

Query Learning and Attribute Exploration



Supervised learning

Input: a training set divided into (for example) two classes w. r. t. a certain target property:

- positive examples;
- negative examples.

Build a classifier that determines whether a previously unseen object has the target property.



Learning with queries (Angluin 1988)

Input: an oracle capable of answering queries of certain predefined types concerning a target property.

Build a classifier that determines whether a previously unseen object has the target property.



Types of queries

Membership query: Does the object have the target property?



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Membership query: Does the object have the target property?

Equivalence query: Does the hypothesis H accurately describe the set of objects with the target property? If not, the oracle must provide

- a **positive counterexample** that has the target property, but is not covered by the hypothesis

or

- a **negative counterexample** that doesn't have the property, but satisfies the hypothesis.



Types of queries

Subset query: Does the hypothesis H describe **only** objects with the target property?

- If not, provide a negative counterexample.

Superset query: Does the hypothesis H describe **all** the objects with the target property?

- If not, provide a positive counterexample.



Learning binary patterns

- A **binary pattern** is a nonempty string consisting of 0, 1, and variables from a countably infinite alphabet X .
- A pattern p defines the **language** $L(p)$ consisting of all words that can be obtained by substituting nonempty binary strings for variables.
- For example, the language of the pattern $x_1 0 x_2 0 x_1$ includes the strings **10101** and **010111001**, but not **10100** or **1001**.



Learning binary patterns

- A hint: I have guessed one of the 2^k variable-free patterns of length k .



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 - No.



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 - No: give a negative counterexample.



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- Membership query: Is word w in the language?
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- Subset query: Is p less general than the target pattern?
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Learning binary patterns

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- Membership query: Is word w in the language?
 - No.
- Equivalence query: Is p the right pattern?
 - No: give a negative counterexample.
- Subset query: Is p less general than the target pattern?
 - No: give a negative counterexample.
- Each of the above queries excludes at most one of 2^k patterns.

No polynomial-time algorithm.



Learning binary patterns with superset queries

1. Determine the length k of the pattern.
2. Determine the constants in the pattern.
3. Determine identical variables in the pattern.



Learning binary patterns with superset queries

1. Determine the length k of the pattern.
 - Query about patterns x_1x_2 , $x_1x_2x_3$, etc., until the answer is “No”.
2. Determine the constants in the pattern.
3. Determine identical variables in the pattern.



Learning binary patterns with superset queries

1. Determine the length k of the pattern.
 - Query about patterns $x_1x_2, x_1x_2x_3$, etc., until the answer is “No”.
2. Determine the constants in the pattern.
 - For $i = 1, 2, \dots, k$, query about the patterns $x_1 \dots x_{i-1} 0 x_{i+1} \dots x_k$ and $x_1 \dots x_{i-1} 1 x_{i+1} \dots x_k$.
3. Determine identical variables in the pattern.



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3. Determine identical variables in the pattern.
 - For each pair of positions $i < j$ of variables in the pattern, query about the pattern $x_1 \dots x_{i-1} y x_{i+1} \dots x_{j-1} y x_{j+1} \dots x_k$.



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Polynomial-time algorithm!



Formal Concept Analysis

Formal context (G, M, I)

- a set G of objects
- a set M of attributes
- objects are described with attributes via a binary relation
 $I \subseteq G \times M$



A formal context

Europe	EU	Euro	Schengen	NATO	Monarchy
Italy	×	×	×	×	
United Kingdom	×			×	×
Poland	×		×	×	
Denmark	×		×	×	×
Norway			×	×	×
Russia					
Spain	×	×	×	×	×
Turkey				×	



Formal Concept Analysis

Derivation operators

For $A \subseteq G$ and $B \subseteq M$:

- $A' = \{m \in M \mid \forall g \in A: glm\}$
- $B' = \{g \in G \mid \forall m \in B: glm\}$

For $g \in G$ and $m \in M$, the set $\{g\}'$ is called an **object intent** and the set $\{m\}'$ is called an **attribute extent**.

$(\cdot)'' : 2^M \rightarrow 2^M$ is a closure operator.



