

MuLLinG

MultiLevel Linguistic Graphs for Knowledge Extraction

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Motivation

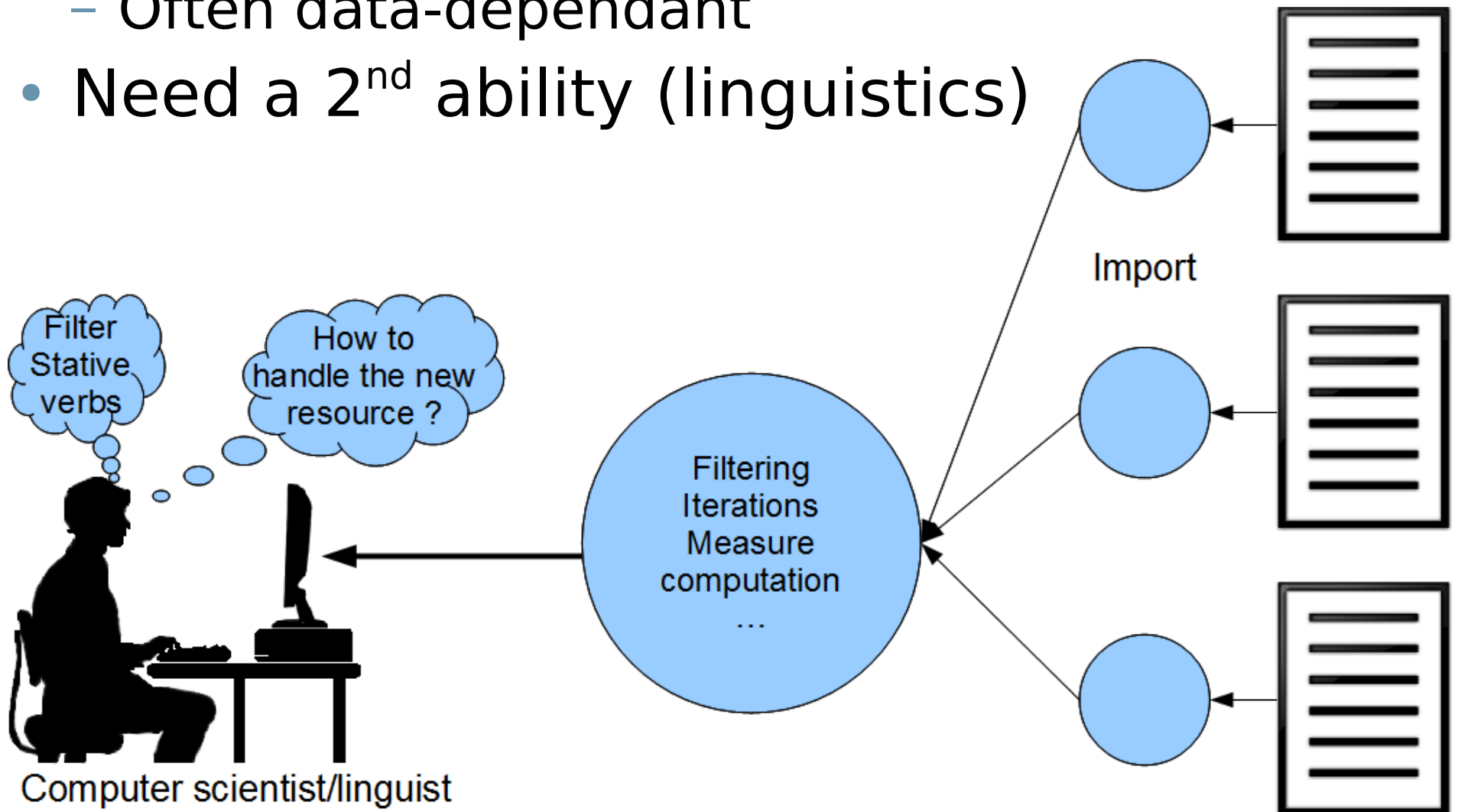
- NLP: Low-quality results
 - Ambiguities, particular phenomena
- Lack of linguistic (lexical) resources
 - Focus: extraction of lexical information
 - Computer scientist POV
 - Should be produced quickly and precisely
 - We must combine human and machine abilities
- Need: Generic tools

Outline

- Make lexical extraction easier
- MuLLinG model
- Application: collocation extraction

