

LTN Examples and Code

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Logic Tensor Networks

- ▶ <https://github.com/logictensornetworks/>
- ▶ git clone
<https://github.com/logictensornetworks/logictensornetworks.git>

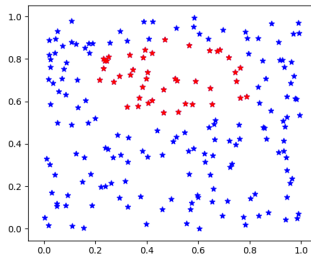
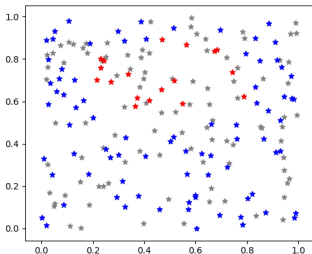
Classification

Domain A set D of points in the $[0, 1]^2$ square;

Predicate A unary predicate A (class)

- Supervisions**
1. **Positive examples** a set of points in D , which are known to be instances of A ;
 2. **Negative examples** a set of points in D , which are known not to be instances of A ;

Task For all the other point in D determine if they are instances of A .



Classification in LTN

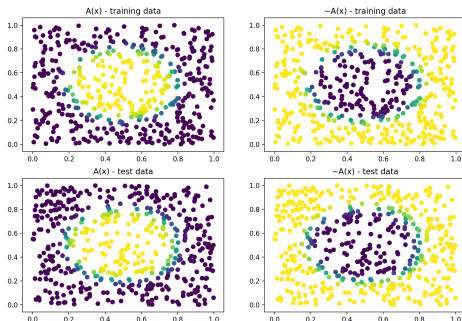
D = random subset of $[0, 1]^2$

$P = \{p \in D \mid p \text{ is a positive example of } A\}$

$N = \{p \in D \mid p \text{ is a negative example of } A\}$

$\forall x \in P : A(x)$

$\forall x \in N : \neg A(x)$



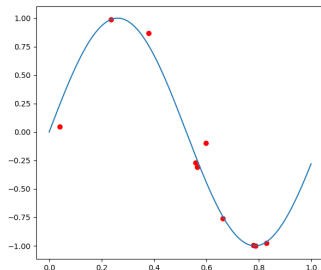
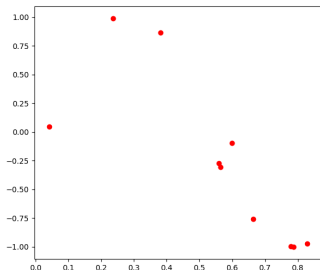
Regression

Domain A set $D = [0, 1]$

Function A unary function $f : D \rightarrow D$

Supervisions A set of **supervision pairs** $S = \{\langle x_1, y_1 \rangle, \dots, \langle x_n, y_n \rangle\}$ such that $y_i = f(x_i)$

Task For all the other points $x \in D$ predict $f(x)$

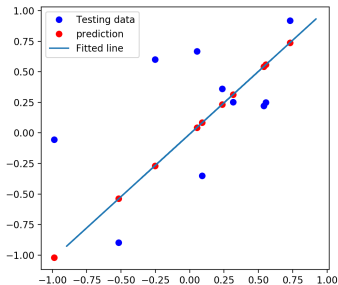
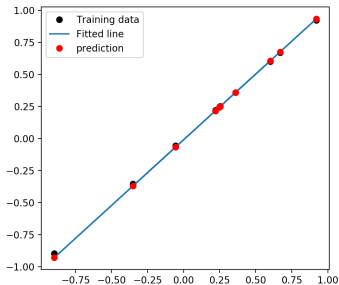


Regression in LTN

D = a uniform sampling of $[0, 1]$

S = $\{\langle x_1, y_1 \rangle, \dots, \langle x_n, y_n \rangle\}$

$$\forall \langle x, y \rangle \in S : f(x) = y$$



Multi-Label Classification with Label Constraints

Domain A set D of points in the $[0, 1]^2$ square;

Predicate A set unary predicate A_1, \dots, A_n (class/labels)

