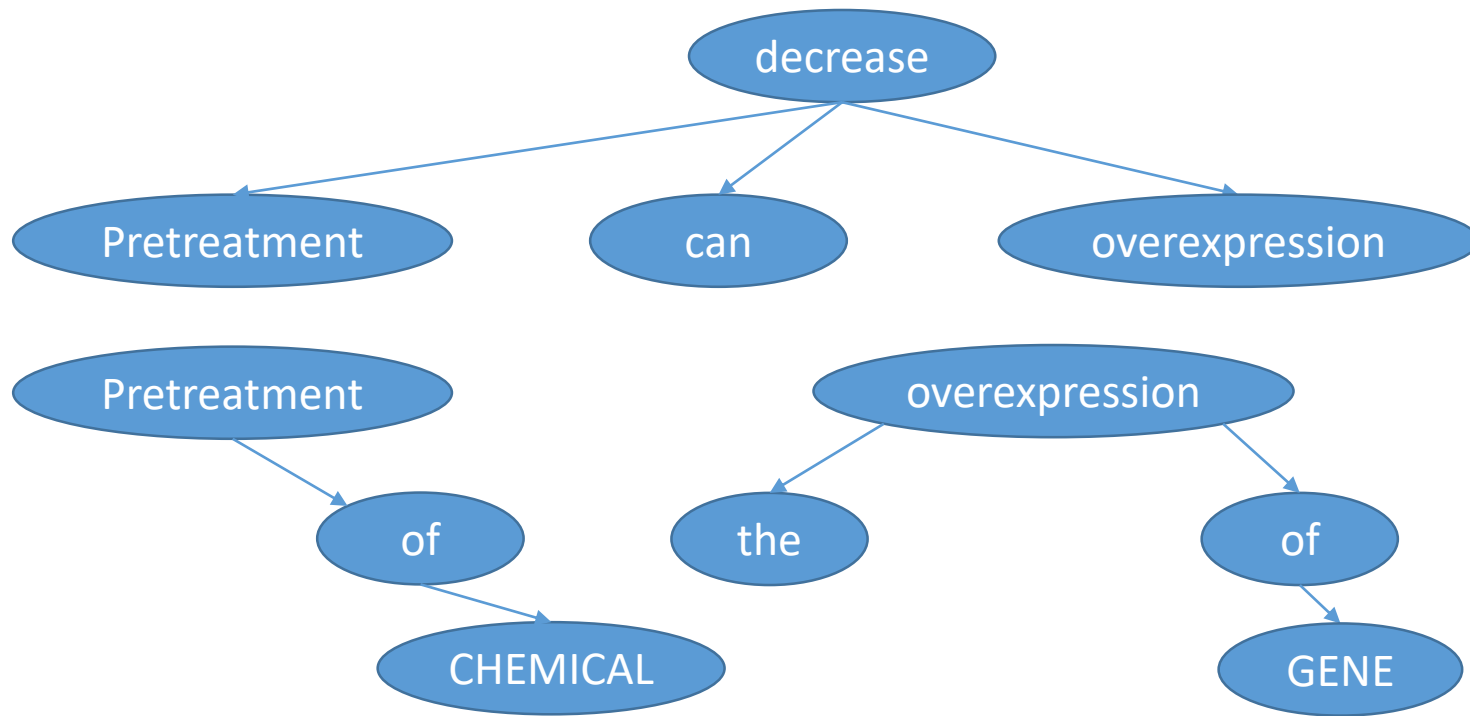
Constraint: words in a pattern should be connected on the tree

Eg. "pretreatment of CHEMICAL" ✓ "and GENE" ✗

Constraint: pattern should contain (one entity + one non-stop-word), or more than one entity

