

A Multi-Grained Self-Interpretable Symbolic-Neural Model For Single/Multi-Labeled Text Classification

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00/ Motivation & What we achieved

- Human rationales

Hierarchical, Multi-grained for language understanding

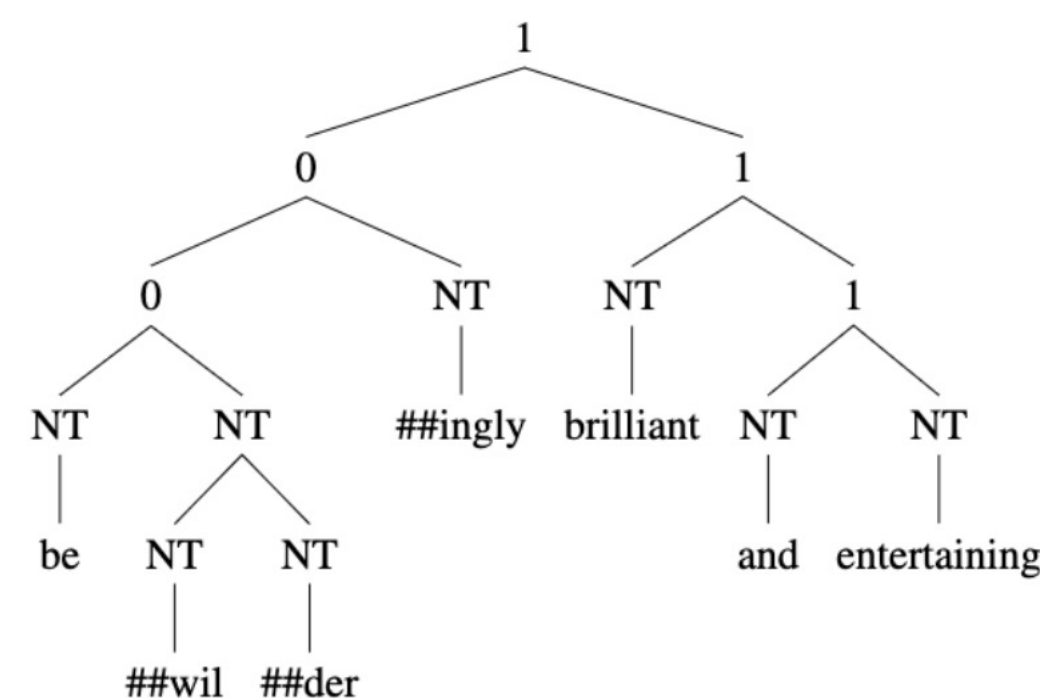
e.g. For multi-intent understanding, we are able to recognize the relationships between intents and text spans.

For semantic analysis, we are able to identify the sentiment and polarity reversal at the levels of words, phrases and clauses.