Supervisions

1. **Positive examples for A_i** a set of points in D , which are known to be instances of A_i ;
2. **Negative examples for A_i** a set of points in D , which are known not to be instances of A_i ;

Label constraints

- **Subset constraint** if x is labelled with A_i then it is labelled with A_j ;
- **Disjoint constraint** x is labelled with A_i iff then x is labelled with $\neg A_j$ and viceversa;

Task For all the points in D determine if they are labelled with A_i or $\neg A_i$ (Notice that it is possible that x is labelled neither with A_i nor $\neg A_i$)

Multi-Label Classification with Label Constraints in LTN

$$\forall x \in A_i^+ : A_i(x)$$

$$\forall x \in A_i^- : \neg A_i(x)$$

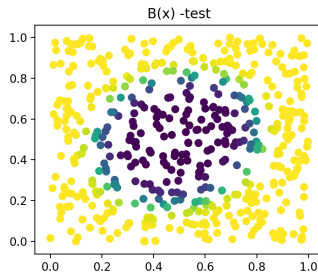
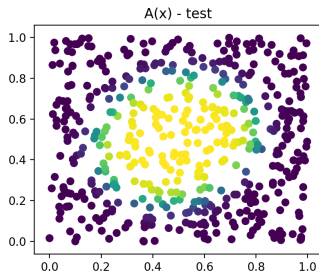
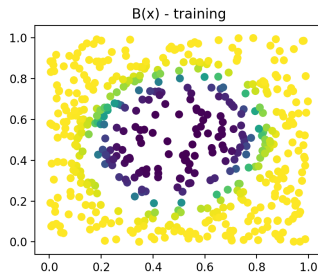
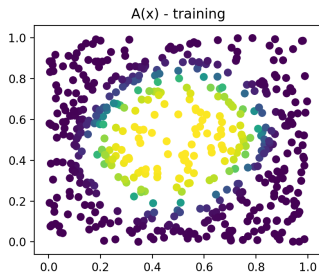
$$\forall x \in D : A_i(x) \rightarrow A_j(x)$$

$$\forall x \in D : A_i(x) \leftrightarrow \neg A_j(x) \wedge \neg A_i(x) \leftrightarrow A_j(x)$$

Notice that FOL allows to express more general constraints between labels, as for instance

$$\forall x \in D : (A(x) \wedge B(x)) \rightarrow (C(x) \vee D(x))$$

Multi-Label Classification with Label Constraints in LTN



Unsupervised learning - Clustering

Domain A set $D \subset [0, 1]$

Cluster labels A set of cluster labels C_1, \dots, C_n

Task Put each point in some cluster C_i minimizing intra-cluster distance and maximizing inter-cluster distance

Clustering Constraints

- ▶ Every point belongs to a cluster

$$\forall x : C_1(x) \vee \dots \vee C_n(x)$$

- ▶ Every point cannot belong to two clusters

$$\forall x : \neg(C_i(x) \wedge C_j(x)) \quad \text{for } i \neq j \in \{1, \dots, n\}$$

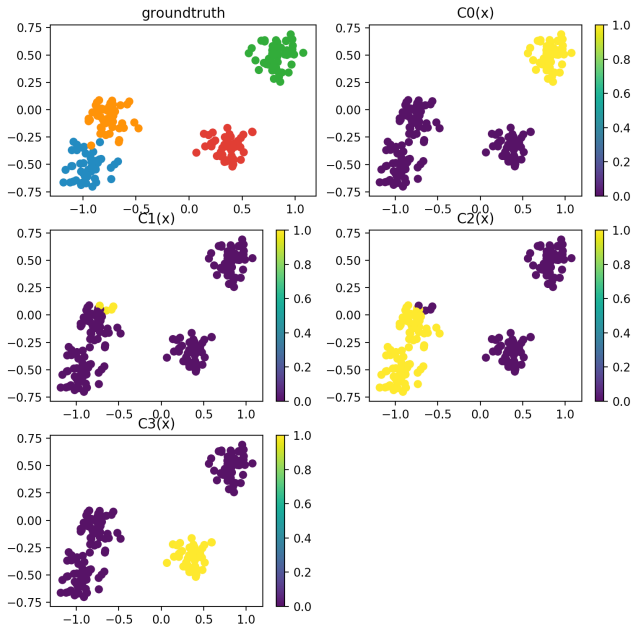
- ▶ Clusters are not empty

$$\exists x : C_i(x) \quad \text{for } i \in \{1, \dots, n\}$$

- ▶ Close points should belong to the same cluster.