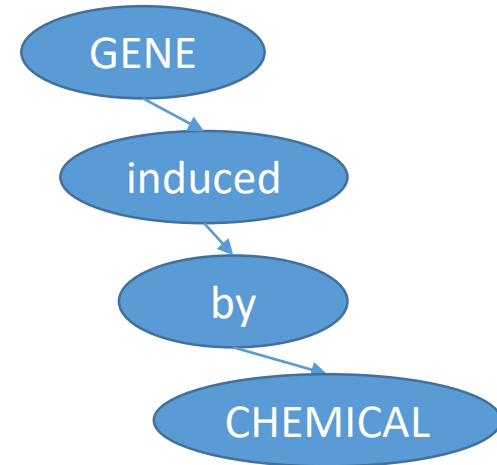
Constraint: frequency is high



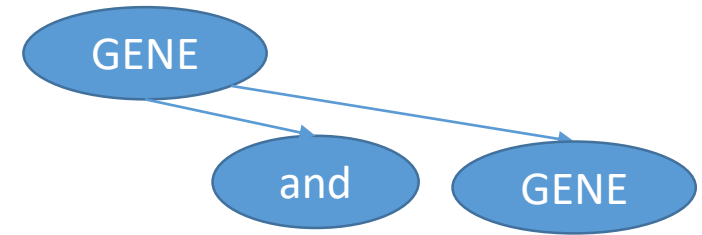
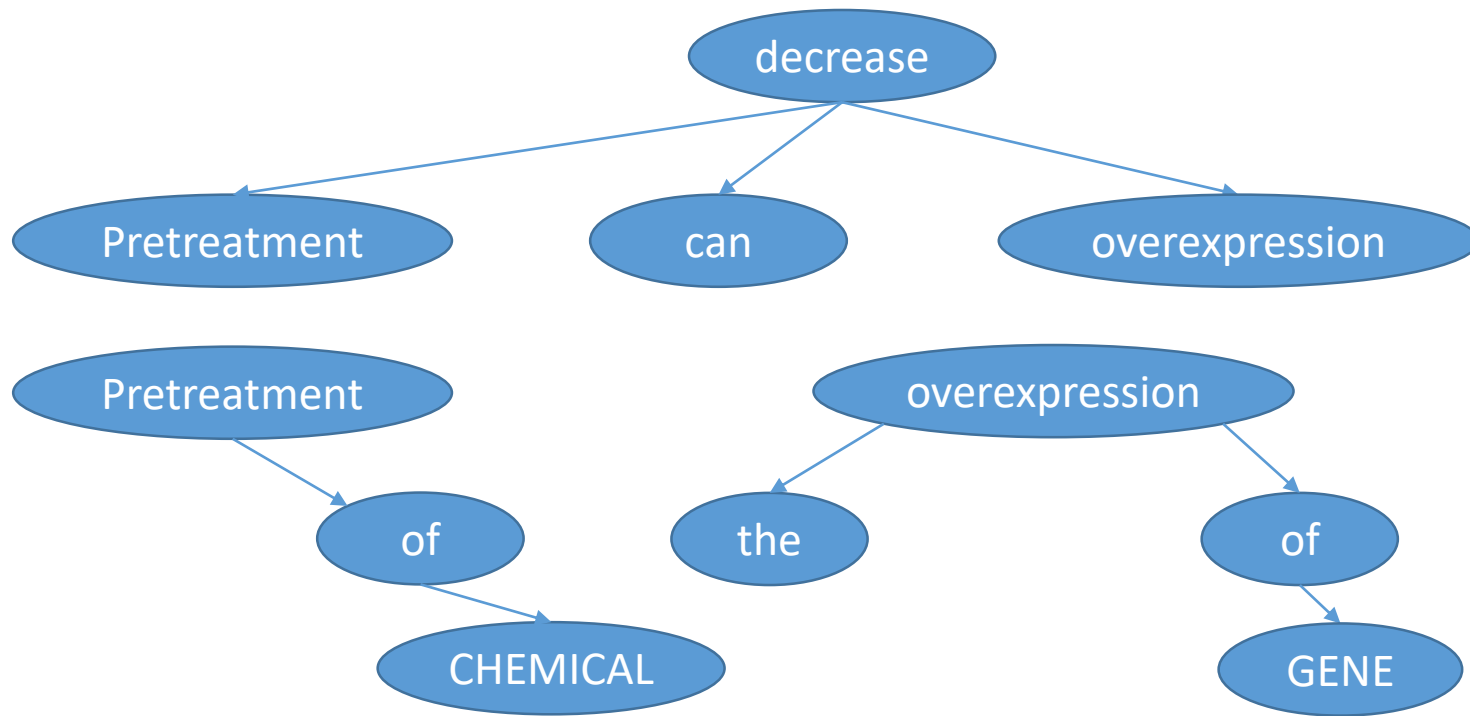


Step 2: Pattern mining on short sentences

pretreatment of **CHEMICAL**  
 the overexpression of **GENE**  
**GENE** and **GENE**  
**GENE** induced by **CHEMICAL**