- Model rationales of ours

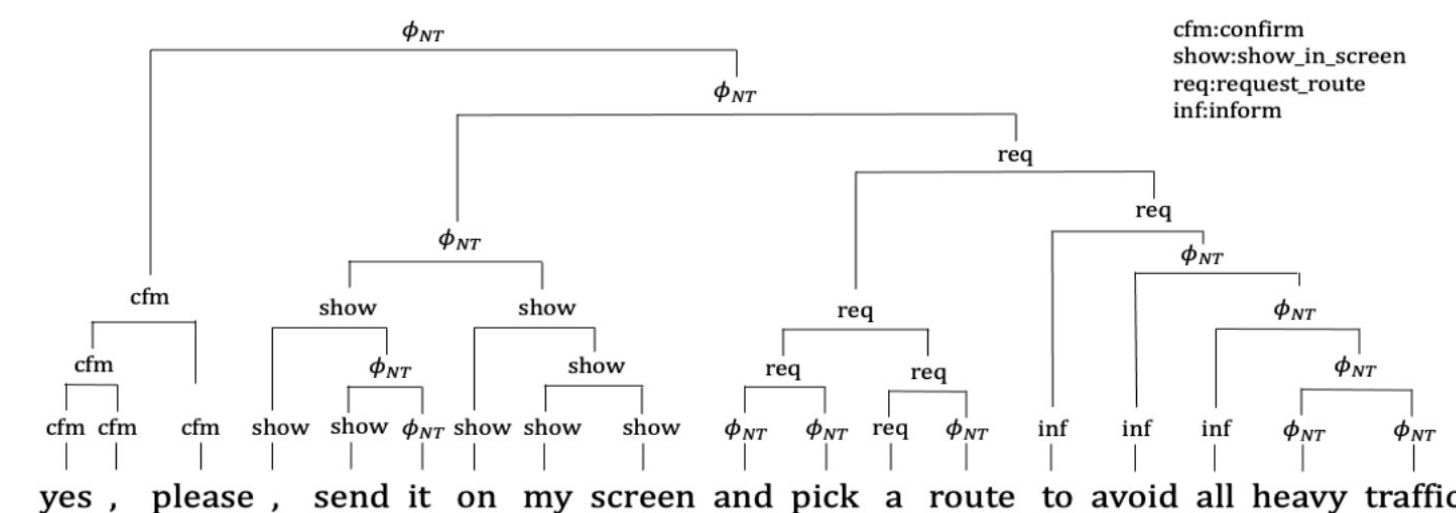
NER given labels only

fromloc.city_name toloc.city_name depart_time.period_of_day
Find a flight between Denver and Oakland the flight **should be in the afternoon**
arrive_time.time flight_stop
and arrive **close to 5 pm** **the flight should be nonstop**



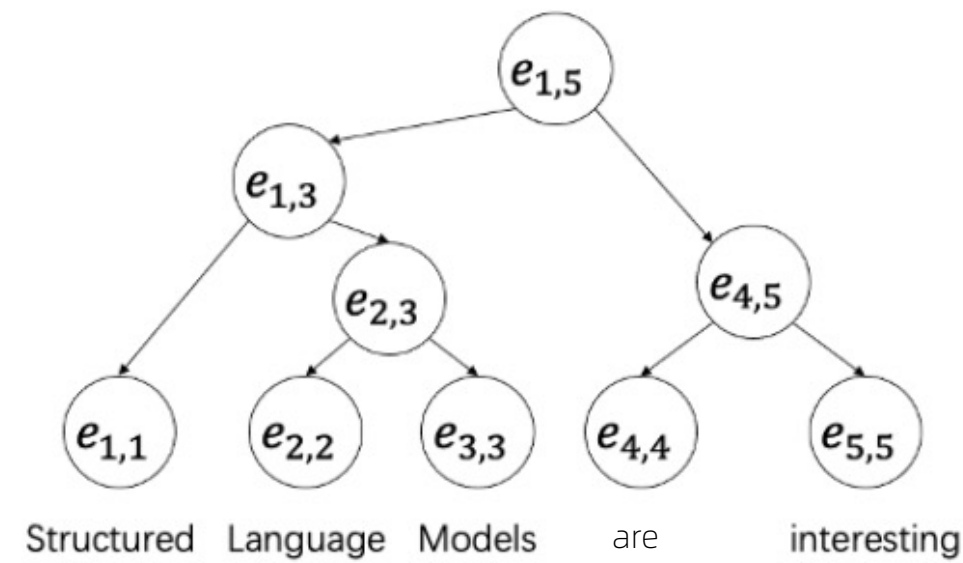
Bewilderingly brilliant and entertaining