$$\forall x, y : \text{Close}(x, y) \rightarrow (C_i(x) \leftrightarrow C_i(y)) \text{ for } i \in \{1, \dots, n\}$$

Exemplary results



Relations

Domain A set D of points in the $[0, 1]^2$ square;

Predicate A set 2-ary relational predicates R_1, \dots, R_n

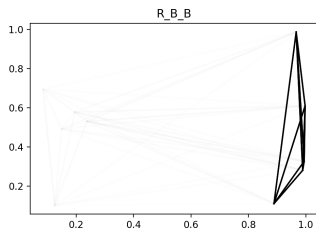
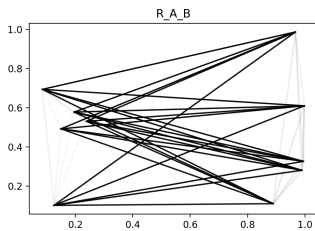
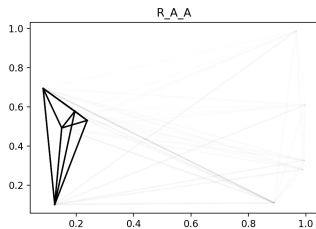
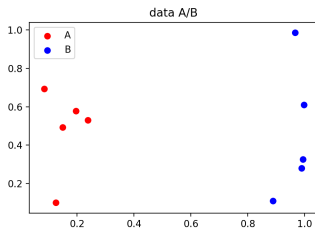
Supervision

1. **Positive examples for $R_i(x, y)$** a set of points in $D \times D$, which are known to be instances of R_i ;
2. **Negative examples for $R_i(x, y)$** a set of points in $D \times D$, which are known not to be instances of R_i ;

Constraints

- ▶ **Symmetry** if $\forall x, y \in D \times D : R(x, y) \rightarrow R(y, x)$
- ▶ **Subset** if $\forall (x, y) \in D \times D : R(x, y) \rightarrow R'(x, y)$
- ▶ ...

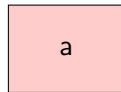
Relations



Learning spatial relations

Domain

Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

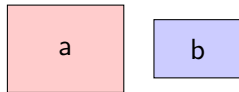


Learning spatial relations

Domain

Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

- ▶ *a* is on the left of *b*

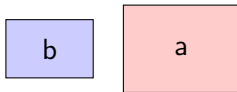


Learning spatial relations

Domain

Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

- ▶ a is on the left of b
- ▶ a is on the right of b

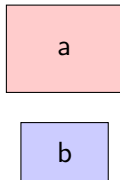


Learning spatial relations

Domain

Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

- ▶ a is on the left of b
- ▶ a is on the right of b
- ▶ a is above of b

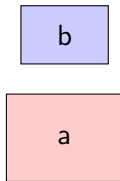


Learning spatial relations

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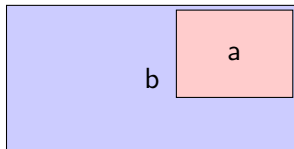


Learning spatial relations

Domain

Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

- ▶ a is on the left of b
- ▶ a is on the right of b
- ▶ a is above of b
- ▶ a is below of b
- ▶ a is contained in b

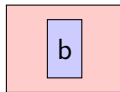


Learning spatial relations

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Our domain is composed of 2d rectangles on a plane; We are interested in the following 6 spatial relations.

- ▶ a is on the left of b
- ▶ a is on the right of b
- ▶ a is above of b
- ▶ a is below of b
- ▶ a is contained in b
- ▶ a contains b



Learning spatial relations

Problem

Given some examples of pairs of rectangles for each specific relation, and some background knowledge about them as for instance:

- ▶ left is the inverse of right
- ▶ an object cannot be at the same time on the left and on the right of another object
- ▶ if an object a is contained in an object b and b is on the left of c , then a is on the left of c

we want to be able to predict if two randomly generated rectangles, are in one of the 6 spatial relation.