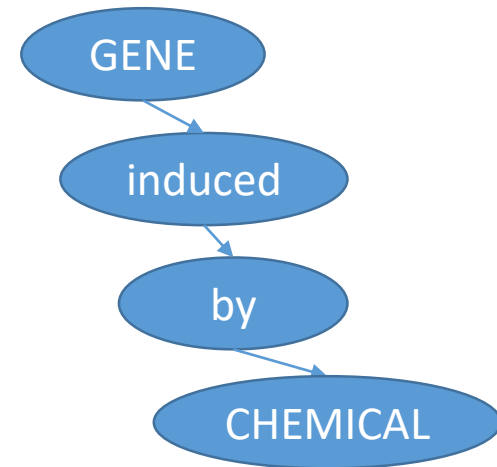
### Step 3: Pattern grouping

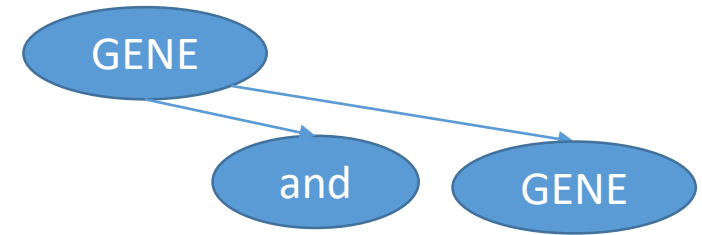
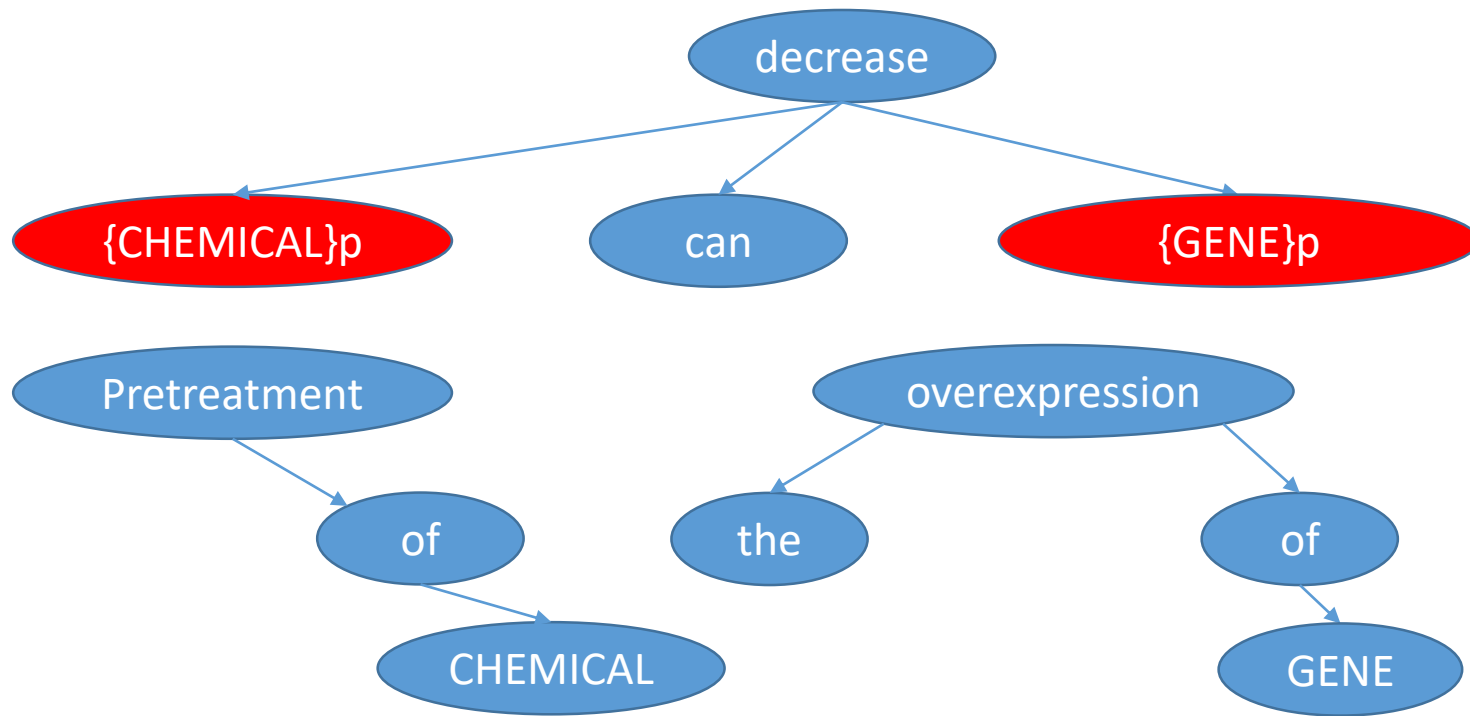
pretreatment of **CHEMICAL** ← **CHEMICAL** phrase