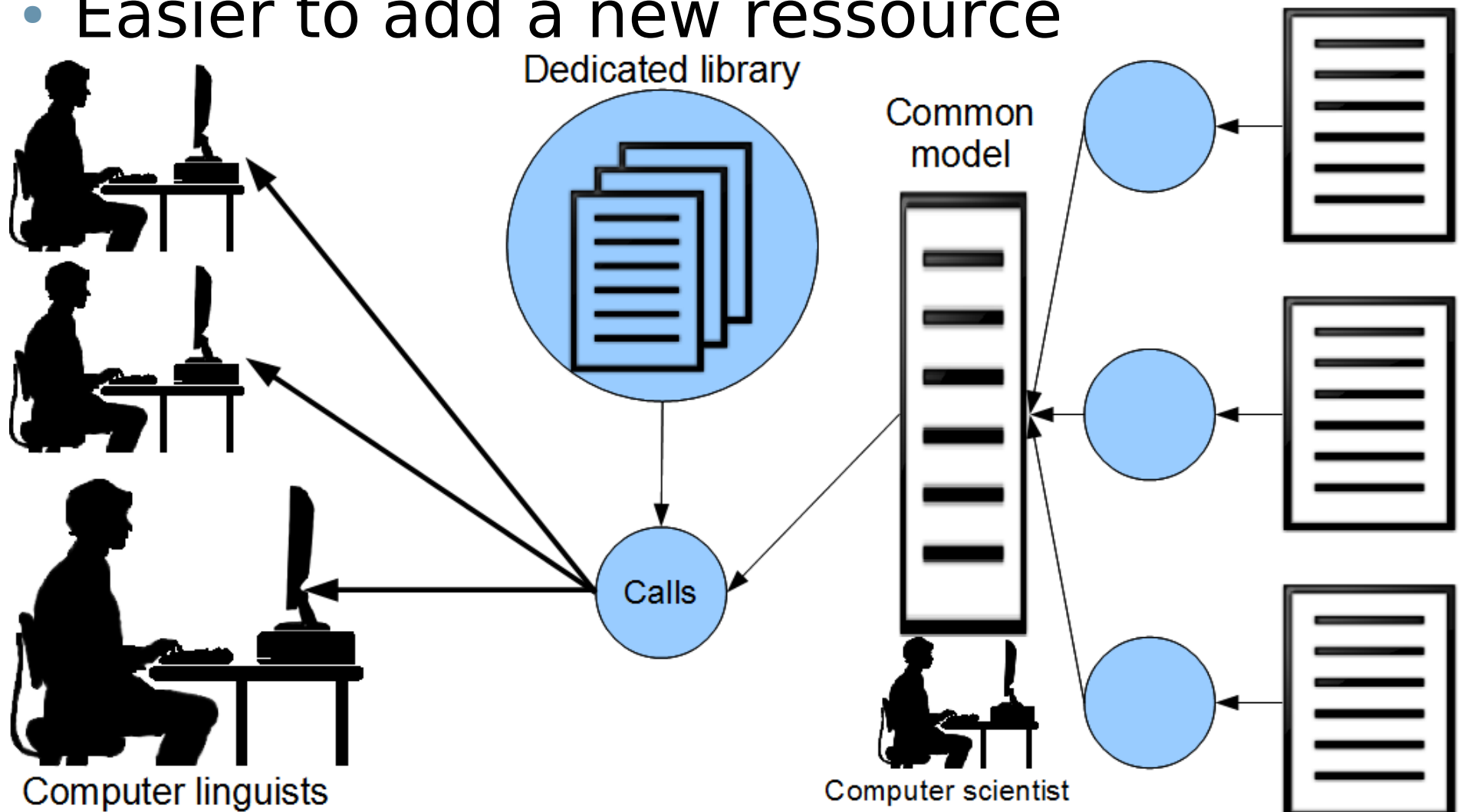
Difficult Programming

- Resource management (add a new one?)
 - Often data-dependant
- Need a 2nd ability (linguistics)



