



DynaLearn, una herramienta de razonamiento cualitativo

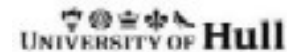
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Asignatura de “Modelos de Razonamiento”, Noviembre 2011

1. Introduction
2. Qualitative Reasoning modelling
3. Semantic Technologies
4. Semantic Grounding
5. Semantic Feedback
6. Research and in-use questions



“Engaging and informed tools for learning conceptual system knowledge”

QUALITATIVE REASONING

- Trata de capturar la **interpretación humana** de la realidad
- Representa sistemas físicos mediante **modelos**
- Estudia su comportamiento mediante **simulación**
- Enfocado en las **variables cualitativas** de los sistemas (ej.: cierto árbol es de tamaño “grande”, la población de cierta especie “aumenta”, etc.) más que en variables numéricas

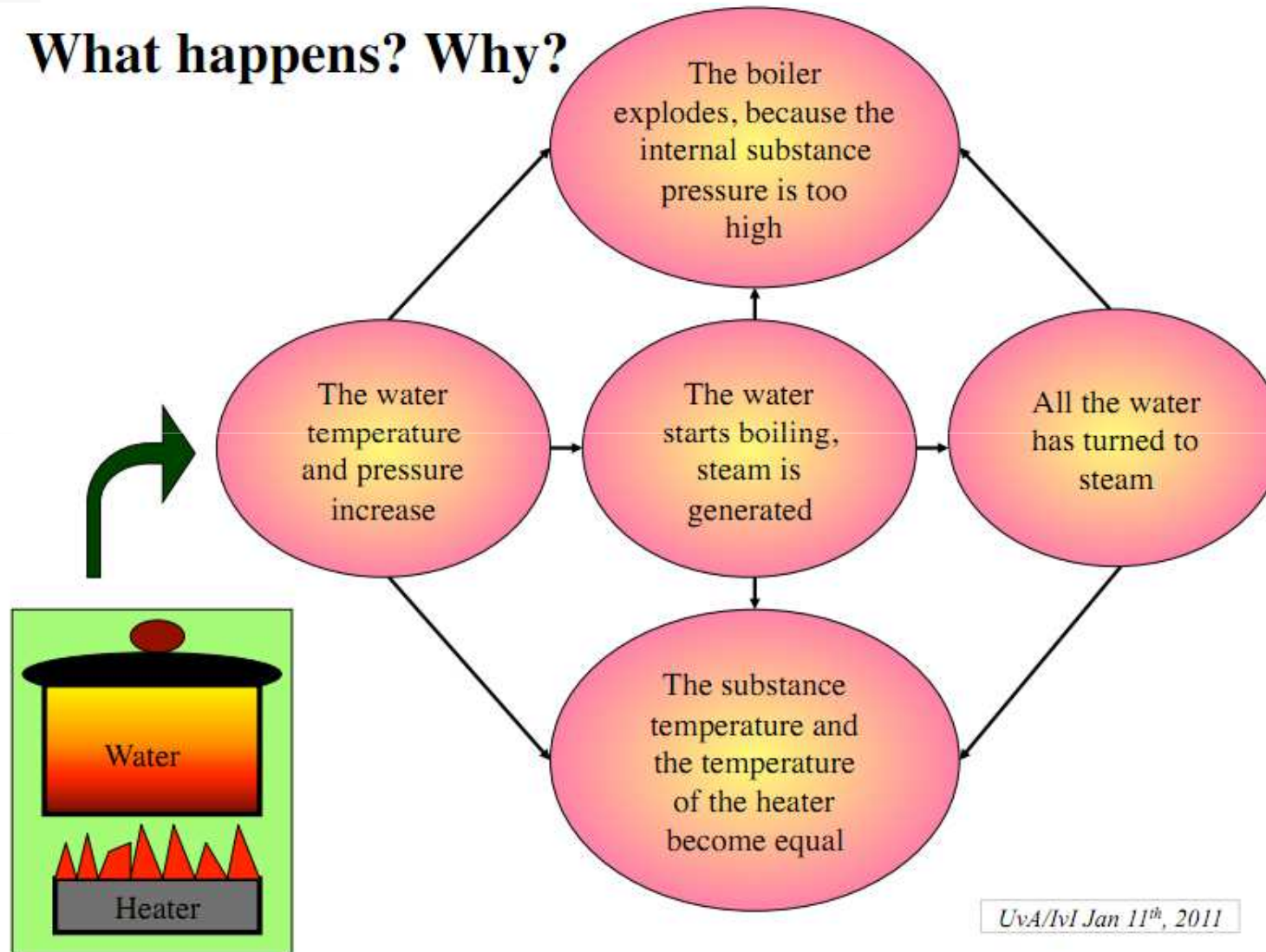


APPLICATION TO LEARNING OF ENVIRONMENTAL SCIENCES

- Idea: **aprendizaje** basado en construir modelos (“Learning by modelling”)
- Herramientas para el aprendizaje:
 - Definición de **terminología** apropiada
 - **Interacción** con el modelo
 - Predicción de su **comportamiento**
- Ejemplos:
 - “*Estudiar la evolución de la población de una especie cuando se introduce otra en su mismo ecosistema*”
 - “*Estudiar el efecto de los agentes contaminantes en un río*”
 -



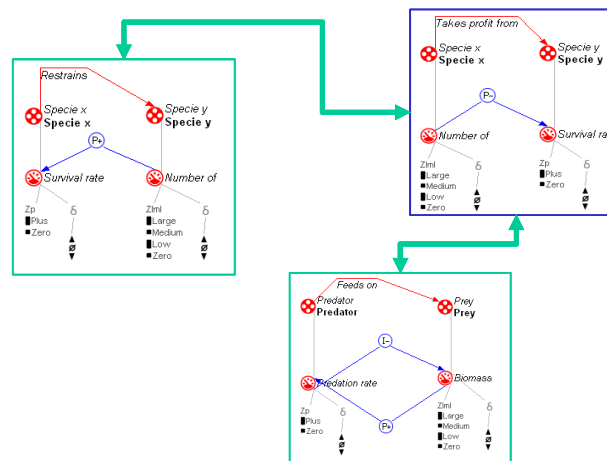
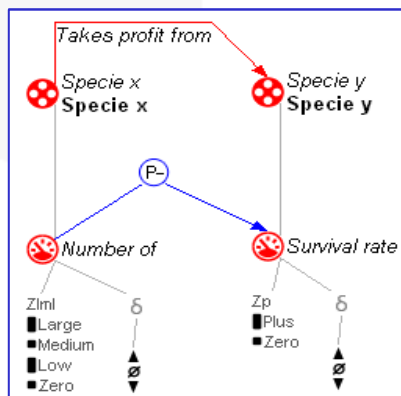
What happens? Why?



