

TAUDoS Theory and Algorithms for the Understanding of Deep learning On Sequential data

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TAUDoS

Context

- Statistical ML, in particular deep learning, allows great practical results
- However: decision process not accessible to human beings (even Machine Learners!)
- A better **understanding** is needed for business development or even legally required (GDPR).

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Overview of TAUDoS

- Focus of this Project : Understanding of neural networks on (discrete) sequential data
- 4 different research paths:

- Theoretical characterizations
- Knowledge distillation of grey/white boxe
- Learning strategies for interpretability or distillation
- Definition and learning of metrics for RNN

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Theoretical Insights

- Recent theorems [Rabusseau et al., 19, Li et al. 20]: linear second order RNN, tensor networks, and Weighted Automata (WA) are equivalent
- Consequences: Proven learning algorithm for WA extended to RNN
- In TAUDoS: extend to other classes (ex: bi-directional RNN and Weighted Context-Free grammars)
- **Possible practical use:** initialization of non-linear RNN

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Knowledge Distillation

- **Goal:** Extract simpler, more explicit models from already learn deep networks
- Recent work: [Eyraud & Ayache, 20; others] extract Weighted Automata from LSTM/GRU with surprising accuracy
- In TAUDoS:
 - direct continuations & improvements of the recent algorithm
 - Subpart detection for subpart distillation

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Learning Strategies

- **Goal:** Design new types of layers or of constraints dedicated to understanding of RNN
- Recent work: topical subject in the field
- In TAUDoS:

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- Discretizing parallel layer to help distillation
- Compositional constraints (a disentangling approach)

- Attention with interpretability-based constraints

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Metric Learning

- Goal: Design and learn metric to compare RNN behavior
- Recent work: few...

- In TAUDoS:
 - Use the link between WA and 2-RNN (as WA come with a computable distance)
 - Metric learning from Euclidean projection

Valorization

- Open-source toolbox with all the developed approaches
- 2 use cases provided by the firm:

Medical (prediction of post-operation complications)

Law (Understanding of legal documents)