Derivation operators

Europe	EU	Euro	Schengen	NATO	Monarchy
Italy	×	×	×	×	
United Kingdom	×			×	×
Poland	×		×	×	
Denmark	×		×	×	×
Norway			×	×	×
Russia					
Spain	×	×	×	×	×
Turkey				×	

$\{\text{EU, Euro, Schengen}\}' = \{\text{Italy, Spain}\}$



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Denmark	×		×	×	×
Norway			×	×	×
Russia					
Spain	×	×	×	×	×
Turkey				×	

$\{\text{Italy, Spain}\}' = \{\text{EU, Euro, Schengen, NATO}\}$



Formal Concept Analysis

Implication $A \rightarrow B$

$(A, B \subseteq M)$

- An attribute subset $X \subseteq M$ is a **model** of $A \rightarrow B$ if A is not a subset of X or B is a subset of X .
- $A \rightarrow B$ **holds** in the context if $A' \subseteq B'$.
- X is a **model** of an implication set L if it is a model of every implication from L .
- Two implication sets are **equivalent** if they have the same models.
- Among equivalent implication sets, there is a particular one called the **canonical** (or Duquenne–Guigues) **basis**. It is minimal in the number of implications among all equivalent implication sets.



Canonical basis

Europe	EU	Euro	Schengen	NATO	Monarchy
Italy	×	×	×	×	
United Kingdom	×			×	×
Poland	×		×	×	
Denmark	×		×	×	×
Norway			×	×	×
Russia					
Spain	×	×	×	×	×
Turkey				×	

EU → NATO

Monarchy → NATO

Schengen → NATO

Euro → EU, Schengen, NATO



Computing implications

- If M is finite, the canonical basis is finite, too.
- If G is also finite, we can compute the canonical basis, e.g., using the NextClosure algorithm (Ganter 1984).
- However, if G is infinite, it is not possible to work with the entire context directly.



Learning implications with queries

- Forget about the context—talk to an oracle to compute the implication set L^* .
- A set of implications describes a set of *models*— attribute combinations that satisfy these implications.

Membership queries: Is $A \subseteq M$ a model of L^* ?

Equivalence queries: Is an implication set L equivalent to L^* ?

- If not, the oracle must provide a counterexample: a set that is a model of L^* , but not of L (**positive counterexample**), or vice versa (**negative counterexample**).



Polynomial-time algorithm (Angluin *et al.* 1992)

Equivalence queries: Is an implication set L equivalent to L^* ?

- If not, the oracle must provide a counterexample: a set that is a model of L^* , but not of L (**positive counterexample**), or vice versa (**negative counterexample**).

How to handle a counterexample X ?

If X is **positive** and it doesn't satisfy some $A \rightarrow B$ from L , weaken $A \rightarrow B$ by replacing it with $A \rightarrow B \cap X$.

If X is **negative**, strengthen L by replacing some $A \rightarrow B$ from L with $A \cap X \rightarrow B$ or by adding a new implication $X \rightarrow M$, so as to exclude X from the set of models of L .



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Initial hypothesis: the **empty** implication set—everything is possible!
- Equivalence query returns a **negative** counterexample $\{a, b, c\}$.
- New hypothesis: $\{a, b, c\} \rightarrow M$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a, b, c\} \rightarrow M$.
- Equivalence query returns a **negative** counterexample $\{a\}$.
- **Strengthen** the current hypothesis to exclude $\{a\}$.
- The new hypothesis: $\{a\} \rightarrow M$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow M$.
- Equivalence query returns a **negative** counterexample $\{c, d\}$.
- Can we **strengthen** $\{a\} \rightarrow M$?



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow M$.
- Equivalence query returns a **negative** counterexample $\{c, d\}$.
- Can we **strengthen** $\{a\} \rightarrow M$?
- **Membership** query w. r. t. $\{a\} \cap \{c, d\} = \emptyset$.
- Answer: **yes!**
 - The empty set is a model of L^* .