Separation of tasks - Generic tools

- Requires less knowledge in programming
- Easier to add a new resource

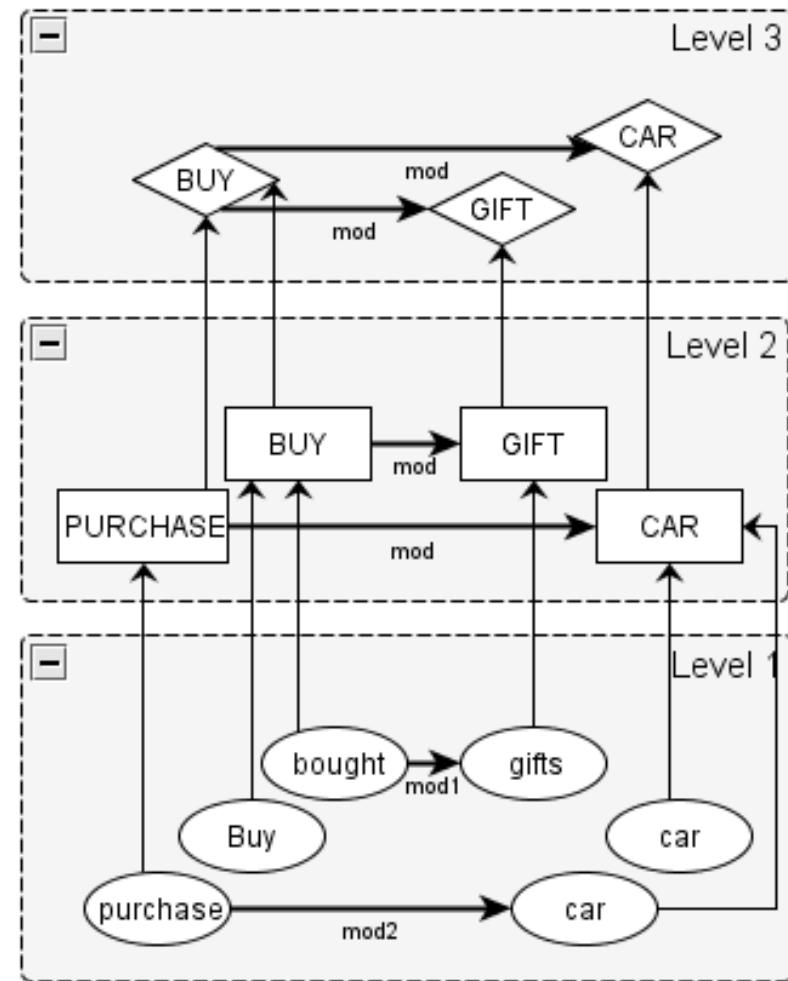


What kind of model ?

- Simple representation of data
 - Expressive (able to model complex data)
 - Generic (corpora, dictionaries, etc.)
- Simple generic operations
 - Task- and data-independant
 - High-level
 - Combine simple operations, rather than write a complex one
- => Graphs
 - Relations (juxtaposition, dependency, etc.)
 - Easy to understand/handle, widely used in NLP ₆

MuLLinG: MultiLevel Linguistic Graph

- Levels=different views
- Grouping by *equivalence classes*
 - 1 class = 1 node at superior level
 - Level hierarchy
 - Interlevel edges
 - Between a node and its class
- Attributes are free
 - No constraint on the data

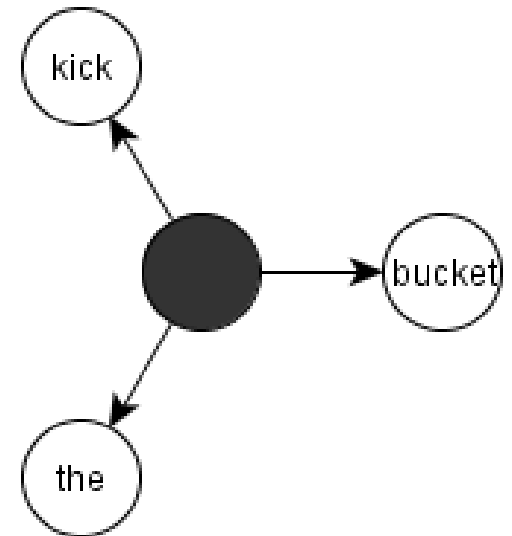


Associated operations

- Modifying the graph:
 - Parameters: level, filtering function, attribute computation functions (for nodes/edges)
 - Given by the user
 - *Emergence* creates a new level
- Union, intersection, difference of graphs
 - Based on identity of nodes/edges
- Basic:
 - Add/delete edge, node (and its descent)
 - Conditional application
 - Measure computation