Classification tasks with sentence labels only

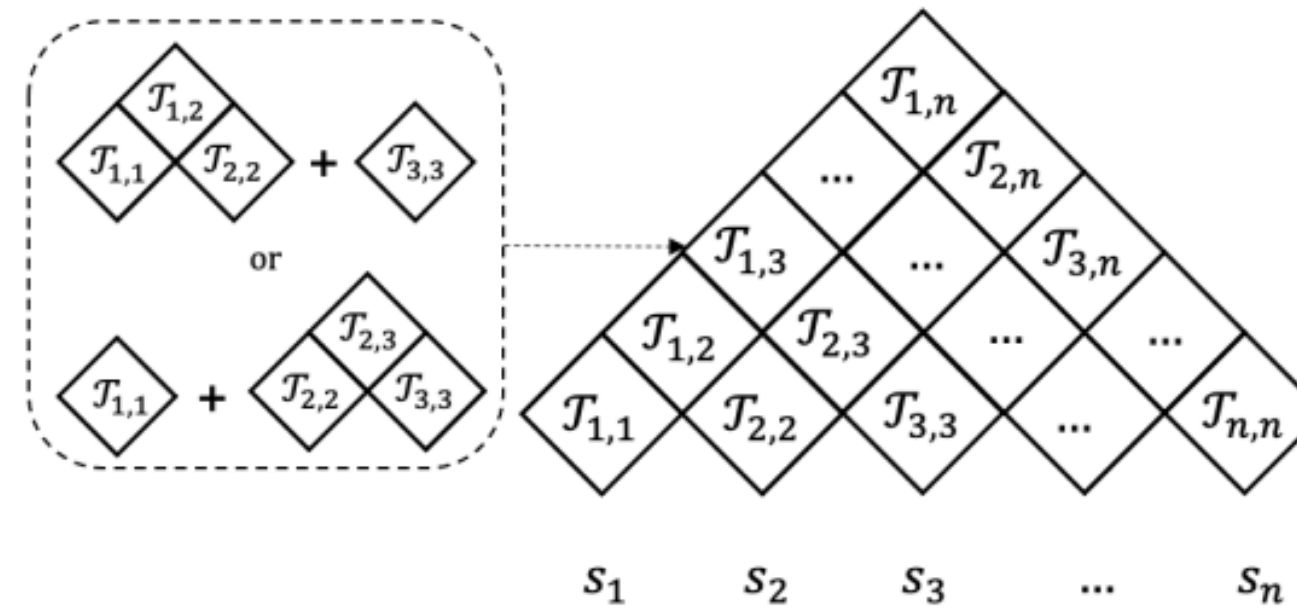


01 / Preliminary & Assumption

- Structured Language models