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow M$.
- Equivalence query returns a **negative** counterexample $\{c, d\}$.
- Can we **strengthen** $\{a\} \rightarrow M$?
- Membership query w. r. t. $\{a\} \cap \{c, d\} = \emptyset$.
- Answer: **yes!**
 - The empty set is a model of L^* .
- New hypothesis: $\{a\} \rightarrow M, \{c, d\} \rightarrow M$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow M$, $\{c, d\} \rightarrow M$.
- Equivalence query returns a **positive** counterexample $\{a, c, d\}$.
- **Weaken** $\{a\} \rightarrow M$ and $\{c, d\} \rightarrow M$.
- New hypothesis: $\{a\} \rightarrow \{a, c, d\}$, $\{c, d\} \rightarrow \{a, c, d\}$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow \{a, c, d\}$, $\{c, d\} \rightarrow \{a, c, d\}$.
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- **Weaken** $\{a\} \rightarrow \{a, c, d\}$.
- New hypothesis: $\{a\} \rightarrow \{a, c\}$, $\{c, d\} \rightarrow \{a, c, d\}$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow \{a, c\}, \{c, d\} \rightarrow \{a, c, d\}$.
- Equivalence query returns a **negative** counterexample $\{a, b, c\}$.
- **Membership** query w. r. t. $\{c, d\} \cap \{a, b, c\} = \{c\}$.
- Answer: **yes!**
- New hypothesis: $\{a\} \rightarrow \{a, c\}, \{c, d\} \rightarrow \{a, c, d\}, \{a, b, c\} \rightarrow M$.



Polynomial-time algorithm (Angluin *et al.* 1992)

An example for $M = \{a, b, c, d\}$:

- Current hypothesis: $\{a\} \rightarrow \{a, c\}, \{c, d\} \rightarrow \{a, c, d\}, \{a, b, c\} \rightarrow M$.
- Equivalence query returns **no** counterexamples.
- **Success!**



Polynomial-time algorithm (Angluin *et al.* 1992)

$\mathcal{L} := \emptyset$

```
while there is a counterexample  $X$  to  $\mathcal{L}$  do                                {Equivalence oracle}
  if  $\mathcal{L}(X) = X$  then                                                {negative counterexample}
    found := false
    for all  $A \rightarrow B \in \mathcal{L}$  do
       $C := A \cap X$ 
      if  $A \neq C$  and  $C \neq \mathcal{L}^*(X)$  then                                {Membership oracle}
         $\mathcal{L} := \mathcal{L} \setminus \{A \rightarrow B\}$ 
         $\mathcal{L} := \mathcal{L} \cup \{C \rightarrow B\}$ 
        found := true
        exit for
      if not found then
         $\mathcal{L} := \mathcal{L} \cup \{X \rightarrow M\}$ 
  else                                                                    {positive counterexample}
    for all  $A \rightarrow B \in \mathcal{L}$  such that  $A \subseteq X$  and  $B \not\subseteq X$  do
       $\mathcal{L} := \mathcal{L} \setminus \{A \rightarrow B\}$ 
       $\mathcal{L} := \mathcal{L} \cup \{A \rightarrow B \cap X\}$ 
```



Polynomial-time algorithm (Angluin *et al.* 1992)

- Computes the canonical basis (Arias and Balcázar 2011)
- Makes $O(m^2n)$ membership and $O(mn)$ equivalence queries
 - m is the size of the basis
 - n is the number of attributes



Attribute exploration

- An alternative technique from formal concept analysis.
- Start with any (possibly empty) set of objects.
- Generate an implication valid in the current subcontext.
- If the implication is not valid in the entire context, provide an object that violates it.
- Go to the next implication, etc.

Follow the canonical basis to ask only questions that are necessary.



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- Generate an implication valid in the current subcontext.
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This is a superset query:

Do the models of the implication $A \rightarrow B$ include all the models of the target implication set L^* ?



Attribute exploration in lexical typology

- Objects are words of different languages.
- Attributes are frames corresponding to individual meanings.
- We want to build a semantic map showing which meanings can be combined together within a single lexeme.



Attribute exploration of the semantic field 'empty'

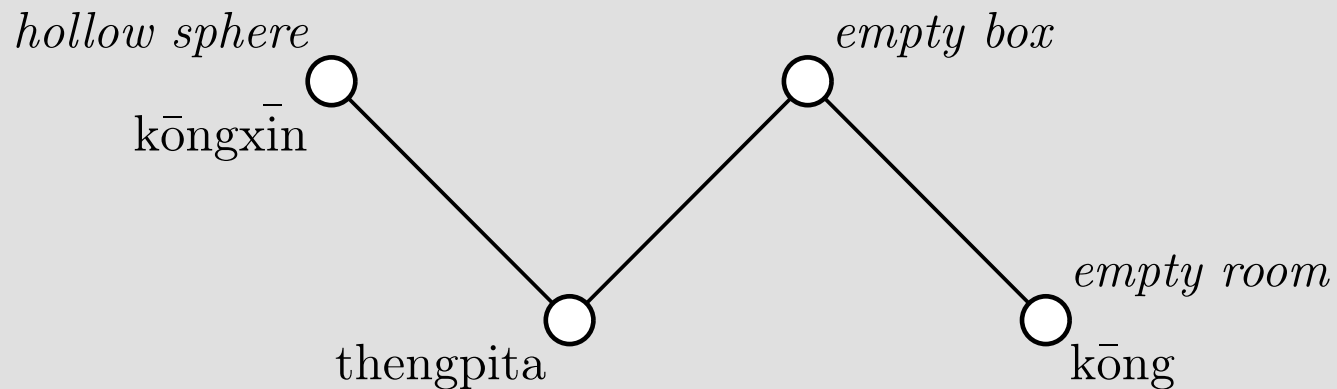
- Start with three words (two Chinese and one Korean) and three frames:

	<i>hollow sphere</i>	<i>empty box</i>	<i>empty room</i>
kōngxin	×		
kōng		×	×
thengpita	×	×	



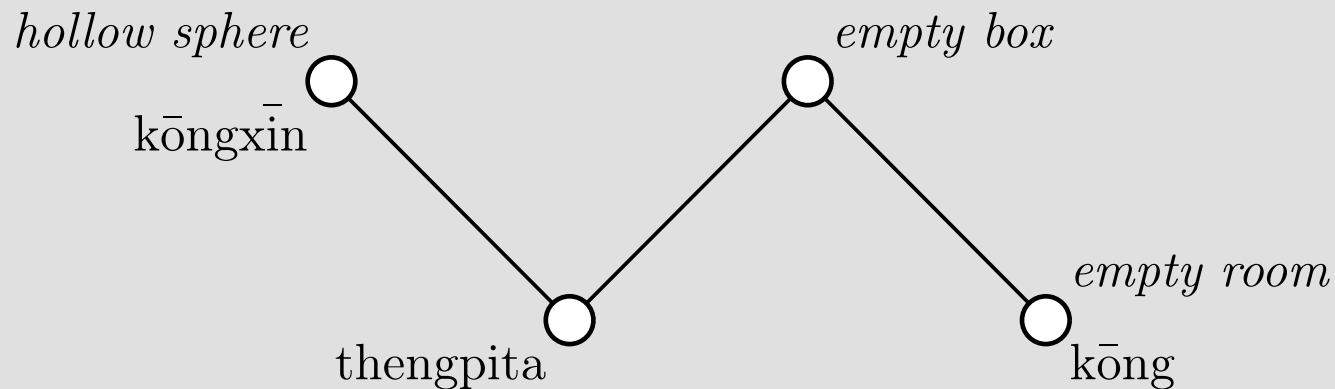
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Attribute exploration of the semantic field 'empty'

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Does every word for *empty room* is also suitable for *empty box*?



Attribute exploration of the semantic field 'empty'

Does every word for *empty room* is also suitable for *empty box*?

No: there is a counterexample in Korean.

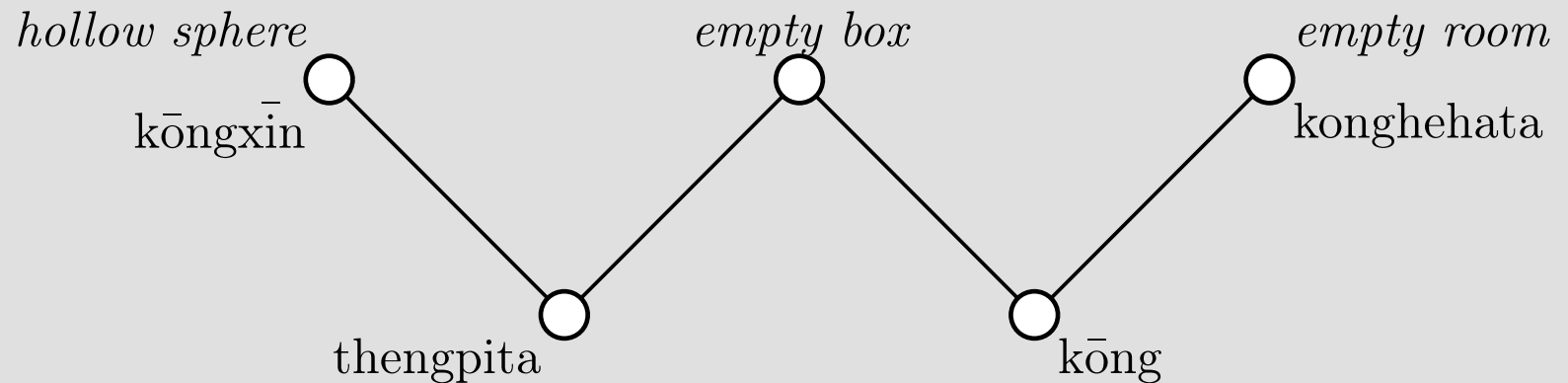
	<i>hollow sphere</i>	<i>empty box</i>	<i>empty room</i>
kōngxin	×		
kōng		×	×
thengpita	×	×	
konghehata			×