the overexpression of **GENE** ← **GENE** phrase

**GENE** and **GENE** ← **GENE** phrase

**GENE** induced by **CHEMICAL** ← **GENE** phrase





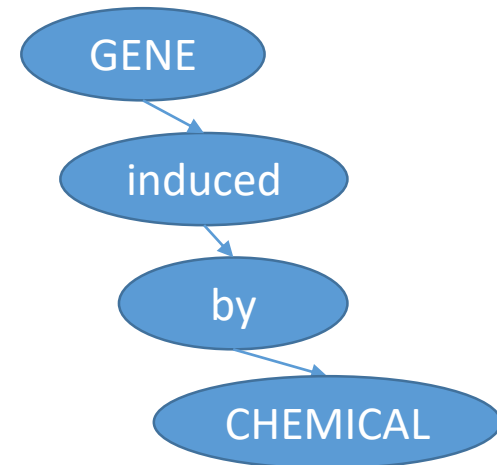
### Step 3: Pattern grouping

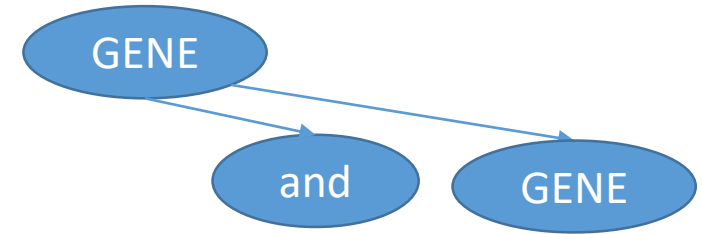
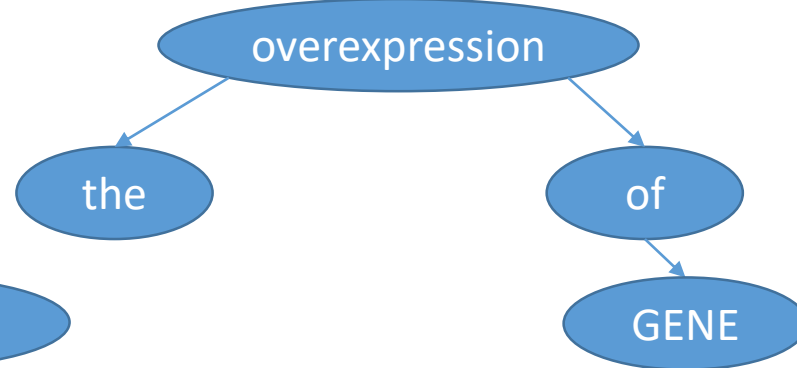
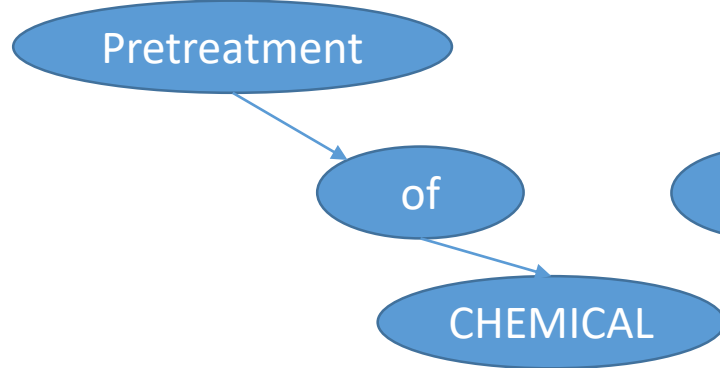
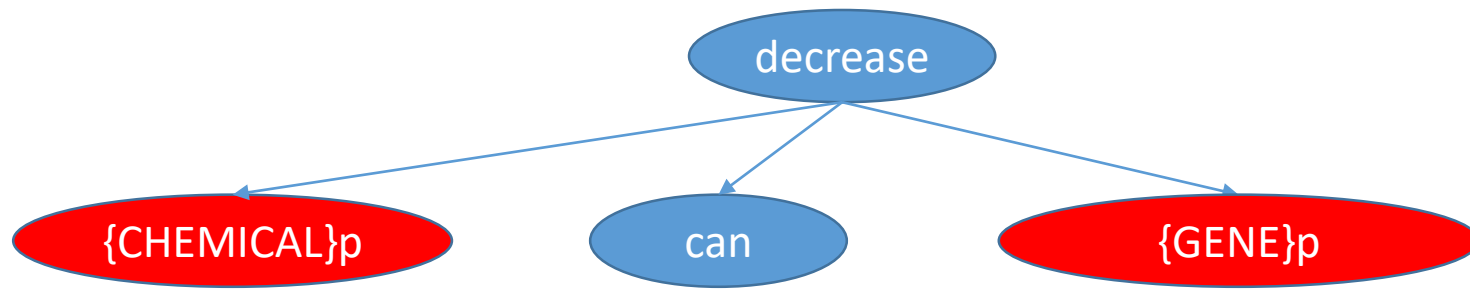
pretreatment of **CHEMICAL** ← **CHEMICAL** phrase