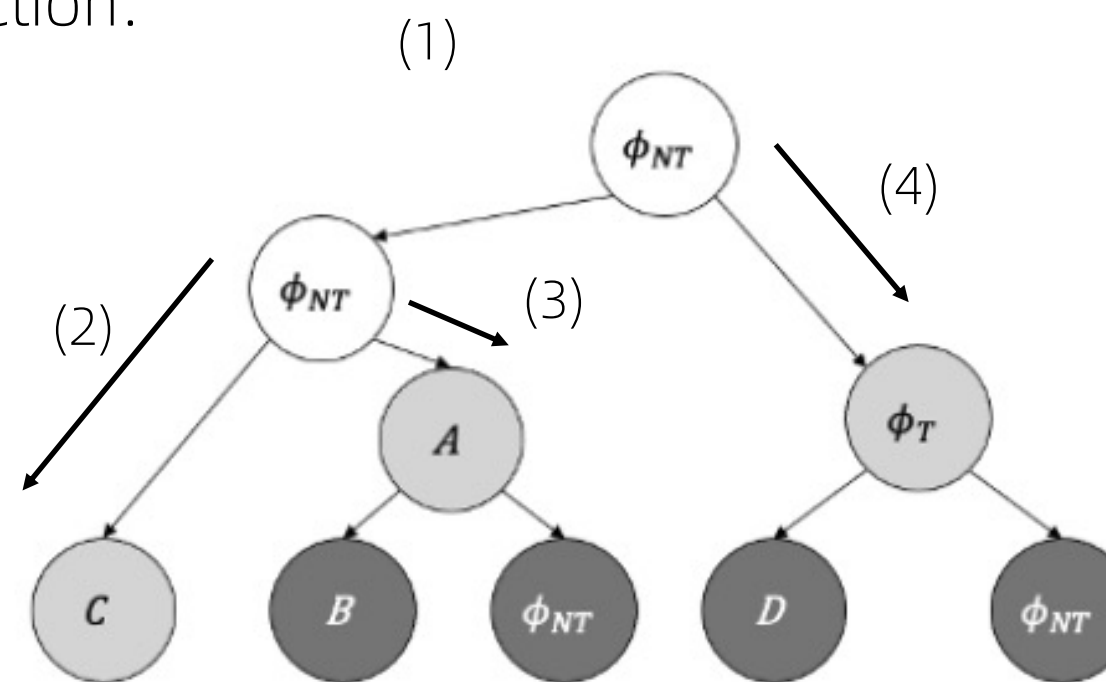
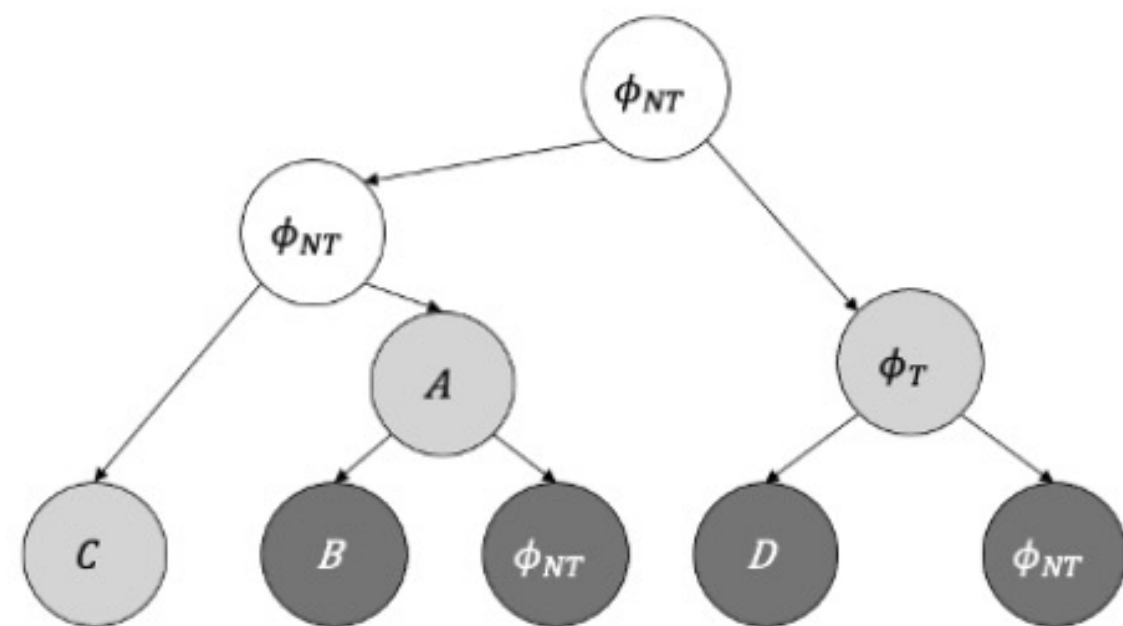


Fast-R2D2



- Our Assumption

Yield function:

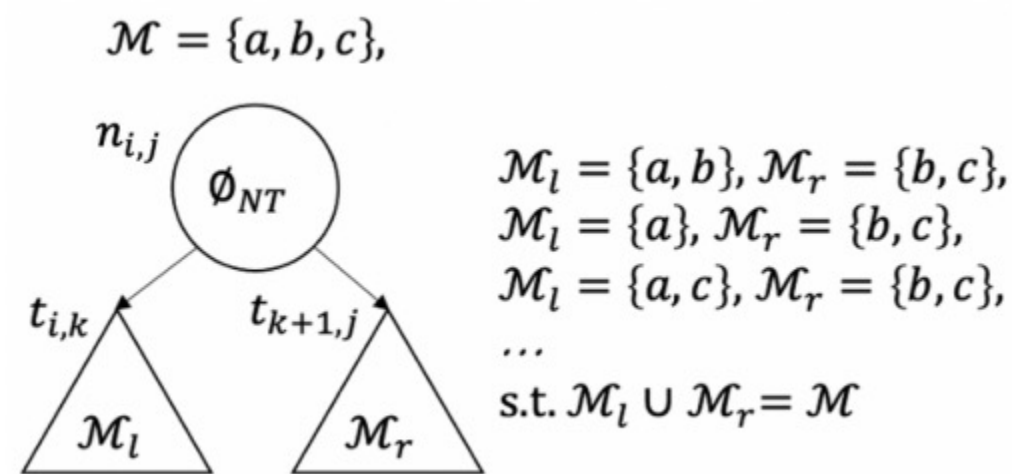
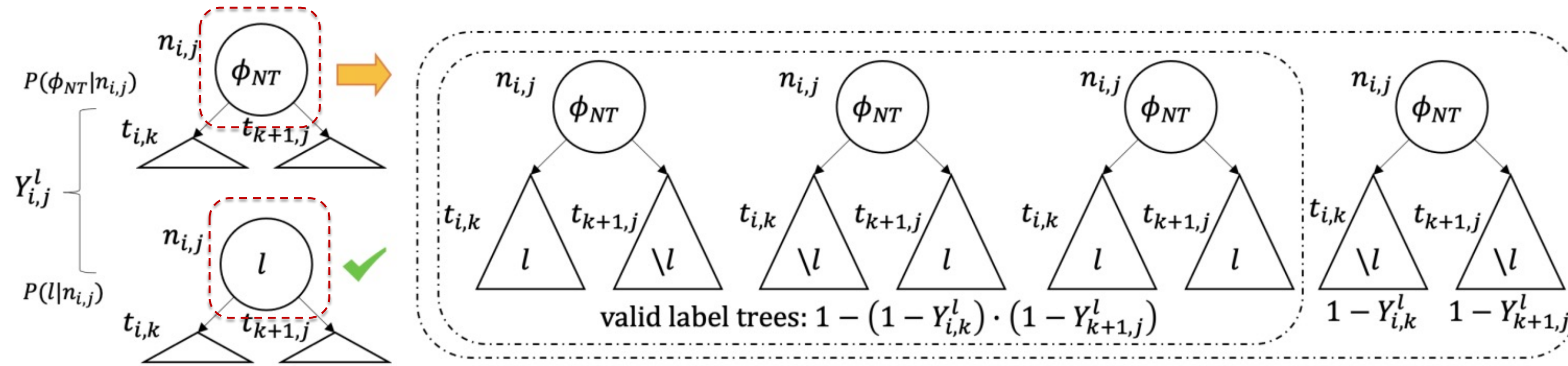


ϕ_{NT} continue traversing (NT stands for non-terminal)
 ϕ_T stop traversing (T stands for terminal)

Final labels: {A, C}

02/ Approaches

What's the probability that yield result of a given tree contains a given label l ?



$$\mathcal{L}_{cls}^t(\Psi) = -\log P_{\Psi}(\hat{t}^{[y(t)=\mathcal{T}]}) = -\sum_{l \in \mathcal{T}} \log Y_{1,|\mathbf{s}|}^l - \log(1 - Y_{1,|\mathbf{s}|}^{\mathcal{O}})$$

\mathcal{T} : target labels, \mathcal{O} : other labels

Thanks