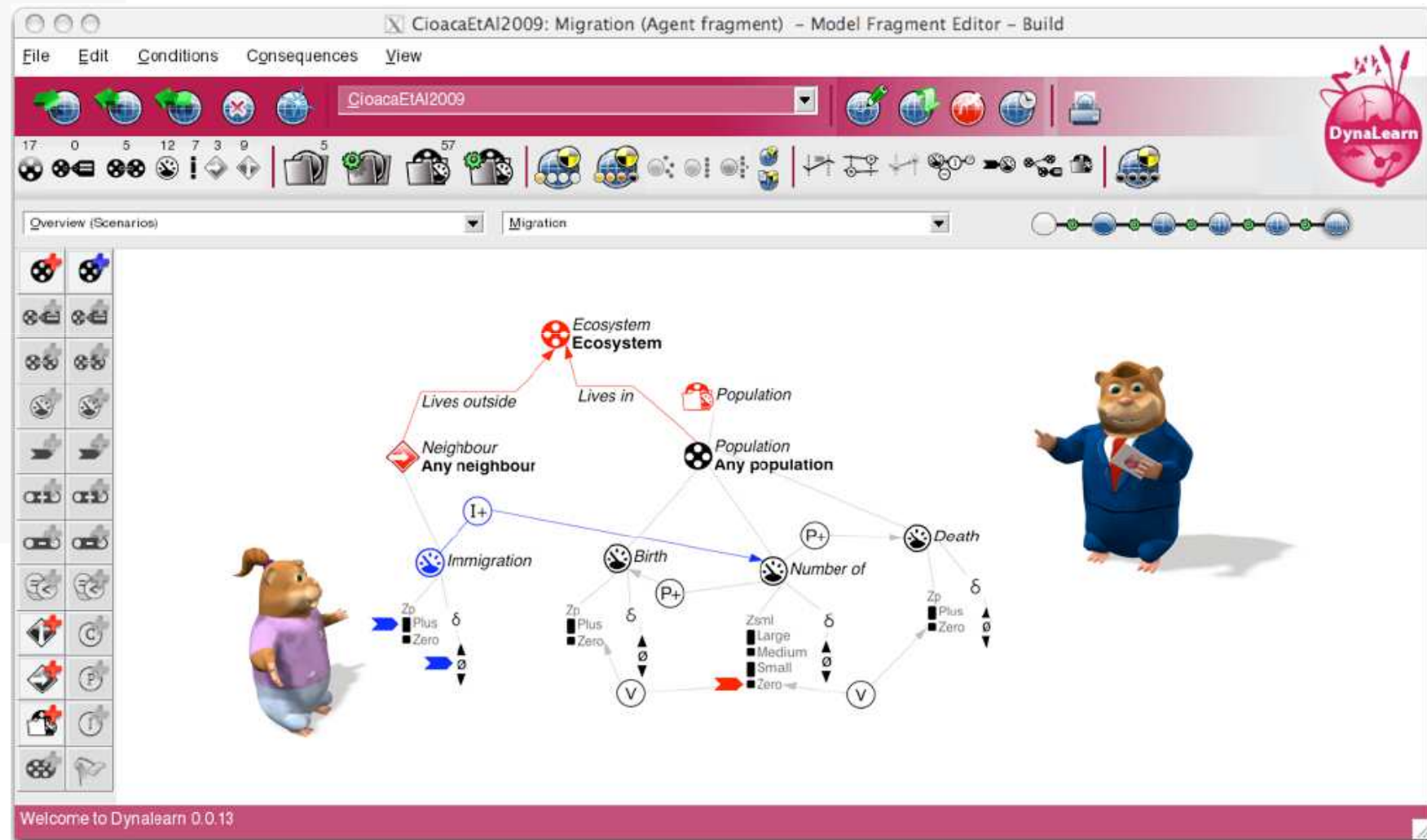
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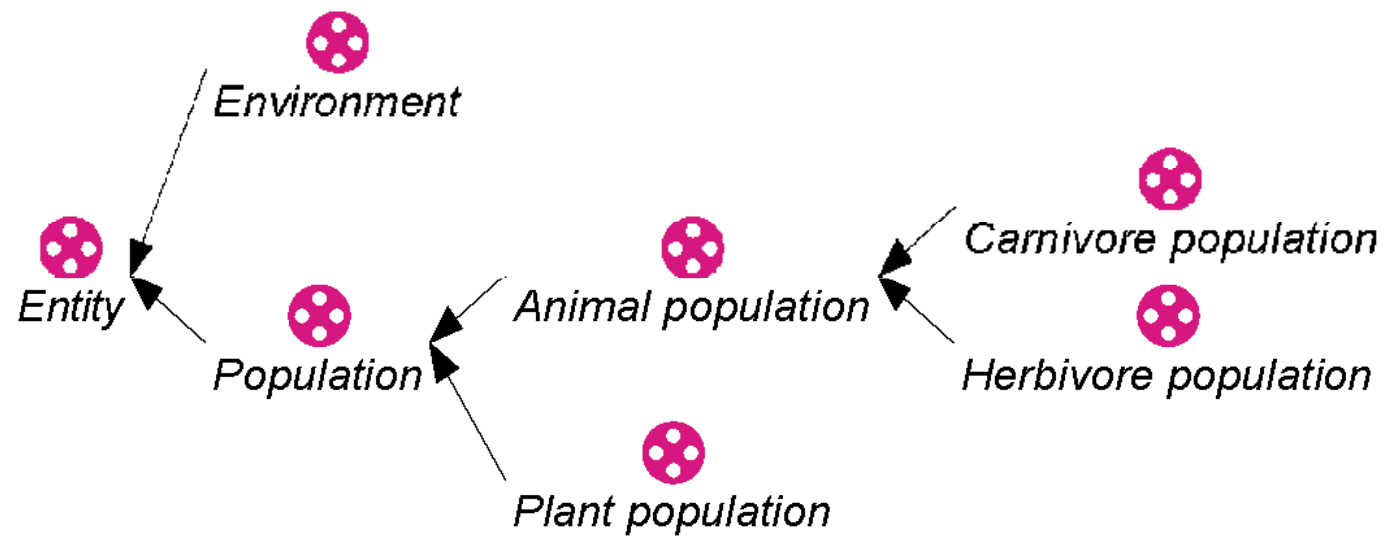
- “Sistema para la adquisición de conocimiento conceptual en el contexto de la enseñanza de ciencias medioambientales”. Combina:
 - Construcción de **modelos** representando un sistema
 - Técnicas semánticas para **relacionar** distintos modelos de alumnos y profesores
 - Uso de **avatares** para interactuar con el sistema