Attribute exploration of the semantic field 'empty'

Does every word for *empty room* is also suitable for *empty box*?

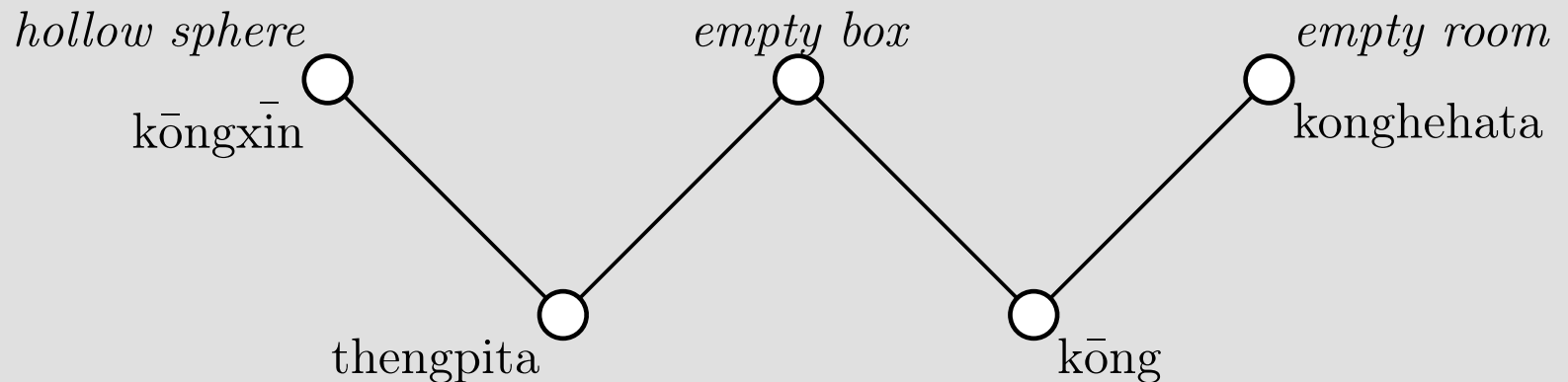
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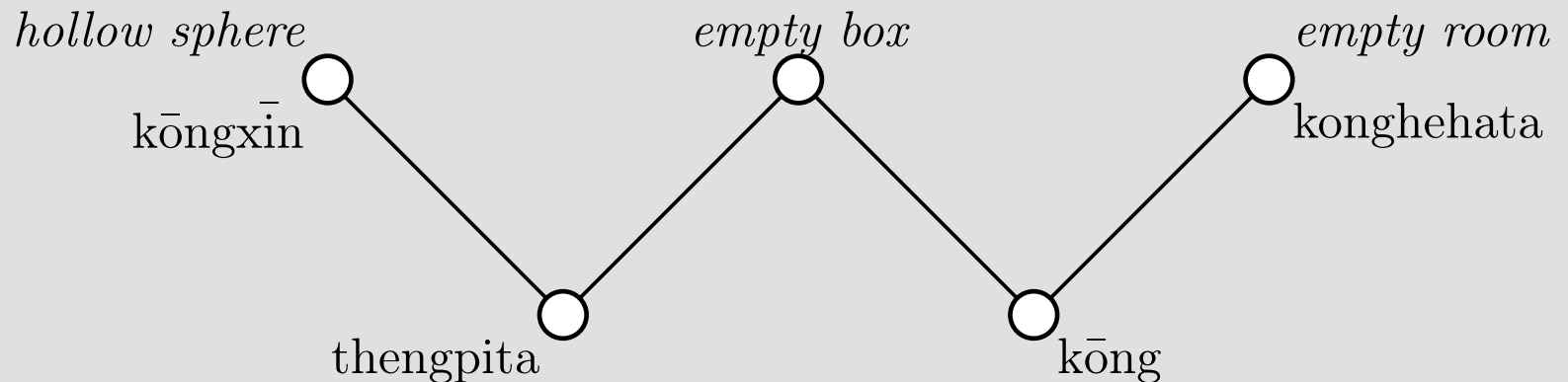
Is there a word used for both *hollow sphere* and *empty room*?