Learning spatial relations

Domain representation

Every rectangle is represented with 4 real numbers,

$$\langle x, y, w, h \rangle$$

encoding the coordinates of the bottom-left corner, and the width and the height

The language

The language is constituted by 6 binary relations

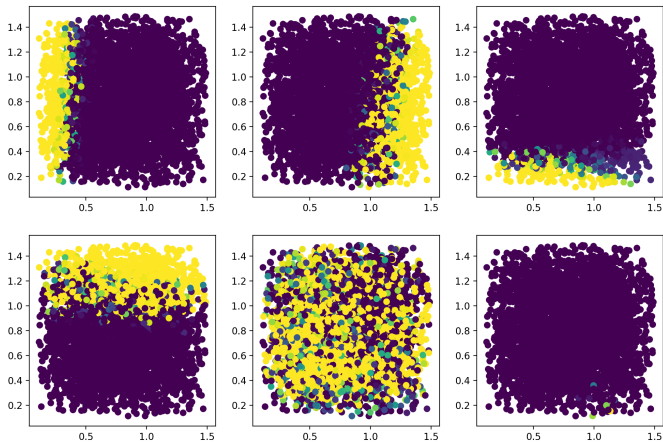
*left(x, y), right(x, y), above(x, y),
below(x, y), contains(x, y), in(x, y)*

Learning spatial relations

The constraints

- ▶ a set of positive examples for each spatial relation:
left(a, b), *right(c, d)*, *above(e, f)*, *below(h, i)*, *contains(j, k)*,
in(l, m), ...
- ▶ axioms about spatial relations:
 - ▶ $\forall x, y : \text{left}(x, y) \rightarrow \neg \text{left}(y, x)$;
 - ▶ $\forall x, y : \text{left}(x, y) \rightarrow \text{right}(y, x)$;
 - ▶ $\forall x, y, z : \text{in}(x, y) \wedge \text{left}(y, z) \rightarrow \text{left}(x, z)$
 - ▶ ...

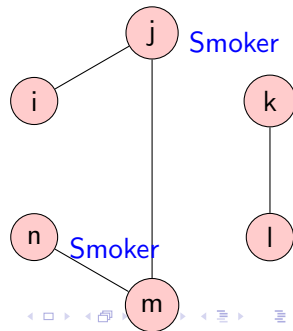
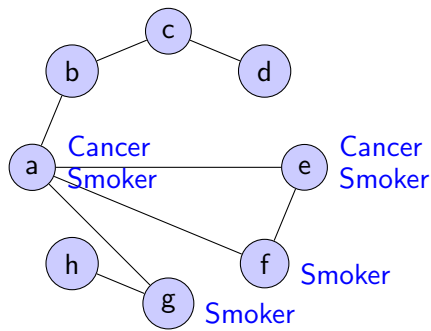
Learning spatial relations



Statistical relational learning

Domain: Smoking-Friends-Cancer, [?]

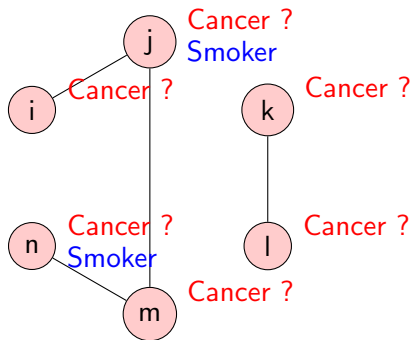
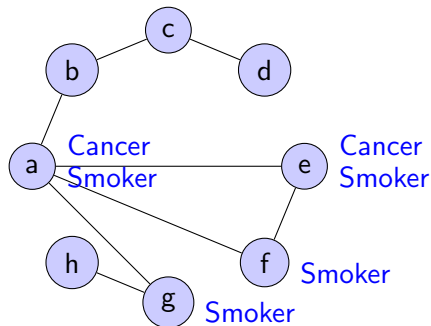
- ▶ Of two groups of people $\{a, b, \dots, h\}$ and $\{i, j, \dots, n\}$; we know if each of them smokes and the friendship relation within each group;
- ▶ for the first group we also know who has a cancer;
- ▶ we know that cancer depends on smoking
- ▶ and that smoking habits depend on the friendship relation