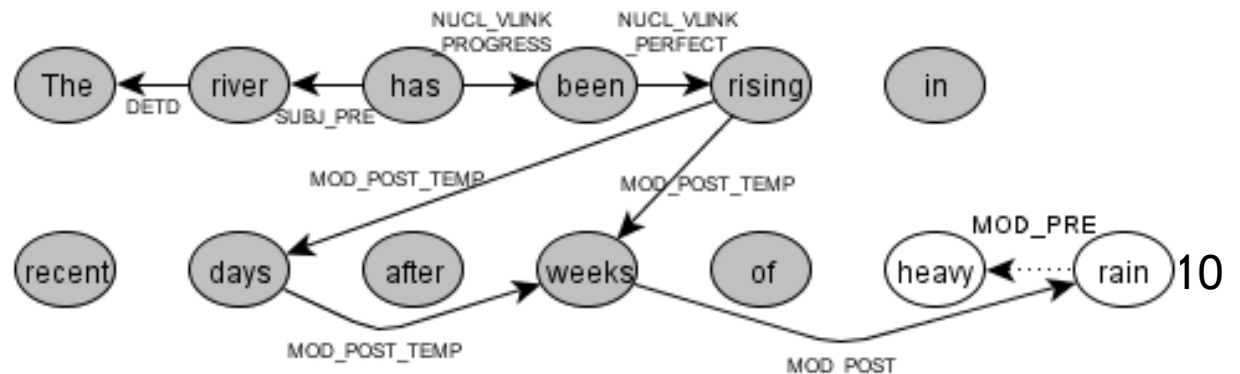
Complex version

- Relations not always binary
- 1 relation =
 - 1 (standard) node materializing the relation
 - + numbered argument edges
- (operators are adapted)



Experiment: collocation extraction

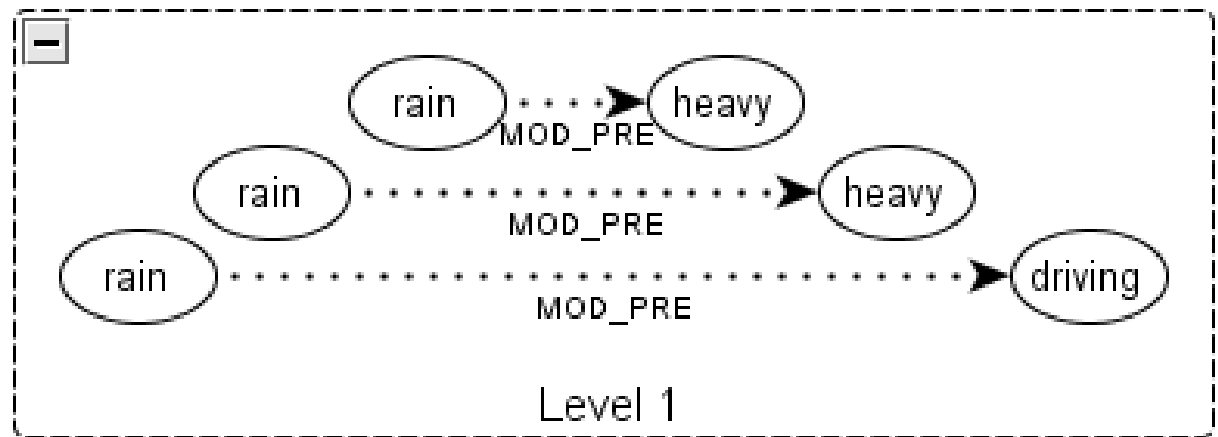
- Collocation (« *Driving rain* ») = semi-fixed
 - One term is chosen arbitrarily
 - In function of the other one
 - To express a particular meaning
 - Problem for translation
- Initial graph:
 - Dependencies produced by the parser (XIP)



Collocation extraction

- *Emergence* produces the superior level
 - Operation based on equivalence classes
 - Before: relations between objets
 - After: (grouped) relations between grouped objects
 - Parameters:
 - Level, filter, attribute computation
 - + function identifying the class of a node/edge