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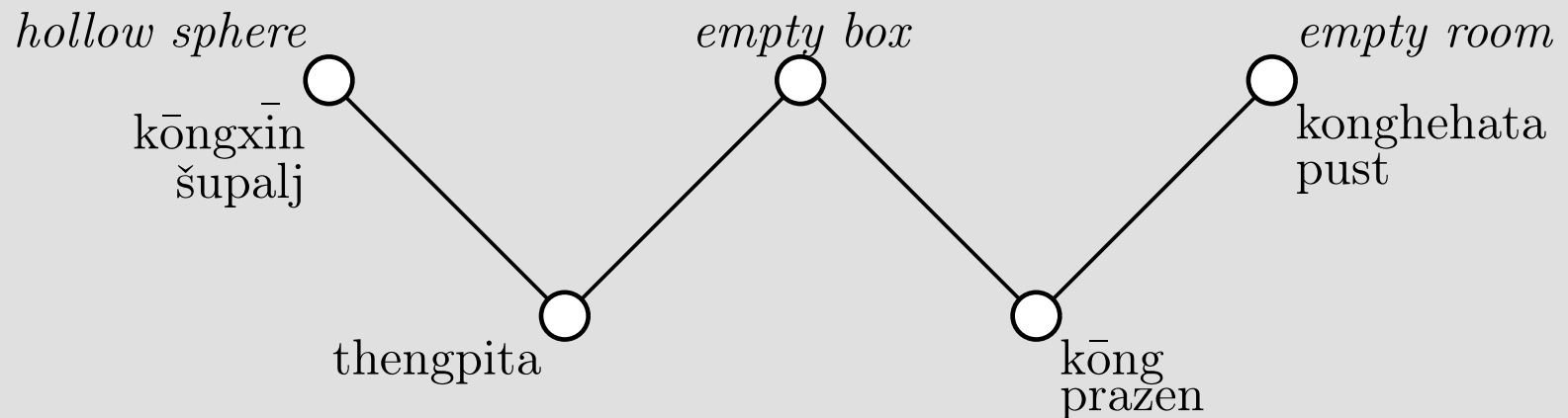
Is there a word used for both *hollow sphere* and *empty room*?

Probably, no.



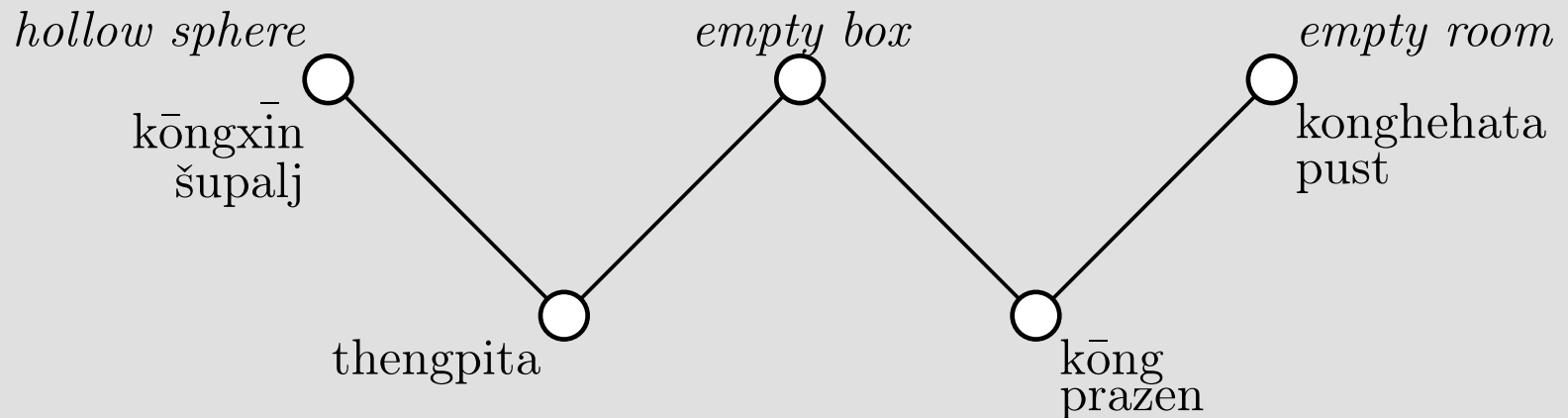
Attribute exploration of the semantic field 'empty'