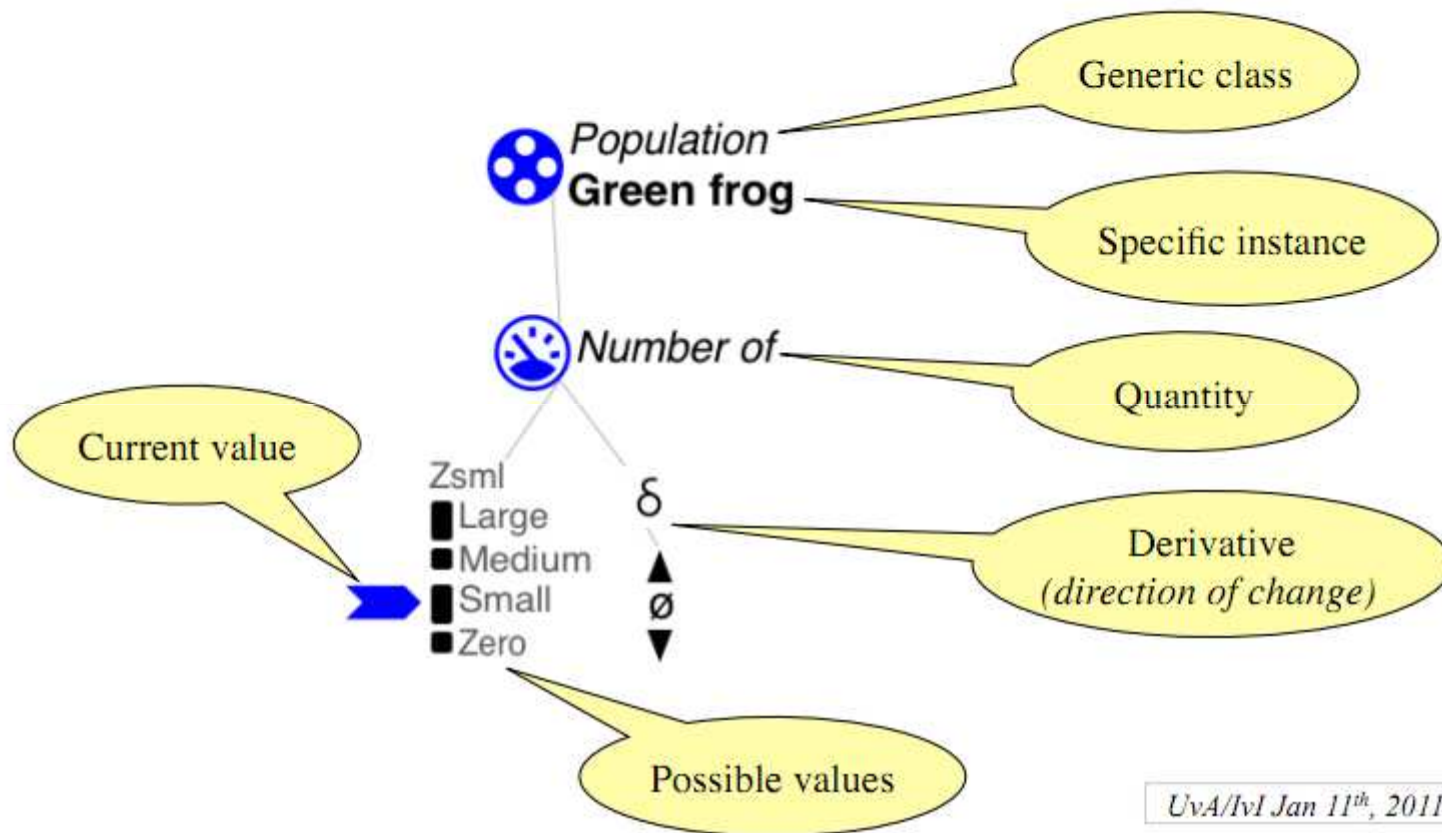
DYNALEARN



Entities

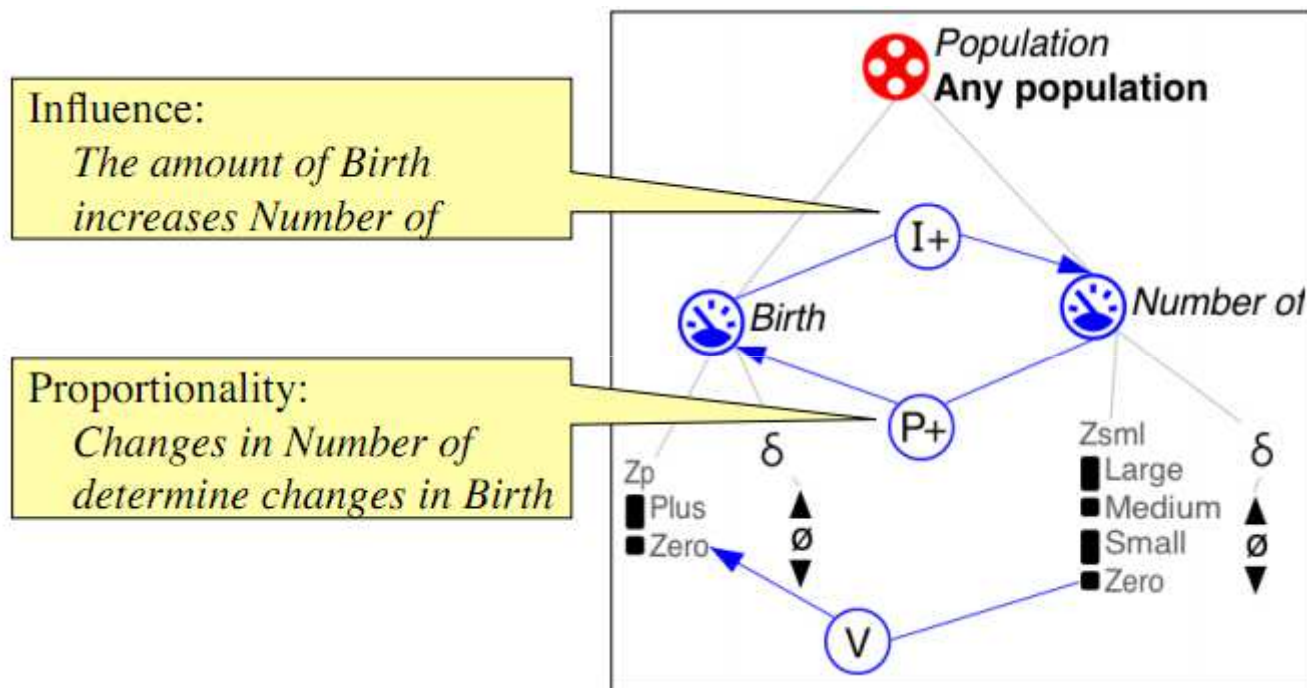


Scenarios (*starting situation*)