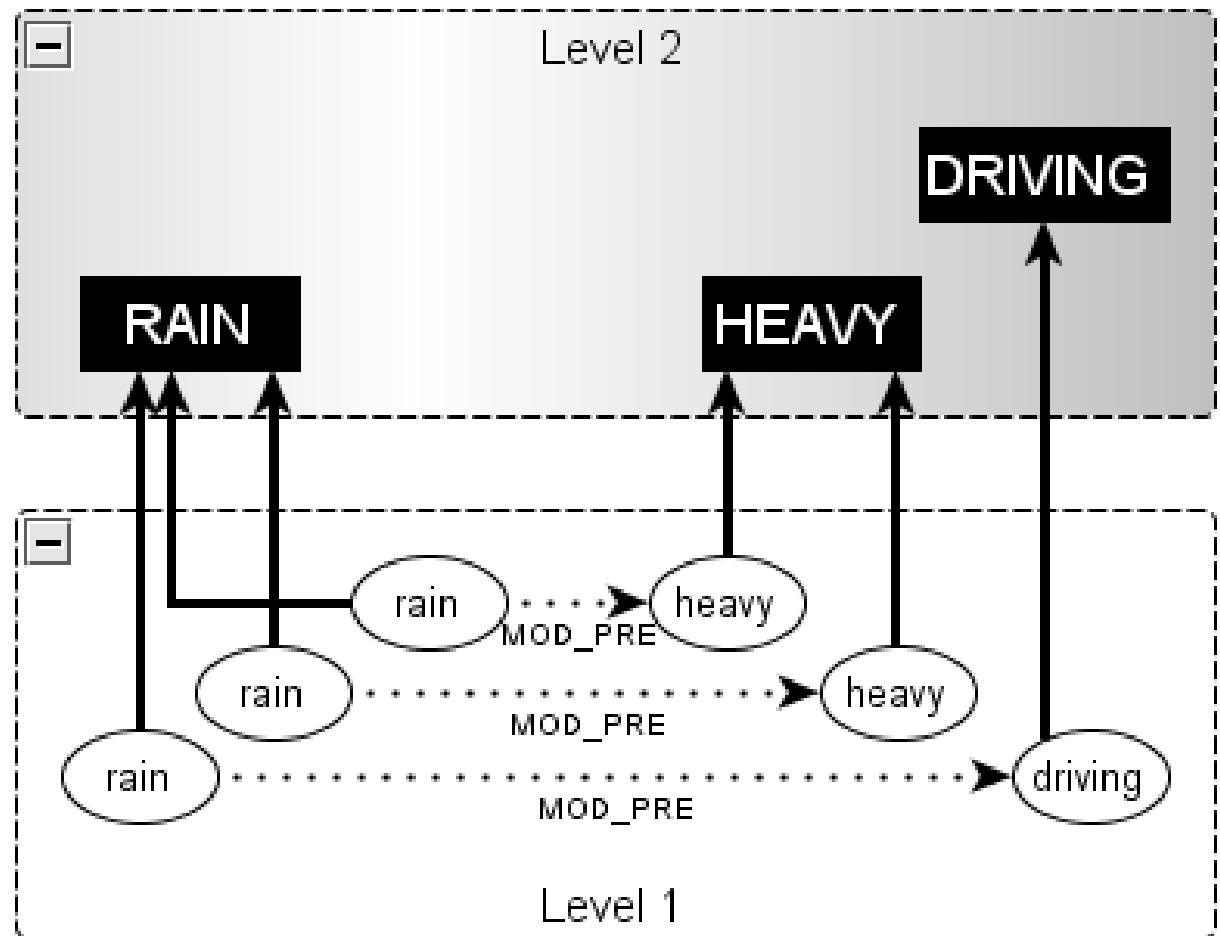
- Filtering relations:
 - Removing nodes



Collocation extraction Node emergence

- 1 node = 1 equivalence class
 - Linked to nodes (at inferior level) elements of the class

```
NodeEmergence(  
1,  
true,  
Class_Lemma_pos(),  
CompAttrNode()  
  //<id, idclass>  
  <type, "term">  
  <nboccs, incr>  
)
```