the overexpression of **GENE** ← **GENE** phrase

**GENE** and **GENE** ← **GENE** phrase

**GENE** induced by **CHEMICAL** ← **GENE** phrase

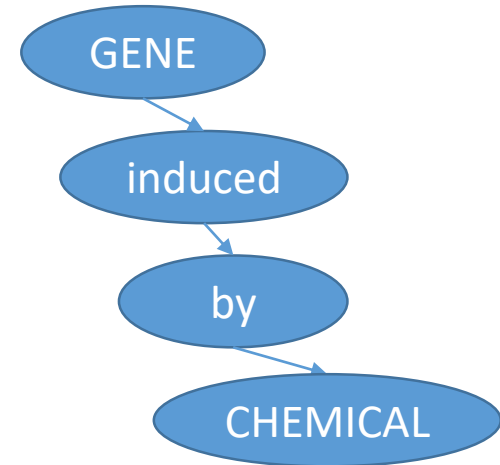


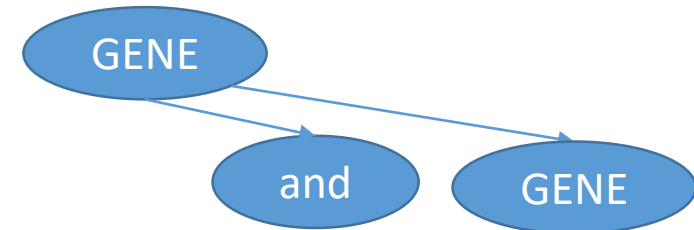
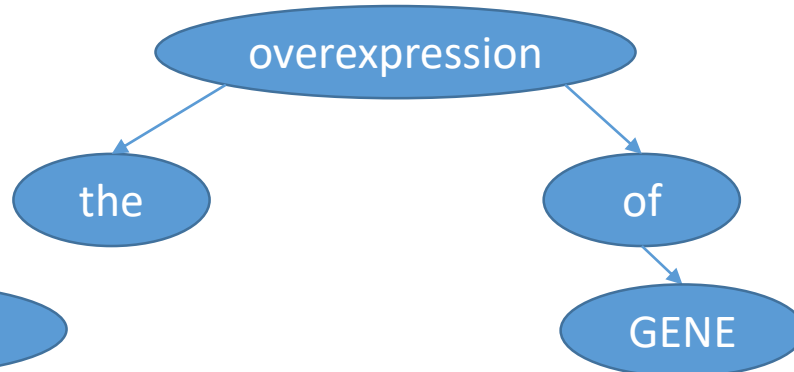
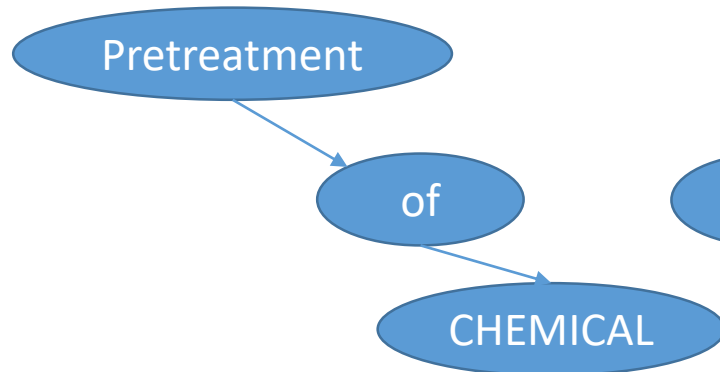
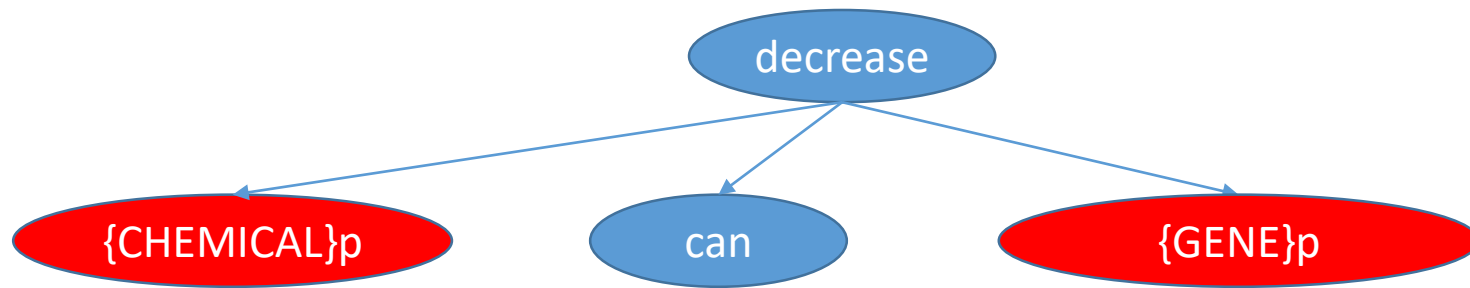