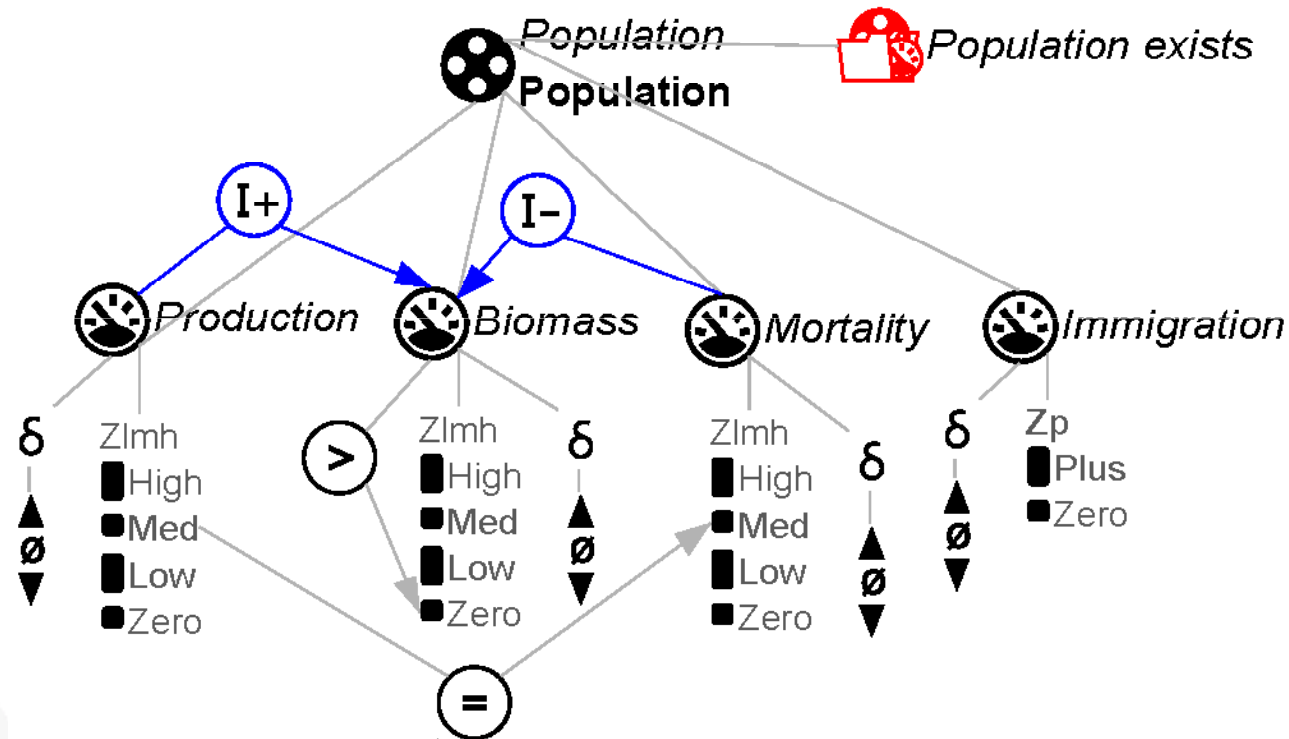


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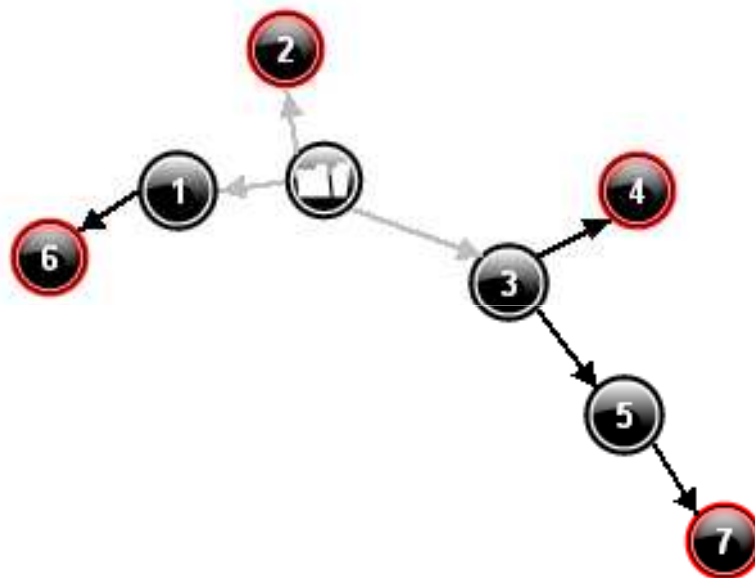
Model fragments (capturing partial knowledge)



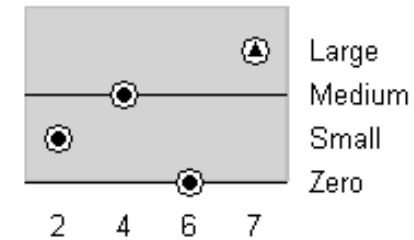
Model fragments (capturing partial knowledge)