Statistical relational learning

Task 1

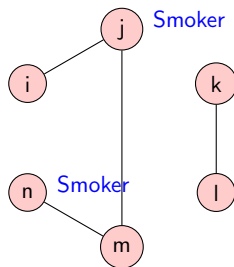
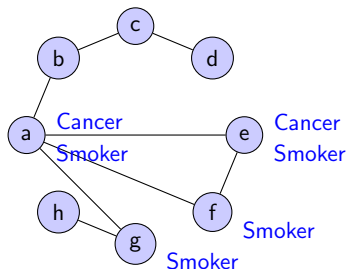
For each of the person of the second group we have to predict if he/she has a cancer or not



Statistical relational learning

Task 2

For each person we want to find a semantic embedding in \mathbb{R}^k consistent with the semantics and the structure. For instance:

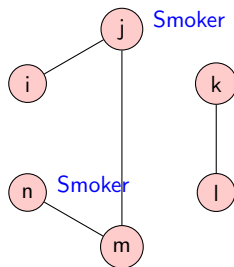
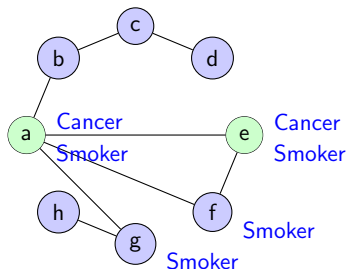


Statistical relational learning

Task 2

For each person we want to find a semantic embedding in \mathbb{R}^k consistent with the semantics and the structure. For instance:

- ▶ $a^{\mathcal{G}} \approx e^{\mathcal{G}}$ since both a and e smoke and have cancer, and they have two friends that smoke, one of which has a cancer

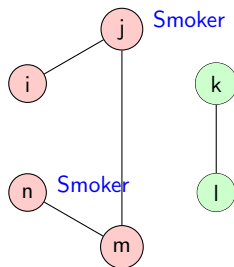
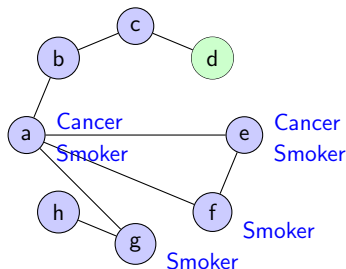


Statistical relational learning

Task 2

For each person we want to find a semantic embedding in \mathbb{R}^k consistent with the semantics and the structure. For instance:

- ▶ $a^{\mathcal{G}} \approx e^{\mathcal{G}}$ since both a and e smoke and have cancer, and they have two friends that smoke, one of which has a cancer
- ▶ $d^{\mathcal{G}} \approx k^{\mathcal{G}} \approx i^{\mathcal{G}}$ because they don't smoke and don't have cancer, and they have only one friend, who does not smoke and does not have a cancer

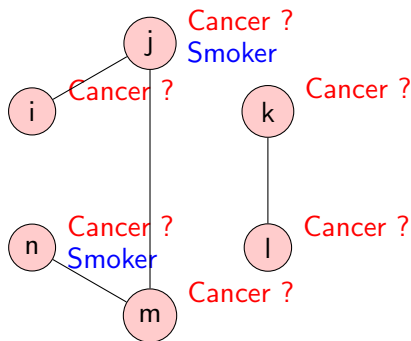
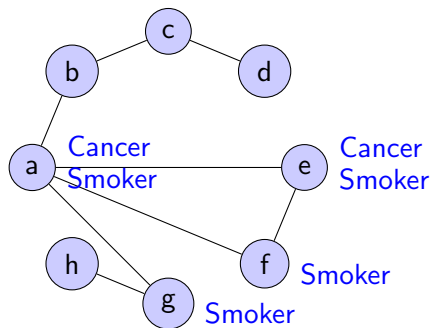


Statistical relational learning

Task 3

We want to know the truth value of certain formulas,

- ▶ e.g., the correlation between friendship and smoking habits;
- ▶ the correlation between smoking habits and cancer