Step 2: Pattern mining

New pattern

{CHEMICAL}<sub>p</sub> can decrease {GENE}<sub>p</sub>

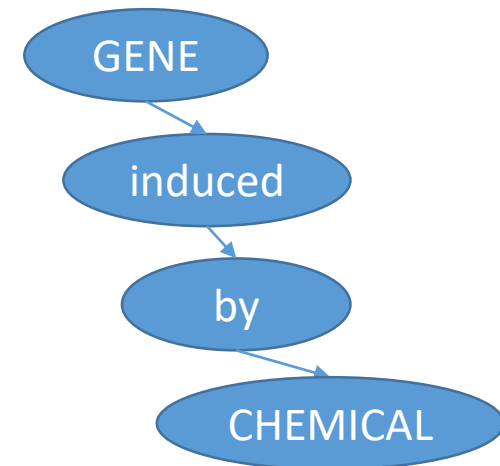




### Step 3: Pattern grouping

perform clustering to group synonymous meta patterns

- {CHEMICAL}p can decrease {GENE}p
- {CHEMICAL}p decrease {GENE}p
- {GENE}p be decrease by {CHEMICAL}p



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# Experiments

- Dataset: A subset of PubMed abstracts, selected using tuples in CTD

BASIC STATISTICS OF THE SUBSET CORPUS.

Abstracts	Sentences	Entity Mentions			
		Gene	Chemical	Disease	Species
28007	302736	215704	314134	129931	86697

- Baselines:
  - ClausIE: adopts clause patterns to handle long-distance relationships.
  - Stanford OpenIE: learns a clause splitter via distant training data.
  - Ollie: utilizes open pattern learning and extracts patterns over dependency path and part-of-speech tags.
  - MinIE: refines tuples extracted by ClausIE by identifying and removing parts that are considered overly specific.

# Performance comparison with state-of-the-art OpenIE systems

- Randomly sample 96 sentences for human labeling
- one tuple will be judged as correct if it reads smoothly and meets the fact described in the sentence

	# Correct extractions	# Valid extractions	Precision
ClausIE [12]	21	142	0.15
Stanford [13]	<b>120</b>	277	0.43
Ollie [11]	43	84	0.51
MinIE [14]	77	126	0.61
WW-PIE	110	150	<b>0.73</b>

- Note: we observe that Stanford OpenIE produces over 60 extractions for one sentence, which may be undesired for some applications.

