Generating Knowledge Networks from Phenotypic Descriptions

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Phenotype Descriptions

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- Morphological structures
- Behavior traits
- Life cycles; etc.

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Examples:

- 1. No dark longitudinal stripes on head and body.
- 2. Scattered breast melanophores (Fuiman et al., 1983). Pteronotropis hubbsi can also be distinguished from Notropis chalybaeus by the presence of two caudal spots, one large spot centered at the base of the caudal fin below the flexed notochord and a smaller spot located dorsally above it, and by the presence of 9 dorsal rays in late metalarvae. Notropis chalybaeus has a single caudal spot in which no part extends above the notochord and 8 dorsal rays (Marshall, 1947).

Biology Knowledge Bases

¹http://www.fishbase.org

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 - e.g., FishBase: knowledge base about fishes¹

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 e.g., FishBase: knowledge base about fishes¹
- Identification Keys (IK)s
 - Artifacts to identify specimens
 - Observable characteristics

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Drawbacks:

- Need previous knowledge
- Need to follow the flow

Goal

To recognize and explicit phenotype elements locked in the Identification Keys. Using the Entity-Quality (EQ) representation:

- Entity: morphological structure
- Quality: qualifier state of the *Entity*

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Related Work

Information Extraction

Reference	Context	Approach
Ciaramita <i>et al.</i> , 2005	Interactions in molecular biology	Unsupervised Learning and Rules over Dependency Trees
Song <i>et al.</i> , 2015	Biomedical anatomic entities	Dictionary-based
Pyysalo and Ananiadou, 2014	Biomedical Anatomic entities	Supervised learning
Ramakrishnan <i>et al.</i> , 2008	Biomedical Anatomical entities	Dictionary-based, Rules over Dependencies Trees and Statis- tical Learning
Fundel et al., 2007	Gene and Protein Interaction	Rules over Dependency Trees
Cui, 2012	Morphological structures of or- ganisms	Unsupervised Learning

General View



Step 1:

It explores isolated sentences

Step 2:

It explores the sentence correlations

Method Step 1 - General View

Assumption:

The typical way in which phenotype descriptions are written can guide the extraction of EQ elements.

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Step 1 - Match Algorithm

Iteration 3 **Iteration 1 Iteration 2** Vertexes: no dorsal fin Vertexes: fin Vertexes: dorsal fin Concept: dorsal fin Concept: dorsal fin Concept: fin Similarity: 1 Similarity: 0.76 Similarity: 1 fin fiņ Dependency Tree dorsal No dorsal No dorsal No fin fin fin **Teleost Anatomy** dorsal dorsal dorsal anal anal anal Ontology fin fin fin fin fin fin (TAO)

Identifying Entities and Qualities:

Method Step 1 - Output



Step 2 - General View

Assumption:

The structure of Identification Keys holds correlations that can be exploited to improve the extraction of EQ statements.

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Assumption:

The structure of Identification Keys holds correlations that can be exploited to improve the extraction of EQ statements.

Generally, in phenotype descriptions:

- 1. Alternative sentences refer to the same *Entities*.
- 2. Alternative sentences assign complementary *Qualities* to *Entity*.



Step 2 - Algorithm



Step 2 - Algorithm



Compare the two relations, based on:

- (a) Existence of antonymy between the quality parts
- (b) Relation Type
- (c) Grammatical classes of quality parts
- (d) Relation Directions

Step 2 - Algorithm



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Similarity
$$= \sum_{i=a}^{d} v_i$$

Method Step 2 - Output



Evaluation - Numerical Assessment

Gold Standard-based Assessment

Gold standard set: 100 phenotype descriptions (randomly selected) were manually annotated

Elements Measures	EQ pair	Entity
Recall	0,45	0,76
Precision	0,87	0,94
F-measure	0,59	0,84

Evaluation - Application Experiments

EQ sharing through taxons



Figure 1: Bipartite network of Species and EQs

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Evaluation - Application Experiments

EQ sharing through taxons



Figure 2: Projection of bipartide network

Conclusion

Original approach to automatically recognize ${\it Entities}$ and ${\it Qualities},$ exploring :

- Writing characteristics of phenotype descriptions
- Organizational structure of IKs

Future Work

- To compare against other approaches
- To recognize complete EQs in Step 2 (not only the quality part)
- To calibrate the parameters and thresholds

Thank you!

Classical Measures

$$Recall = \frac{TP}{TP + FN}$$
(1)

$$Precision = \frac{TP}{TP + FP}$$
(2)

$$F\text{-}measure = \frac{2*Precision*Recall}{Precision+Recall}$$
(3)

Examples of:

- True Positive:
 - expected: E[lips]Q[notfringed]
 - recognized: E[lips]Q[notfringed]
- False Positive:
 - expected E[vertebrae]Q[119 to 132]
 - recognized: E[vertebrae]Q[132]
- False Negative:
 - recognized E[breastmelanophores]Q[Scattered]

Considering Partial Matches

- Complete Miss (CM): false negative
- Wrong Hit (WH): false Positive
- Full Match (FM): true Positive

$$Partial \ Precision = \frac{Partial \ Match}{Full \ Match + Partial \ Match + Wrong \ Hit}$$
(4)

$$Full Precision = \frac{Full Match}{Full Match + Partial Match + Wrong Hit}$$
(5)

 $Full Recall = \frac{FullMatch}{Full Match + Partial Match + Complete Miss}$ (7)

Considering Partial Matches Total Precision = Partial Precision + Full Precision Total Recall = Partial Recall + Full Recall

Elements Measures	EQ pair	Entity
Partial-Recall	0.05	0.08
Full-Recall	0.39	0.67
Partial-Precision	0.11	0.1
Full-Precision	0.75	0.84

Table 1: Results concerning Perfect and also Partial Matches

Elements Measures	EQ pair	Entity
Total Recall	0,45	0,76
Total Precision	0,87	0,94
Total F-measure	0,59	0,84