Running simulations



Green frog: Biomass



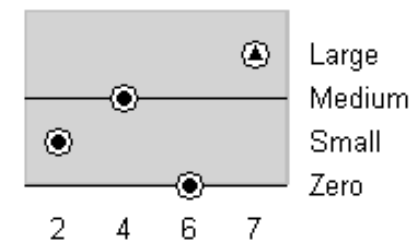
Green frog: Birth



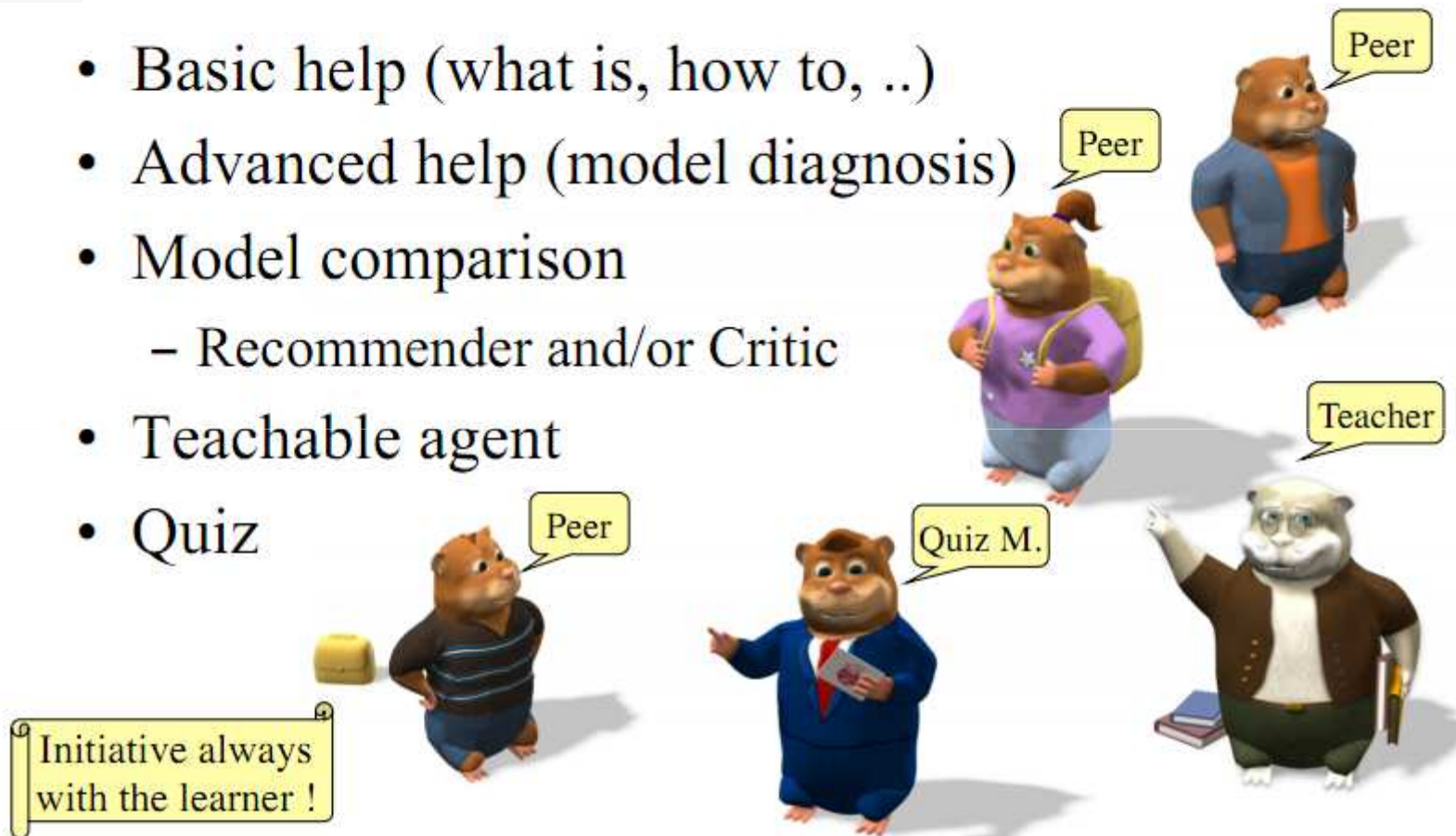
Green frog: Death

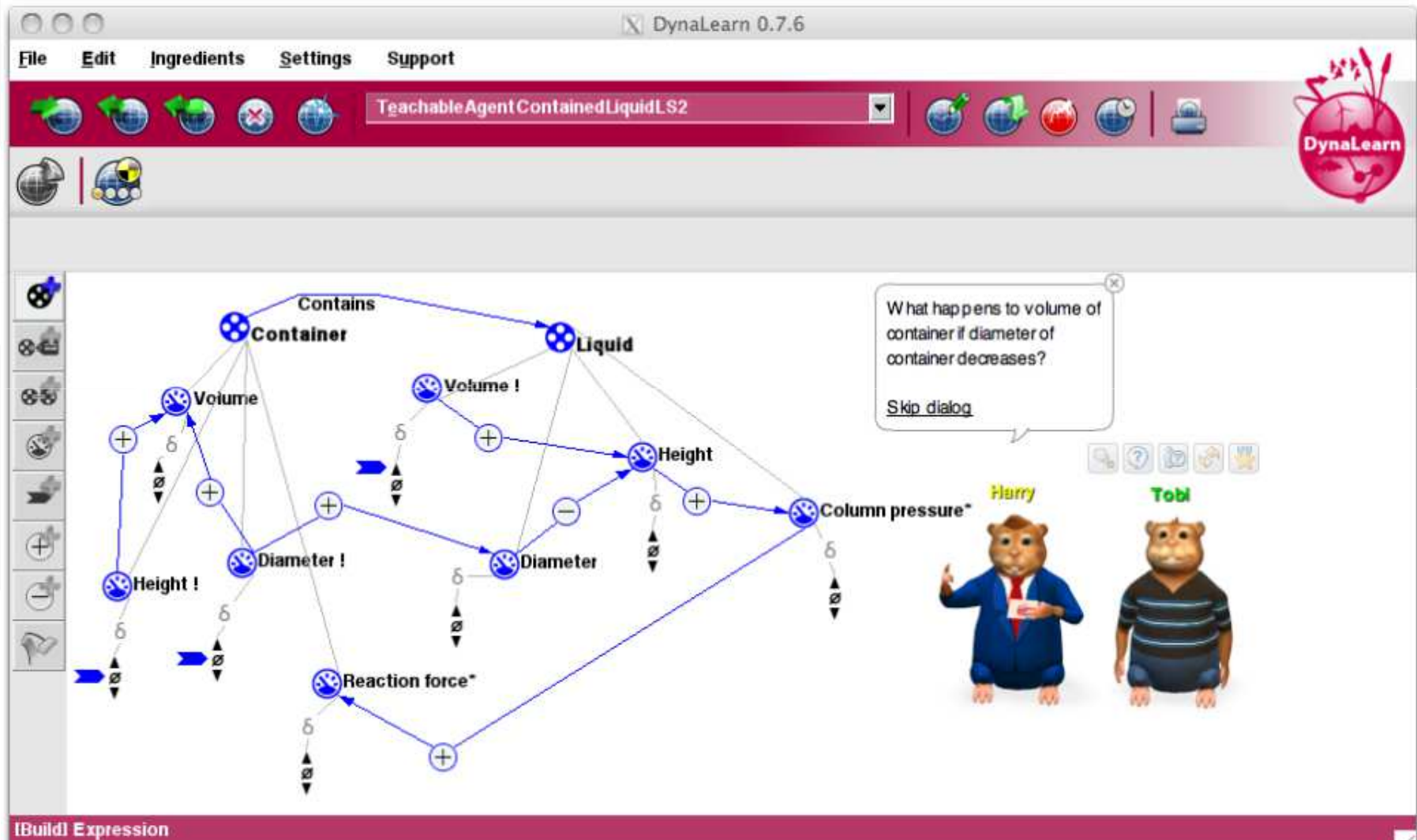


Green frog: Number of



- Basic help (what is, how to, ..)
- Advanced help (model diagnosis)
- Model comparison
 - Recommender and/or Critic
- Teachable agent
- Quiz