# Pattern and Extraction Examples

Meta Pattern {CHEMICAL} reduce {DISEASE}	
Extractions in Expression Format	Extractions in Tuple Format
Ranitidine reduce ischemia/reperfusion-induced liver_injury in rats	<Ranitidine, reduce, liver_injury: <ischemia/reperfusion, induce, liver_injury, in, rats> >
resveratrol reduce brain_injury	<resveratrol, reduce, brain_injury >
Resveratrol reduce renal_and_lung_injury cause by sepsis in rats	<Resveratrol, reduce, renal_and_lung_injury: <sepsis, cause, renal_and_lung_injury, in rats> >
Resveratrol reduce TNF-a-induced U373MG human glioma_cell_invasion	<Resveratrol, reduce, glioma_cell_invasion: <TNF-a, induce, human glioma_cell_invasion> >
caffeine treatment reduce glioma cell proliferation	<caffeine, reduce, glioma >

Meta Pattern {CHEMICAL} inhibit {GENE}	
Extractions in Expression Format	Extractions in Tuple Format
Progesterone inhibit COX-2 expression	<Progesterone, inhibit, COX-2>
NAC treatment inhibit phosphorylation of Akt	<NAC, inhibit, Akt >
ATRA inhibit the expression of Ccnb1 and Ccna1	<ATRA, inhibit, (Ccnb1, Ccna1)>
Cypermethrin inhibit the interaction between the AR_AF1 and SRC-1	<Cypermethrin, inhibit, AR_AF1:<AR_AF1, interaction, SRC-1> >
PGF and H2O2 inhibit SOD1 protein expression and activity	<(PGF,H2O2), inhibit, SOD1>

Meta Pattern {GENE} cause {DISEASE}	
Extractions in Expression Format	Extractions in Tuple Format
mutations in the CSB gene cause Cockayne_syndrome	<CSB, cause, Cockayne_syndrome>
mutations in FOXP2 cause developmental_verbal_dyspraxia (DVD)	<FOXP2, cause, developmental_verbal_dyspraxia: < <developmental_verbal_dyspraxia, abbr, DVD> >
mutations in the hENT3 gene cause an autosomal_recessive_disorder in humans	<hENT3, causes, autosomal_recessive_disorder: <autosomal_recessive_disorder, in, humans> >
germline mutations in DIS3L2 cause the Perlman_syndrome_of_overgrowth and Wilms_tumor susceptibility	<(DIS3L2, cause, (Perlman_syndrome_of_overgrowth, Wilms_tumor) >



# Top 10 Single Entity Patterns

<b>Meta Patterns with Single Entity</b>	<b>#</b>
DISEASE cell	11210
effect of CHEMICAL	9507
GENE expression	6551
expression of GENE	4940
CHEMICAL treatment	4896
GENE gene	4229
CHEMICAL exposure	3957
the effect of CHEMICAL	3721
GENE mrna	3211
CHEMICAL level	3076

- Can be helpful in named entity recognition tasks
- PENNER: Pattern-enhanced Nested Named Entity Recognition in Biomedical Literature

# Synonymous Pattern Group Examples

<b>Synonymous group</b>	<b>Meta Patterns</b>
<b>CHEMICAL_induced inhibition of GENE</b>	GENE inhibition by CHEMICAL CHEMICAL block GENE GENE inhibitor , CHEMICAL GENE inhibitor CHEMICAL
<b>CHEMICAL activate GENE</b>	CHEMICAL_activated GENE GENE activator CHEMICAL GENE agonist CHEMICAL GENE agonist , CHEMICAL GENE ligand CHEMICAL GENE ligand , CHEMICAL
<b>DISEASE cause by CHEMICAL</b>	CHEMICAL_induced DISEASE CHEMICAL can cause DISEASE CHEMICAL induce DISEASE CHEMICAL cause DISEASE DISEASE be induce by CHEMICAL DISEASE induce by CHEMICAL DISEASE produce by CHEMICAL
<b>SPECIES treat with CHEMICAL</b>	CHEMICAL administration to SPECIES CHEMICAL_treated SPECIES CHEMICAL_exposed SPECIES CHEMICAL treat SPECIES SPECIES be inject with CHEMICAL SPECIES be administer with CHEMICAL ...

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# Conclusion and Future Work

- WW-PIE
  - can extract all variety of the relation tuples from large biomedical literature corpora
  - resolves the long and complicated sentence structures by breaking down the sentences
  - groups meta-patterns hierarchically to extract n-ary hierarchical tuples
- Discussion and Future Work
  - Pattern grouping can be enhanced
  - Negation structures. For example, “there is no evidence that ...”
  - Dependency parser may introduce noise

Thank you! Questions?