Collocation extraction

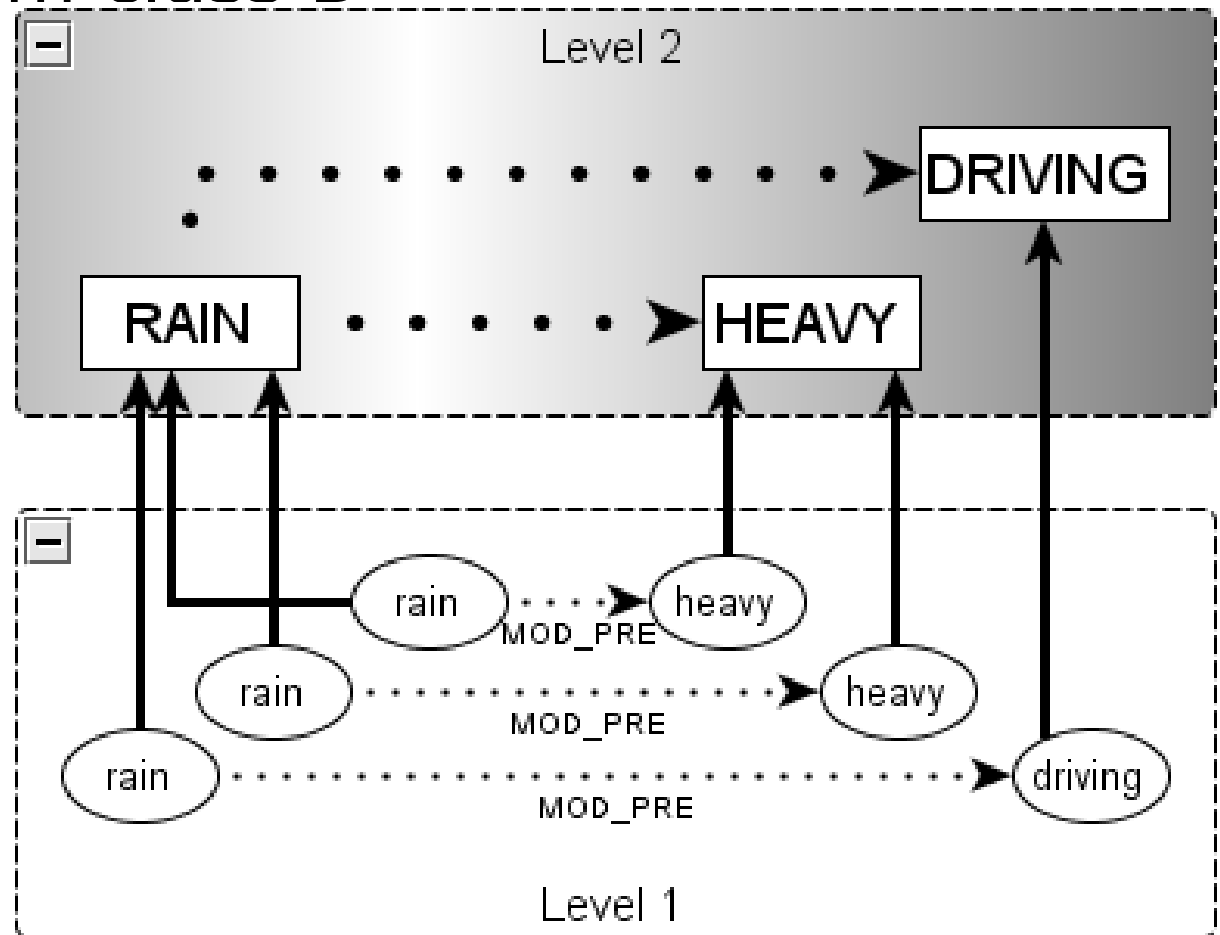
Edge emergence

- 1 edge between A and B = 1 set of edges
 - Between an element from class A, and an element from class B
 - Equivalent

```

EdgeEmergence(
  1,
  true,
  Class_Type()
  CompAttrEdge(),
    //<id, idclass>
    <type, "classmod">
    <nboocs, incr>
  CompAttrSource(),
    //<d+, incr>
  CompAttrTarget() )
    //<d-, incr>

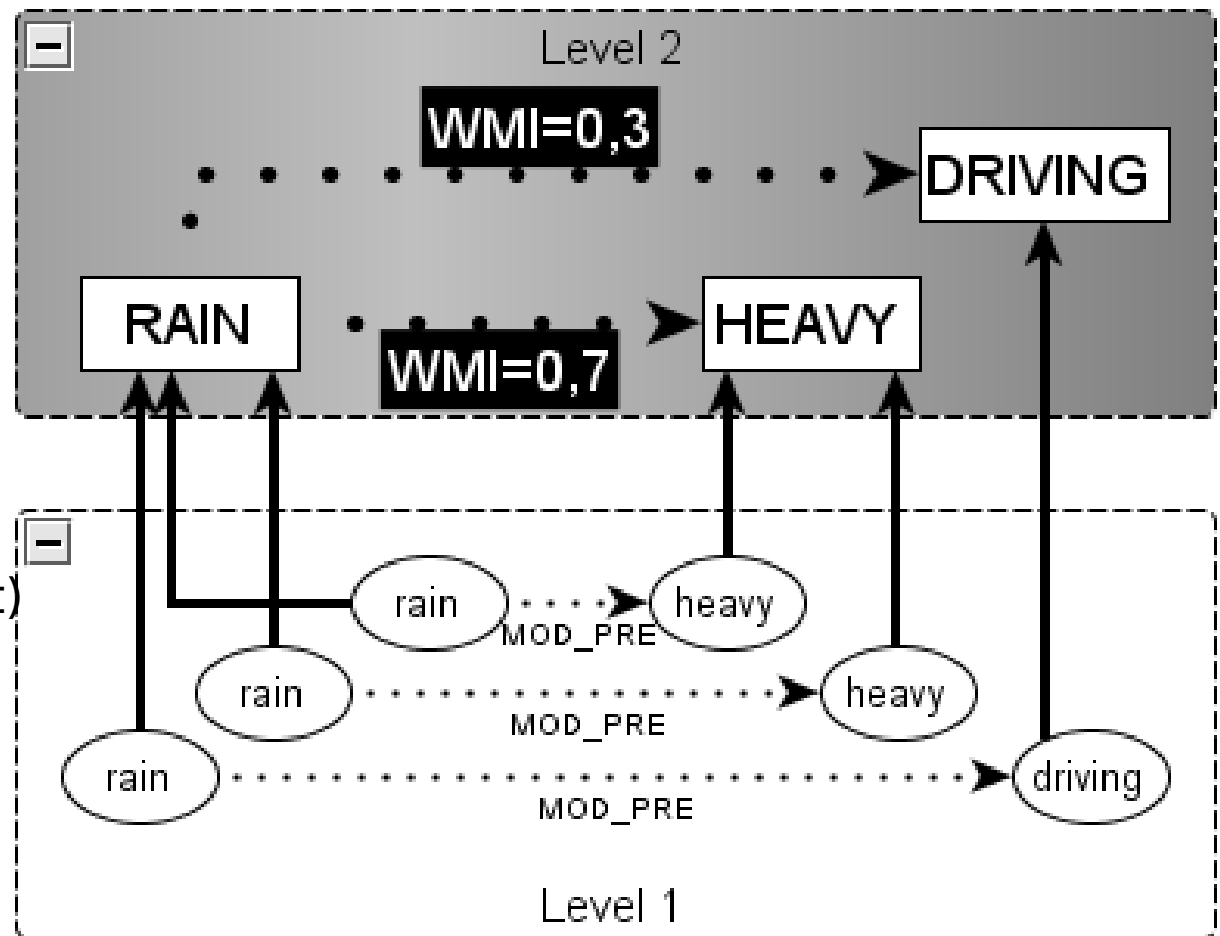
```



Collocation extraction Measure computation

- Using values previously computed
 - Number of (co-)occurrences, in/out degree...
- Candidates :
level2 edges

```
ComputeMeasure(  
WMI(),  
    //association measure  
is_classmod(),  
"measure",  
    //where to write (result)  
"nboccs",  
"nboccs",  
    //where to read  
NumberSentences  
...)
```



Observations

- Coherent results
- Mulling (open-source C++ library)
 - <http://mulling.ligforge.imag.fr>
 - In/out file format: GraphML
 - **~70 lines** (calls)
 - vs. *Ad hoc*: **~400 lines** (iterations on the data)
 - Much faster description / Avoid programming errors
 - Import : ~250 lines (vs. 200 lines ad hoc)
 - Execution quite slower
 - less optimized
 - Generic: reusable with any kind of relation

Future works

- Library usability
 - Import
 - High-level (request) language
 - Graphic interface
 - Memory: use databases (+cache) to store large graphs
- Graph clustering
- Applications to other graphs
 - Less NLP-centered
 - Semantic web (RDF/SPARQL)
 - Social networks