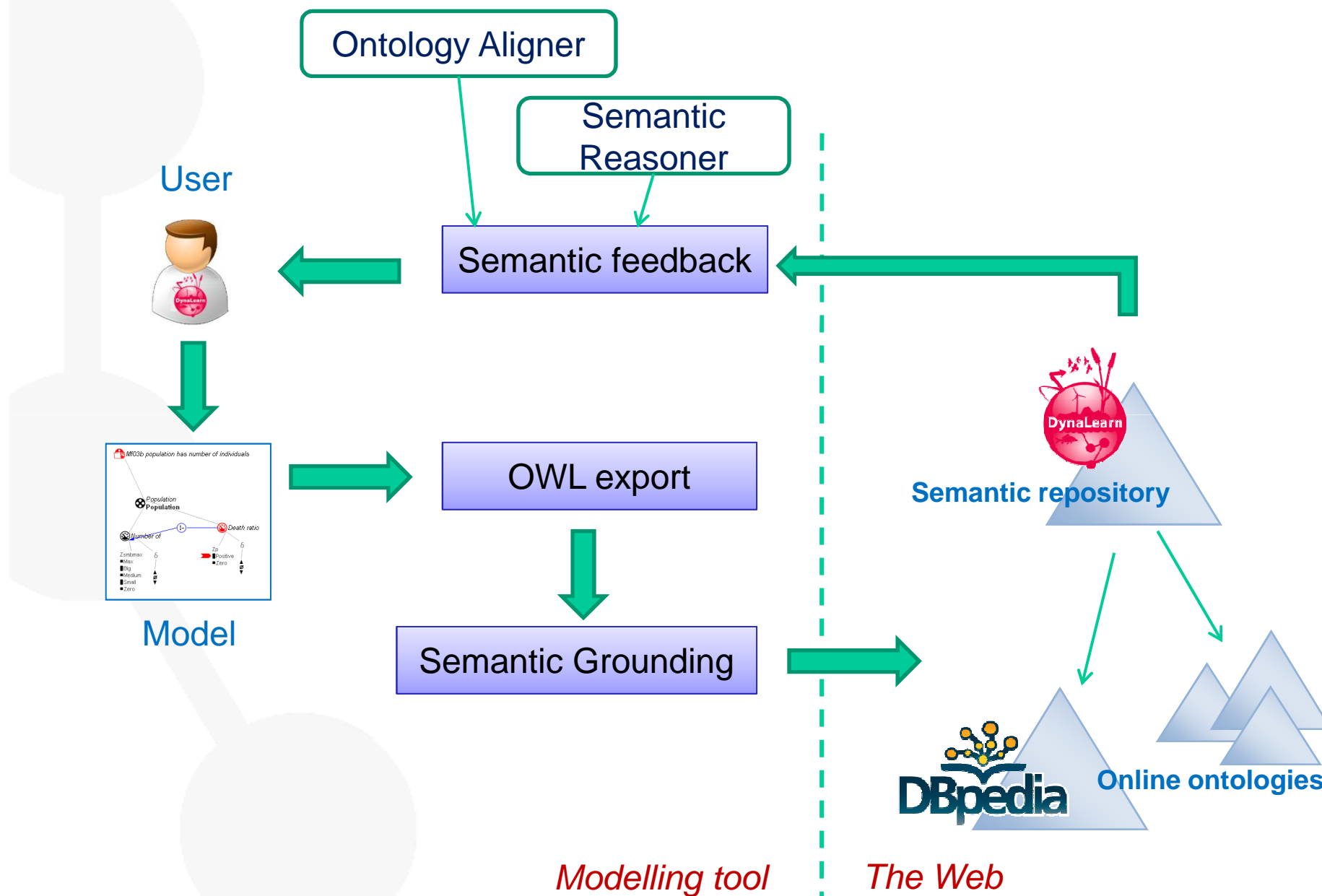




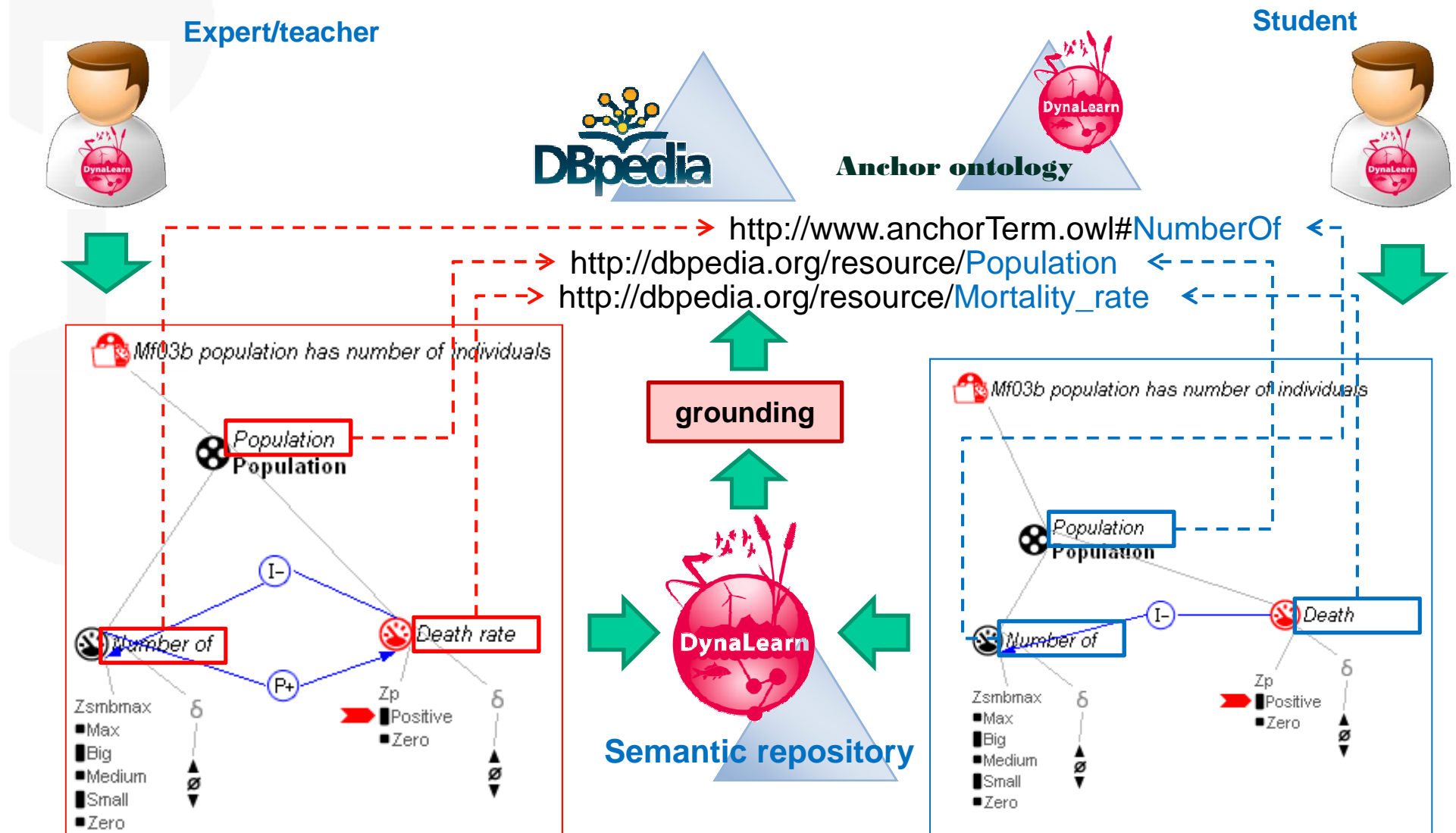
SEMANTIC TECHNIQUES

- To bridge the gap between the loosely and imprecise **terminology** used by a learner and the well-defined **semantics** of an ontology
- To put in **relation** to the QR models created by other learners or experts in order to automate the acquisition of **feedback** and **recommendations** from others

Semantic Techniques



Semantic Grounding



Benefits of grounding

- Support the process of **learning** a domain vocabulary
- Ensure lexical and semantic **correctness** of terms
- Ensure the **interoperability** among models
- Extraction of a common **domain knowledge**
- Detection of **inconsistencies** and contradictions between models
- Inference of **new, non declared, knowledge**
- Assist the model construction with **feedback** and **recommendations**

Semantic Grounding

The diagram on the left shows a semantic network. At the top is a red circle labeled "Population". Below it are three blue circles: "Production", "Biomass+", and "Mortality". Each of these blue circles has a vertical stack of three small squares labeled "High", "Med", and "Low", and a "Zero" label below them. Arrows point from "Population" to each of the three blue circles. The "Grounding editor" window on the right is titled "Grounding editor". It has a "Possible groundings:" list on the left, a "Mortality rate" section on the right, and a "Related models:" section at the bottom. The "Possible groundings:" list includes: Fish mortality, Mortality salience, **Mortality rate** (highlighted), Risk of mortality, Mortality displacement, Bills of Mortality, Child mortality, Perinatal mortality, Extraction from Mortalit, Force of mortality, Mortality drag, Infant mortality, Perioperative mortality, Compensation law of m, Morbidity and Mortality, Gompertz-Makeham la, Human mortality from H, Standardised mortality r, Age-standardized mortal, Mortality Medical Data, and Mortality (computability). The "Mortality rate" section contains a URL: http://dbpedia.org/resource/Mortality_rate and a paragraph of text: "Mortality rate is a measure of the number of deaths (in general, or due to a specific cause) in some population, scaled to the size of that population, per unit time. Mortality rate is typically expressed in units of deaths per 1000 individuals per year; thus, a mortality rate of 9.5 in a population of 100,000 would mean 950 deaths per year in that entire population." The "Related models:" section contains the text: "Mortality is ungrounded." and "Mortality will be grounded as Mortality rate." The window also has a "Synonyms:" section and a "Related models:" section.