Let us add some Serbian words:



Object exploration of the semantic field 'empty'

Let us add some Serbian words:



and run attribute exploration the other way round:

Are all the meanings of "pust" shared by "kōng",
"konghehata", and "prazen"?



Object exploration of the semantic field 'empty'

Are all the meanings of "pust" shared by "kōng",
"konghehata", and "prazen"?

No: "prazen" is not used to denote local spaces without people
(but only those without inanimate objects).

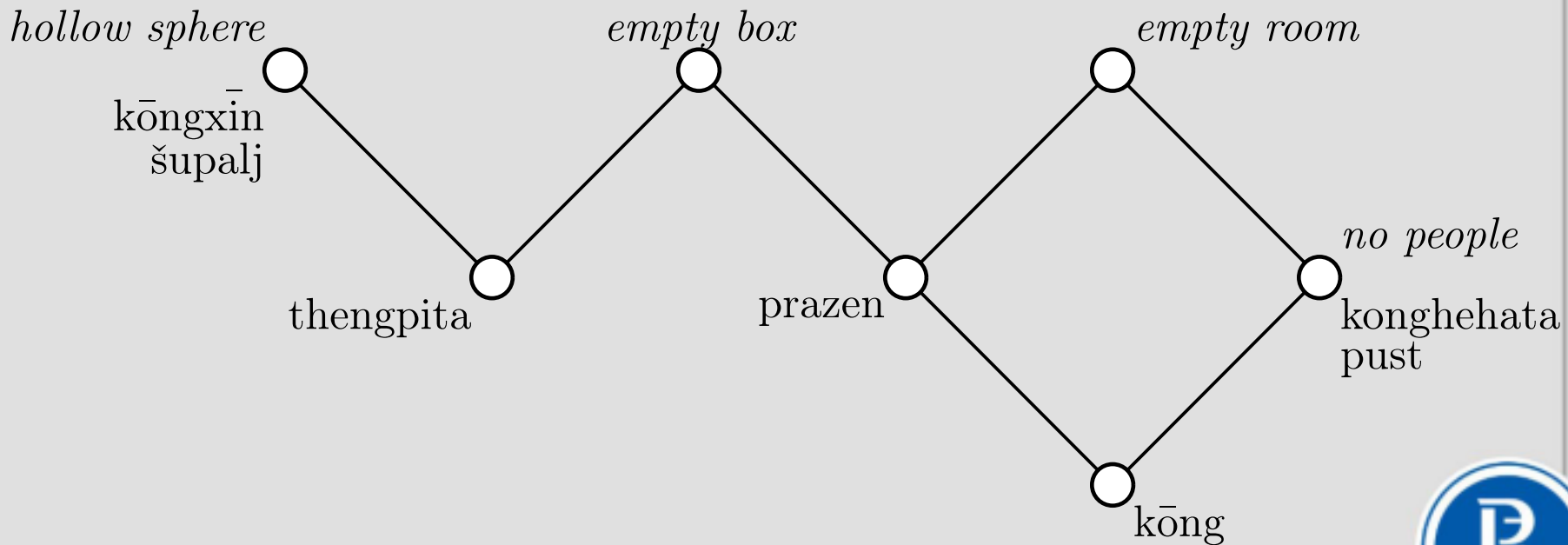
	<i>no people</i>
kōngxin	
kōng	×
thengpita	
konghehata	×
šupalj	
prazen	
pust	×



Object exploration of the semantic field 'empty'

Are all the meanings of "pust" shared by "kōng",
"konghehata", and "prazen"?

No: "prazen" is not used to denote local spaces without people
(but only those without inanimate objects).



Many possible extensions

- Background knowledge
- Exceptions
- Symmetries
- Incompletely specified examples

- First-order rule exploration
- Exploration for description logics
- Concept exploration
- Exploration of noisy, fuzzy, triadic data, etc.

- Collaborative exploration