Statistical relational learning

Representation of the domain in \mathbb{R}^k

For each individual of the domain $\{a, \dots, n\}$ we don't provide an explicit mapping to \mathbb{R}^k , which instead is generated, as the result of the constraint optimization. We only provide the dimension of the domain (i.e., k)

The language

- ▶ unary predicates $S(x)$ and $C(x)$ for “x smokes” and “x has a cancer” and a binary predicate $F(x, y)$ for “y is a friend of x”

Constraints

- ▶ $S(a), \neg S(b), \neg S(c), \neg S(d), S(e), \dots, \neg S(i), \dots, S(n)$;
- ▶ $C(a), \neg C(b), \neg C(c), \dots, \neg C(h)$;
- ▶ $\forall x : \neg F(x, x)$
- ▶ $\forall xy : F(x, y) \rightarrow F(y, x)$

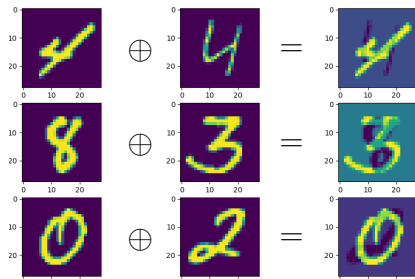
Mnist with constraints

Dataset

Contains pictures resulting from overlaying two MNIST digit pictures, where the smaller digit is in black-on-white and the smallest in white-on-black. I.e., d_x and d_y are the pixel matrices of the digits x and y , then the pixel matrix $d_{xy} = d_x \oplus d_y$ is defined as

$$d_x \oplus d_y = \begin{cases} d_x - w \cdot d_y & \text{if } x \leq y \\ d_y - w \cdot d_x & \text{Otherwise} \end{cases}$$

where w is randomly generated number in $[0, 1]$

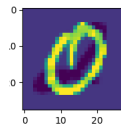
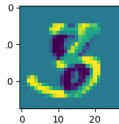
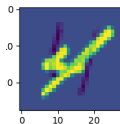


Mnist with constraints

The Task

Given an image $d_{xy} = d_x \oplus d_y$ we have to predict x and y .

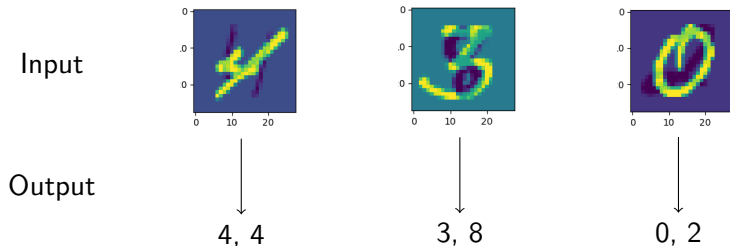
Input



Mnist with constraints

The Task

Given an image $d_{xy} = d_x \oplus d_y$ we have to predict x and y .



Mnist with constraints

The language

20 unary predicates, two for every digit;

$$\begin{array}{llll} zero_1(x) & one_1(x) & \dots & nine_1(x) \\ zero_2(x) & one_2(x) & \dots & nine_2(x) \end{array}$$

$zero_1(x)$ (resp. $zero_2(x)$) means: “the smaller (resp the larger) digit of x is a 0

Constraints

- ▶ $four_1 \left(\begin{array}{c} \text{[Handwritten 4]} \\ \text{0 10 20} \end{array} \right), three_1 \left(\begin{array}{c} \text{[Handwritten 3]} \\ \text{0 10 20} \end{array} \right), zero_1 \left(\begin{array}{c} \text{[Handwritten 0]} \\ \text{0 10 20} \end{array} \right),$
 $four_2 \left(\begin{array}{c} \text{[Handwritten 4]} \\ \text{0 10 20} \end{array} \right), eight_2 \left(\begin{array}{c} \text{[Handwritten 8]} \\ \text{0 10 20} \end{array} \right), two_2 \left(\begin{array}{c} \text{[Handwritten 2]} \\ \text{0 10 20} \end{array} \right), \dots$
- ▶ $\forall x : zero_1(x) \rightarrow \neg one_1(x), \dots$
- ▶ $\forall x : \neg(one_1(x) \wedge zero_2(x)), \dots$