Grounding editor

Possible groundings:

- Fish mortality
- Mortality salience
- Mortality rate**
- Risk of mortality
- Mortality displacement
- Bills of Mortality
- Child mortality
- Perinatal mortality
- Extraction from Mortalit
- Force of mortality
- Mortality drag
- Infant mortality
- Perioperative mortality
- Compensation law of m
- Morbidity and Mortality
- Gompertz-Makeham la
- Human mortality from H
- Standardised mortality r
- Age-standardized mortal
- Mortality Medical Data
- Mortality (computability)

Mortality rate

http://dbpedia.org/resource/Mortality_rate

Mortality rate is a measure of the number of deaths (in general, or due to a specific cause) in some population, scaled to the size of that population, per unit time. Mortality rate is typically expressed in units of deaths per 1000 individuals per year; thus, a mortality rate of 9.5 in a population of 100,000 would mean 950 deaths per year in that entire population.

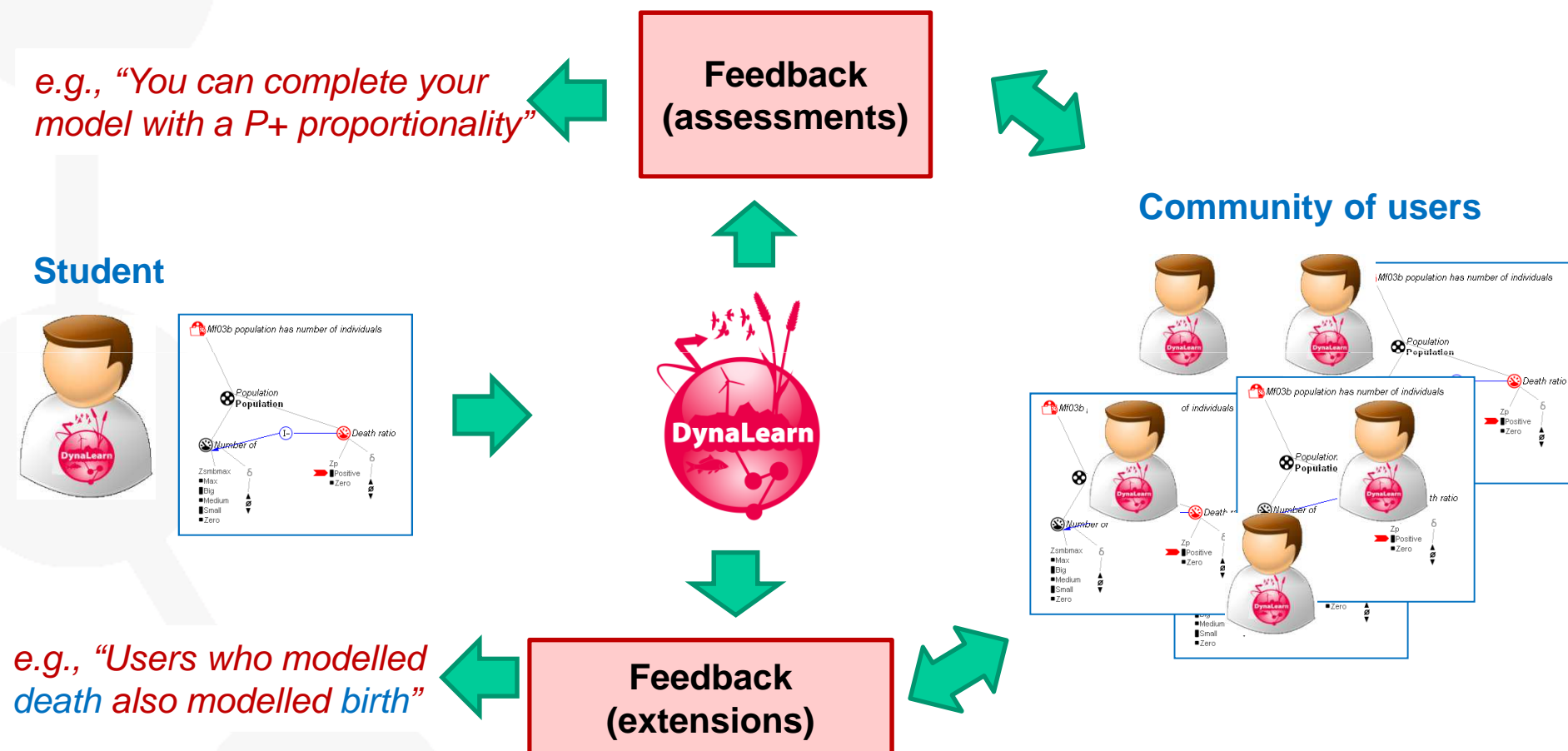
Related models:

Synonyms:

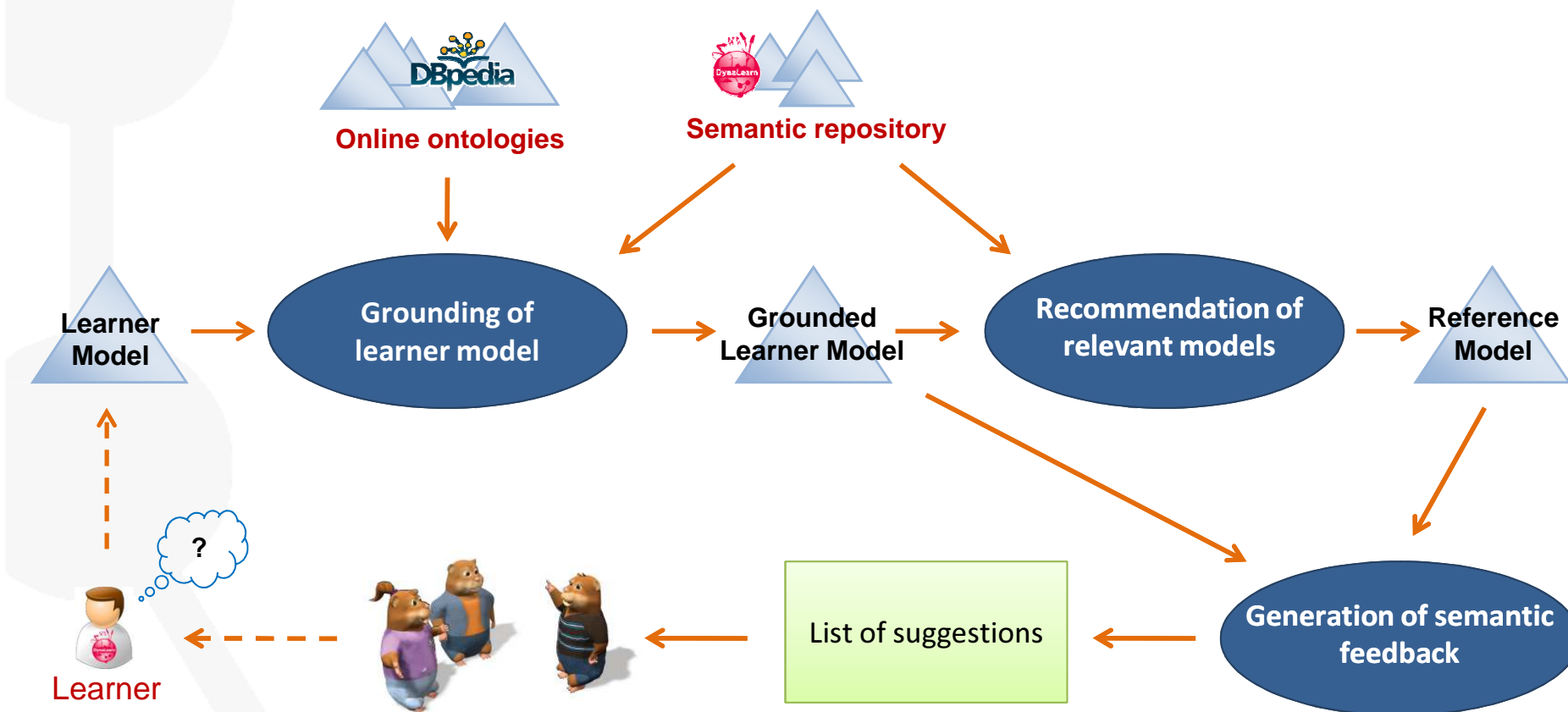
Mortality is ungrounded.

Mortality will be grounded as Mortality rate.

Semantic Feedback



Semantic Feedback



- Collaborative filtering for (subjective) community-driven recommendation. E.g.:
 - *“Users who liked model X also liked model Y”*
 - *“Users who modeled X also modeled Y”*
- Model-based. Recommendations based on the properties of the model under an objective perspective. E.g.:
 - *“Find a model fragment with inverse behaviour to the current one”*
 - *“List all models which are more specific than the current one”*

SOME RESEARCH QUESTIONS

- Q1. Which **external knowledge** source is the most suitable for grounding?
- Q2. Are the proposed groundings suitable according to **human opinion**?
- Q3. Are the state-of-the-art **ontology matching** tools suitable for the alignment of QR models?

SOME RESEARCH QUESTIONS

Q1. Which **external knowledge** source is the most suitable for grounding?

Q2. Are the proposed groundings suitable according to human opinion?

Q3. Are the state-of-the-art ontology matching tools suitable for the alignment of QR models?

Q1 – Semantic Web resources

Coverage Study

Tested **1686** different English words coming from DynaLearn glossaries

Knowledge source	Coverage Ratio
DBpedia	72%
OpenCyc	69%
WordNet	45%
Watson	47%

Behaviour of Dbpedia in other **languages**?

Labels in other languages for the covered English terms:

Language	ratio
English	100%
German	72%
Spanish	64%
Dutch	61%
Portuguese	58%

Q1 – Semantic Web resources

Coverage Study

What if we fix spelling errors and suggest nearby terms?: “fiter feeding” → “filter feeding”

Knowledge source	Coverage Ratio
DBpedia + Yahoo Spelling Suggestion	78%

What if we combine several sources?:

Knowledge source	Coverage Ratio
DBpedia + OpenCyc	87%
DBpedia + Watson	73%
Dbpedia + WordNet	72%
Dbpedia + OpenCyc + WordNet + Watson	88%

SOME RESEARCH QUESTIONS

Q1. Which **external knowledge** source is the most suitable for grounding?

***DBpedia** (combined with OpenCyc for English)*

Q2. Are the proposed groundings suitable according to human opinion?

Q2. Are the state-of-the-art ontology matching tools suitable for the alignment of QR models?

HULL'S EXPERIMENT

- Tested 909 English labels covered by DBpedia, randomly selected from DynaLearn glossaries
- Asked 8 expert evaluators, each one evaluated between 200-300 groundings. Each grounding was double-evaluated
- Question: *For each grounded term, are all its relevant meanings contained in the set of grounding candidates? If yes, mark the relevant ones.*

Accuracy (Average)	precision@1st (Average)
83%	76%

Inter-evaluator Agreement Level (Average)		
Agreem.1 (chosen grounding)	Agreem.2 (existence of suitable grounding)	Cohen's Kappa
78%	85%	0,47

SOME RESEARCH QUESTIONS

Q1. Which external knowledge source is the most suitable for grounding?

Dbpedia (combined with OpenCyc for English)

Q2. Are the proposed groundings suitable according to **human opinion**?

Yes... *(with an 83% accuracy)*

Q3. Are the state-of-the-art ontology matching tools suitable for the alignment of QR models?

MODEL MATCHING EXPERIMENT

- Reference Alignments:

Provided by [experts](#), manually aligned:

Case 1 - [Social aspects of population growth](#), v1 vs. v2

Case 2 - [Soil contamination](#), v1 vs. v2

Case 3 - [Hervibory](#) vs. Predation

Case 4 - [Amensalism](#) vs. Commensalism

- Ontology [alignment tools](#) tested: CIDER, Falcon

Q4 - Ontology Matching

	CIDER			Falcon		
	Precision	Recall	Time (s)	Precision	Recall	Time (s)
Case 1 (pop.)	1.00	1.00	10.8	0.80	1.00	1.9
Case 2 (soil)	1.00	1.00	9.2	0.79	1.00	2.0
Case 3 (h./p.)	1.00	1.00	4.2	0.63	1.00	1.8
Case 4 (a./c.)	0.67	0.80	4.6	0.44	0.80	1.7
AVERAGE	0.92	0.95	7.2	0.67	0.95	1.9

SOME RESEARCH QUESTIONS

Q1. Which external knowledge source is the most suitable for grounding?

Dbpedia (combined with OpenCyc for English)

Q2. Are the proposed groundings suitable according to human opinion?

Yes... (with an 83% accuracy)

Q3. Are the state-of-the-art **ontology matching** tools suitable for the alignment of QR models?

Yes

Some references

- [1] Bredeweg, B., Liem, J., Linnebank, F., Bühling, R., Wißner, M., Gracia, J., Salles, P., Beek, W. and Gómez Pérez, A. *DynaLearn: Architecture and Approach for Investigating Conceptual System Knowledge Acquisition*. In Intelligent Tutoring Systems 2010, Part II, pp. 272-274, LNCS 6095, Springer-Verlag.
- [2] Gracia, J., Liem, J., Lozano, E., Corcho, O., Trna, M., Gómez-Pérez, A., and Bredeweg, B. (2010). *Semantic Techniques for Enabling Knowledge Reuse in Conceptual Modelling*. Proc. of 9th International Semantic Web Conference (ISWC2010), Shanghai (China), Springer, volume 6414, November 2010. **Nominated as best in-use paper.**
- [3] DynaLearn web site, www.dynalearn.eu

Thanks for your attention!

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