

Using DeepProbLog to perform Complex Event Processing on an Audio Stream

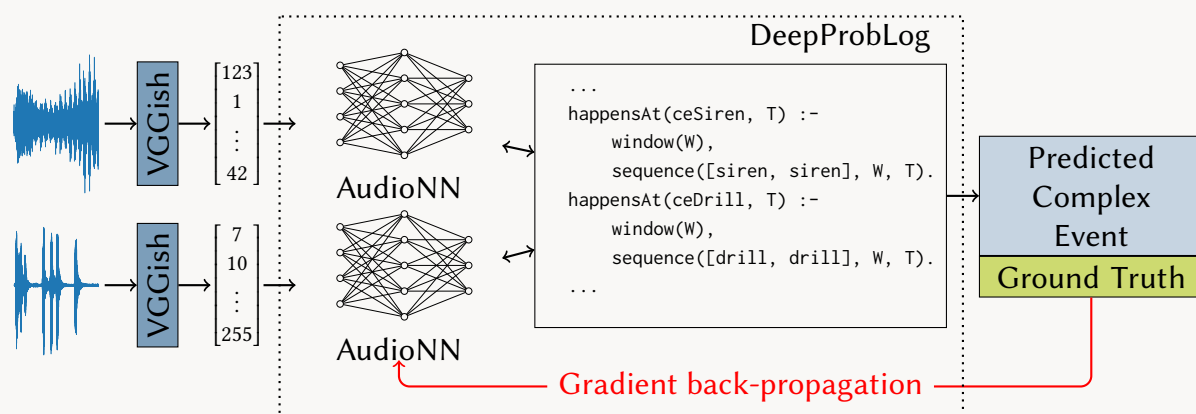
Motivating scenario

- ▶ Audio sensors have been deployed in a city.
- ▶ We want to detect *complex events*:
 - These are combinations of events distributed in time and space.
 - These can be detected by combinations of sounds defined by an expert.
 - For instance, if there is *shouting*, *glass shattering* and *sirens*, we can detect a *riot*.
- ▶ However, we do not have large amounts of training data, as these situations are rare.

Objectives

- ▶ Being able to use subsymbolic data as input.
- ▶ Retaining flexibility in rule definitions.
- ▶ Being able to perform end-to-end training.
- ▶ Being robust against noisily labelled data.

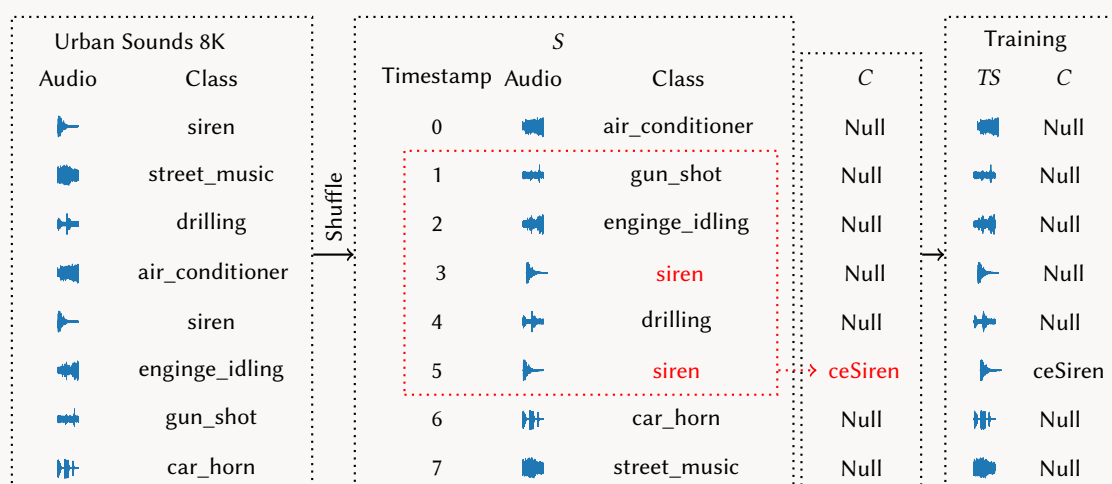
Proposed approach



The ProbLog code in the figure above shows the example rules used in the experiments below. These use the clause *sequence* to define event patterns. Users are able to define more complex rules with the full expressiveness of ProbLog.

Base dataset

We generated a synthetic dataset based on Urban Sounds 8K. The resulting dataset contains 10 types of complex events, which happen when two instances of the same class of sound occur within a window of time. In our experiments, we used window sizes between 2 and 5 seconds.



Random noise datasets

- ▶ To evaluate robustness to noisily labelled data, we generated *random noise datasets*.
- ▶ These have different percentages of noisy data (0% to 60% of the dataset).
- ▶ For this noisy data, the correct label has been replaced by a random one.

Related work and its limitations

There are three main types of approaches to this type of problem:

- ▶ Combining pre-trained neural networks with symbolic approaches. [1,2]
 - Requires access to pre-trained neural networks.
- ▶ Neural network approaches such as LSTMs [3] or Convolutional 3Ds [4].
 - They require very large amounts of data to train.
 - Does not allow for rule definitions by experts.
- ▶ Neuro-symbolic approaches. [5]
 - Currently, they significantly limit rule definitions.

- [1] Roldán, J.; Boubeta-Puig, J.; Luis Martínez, J.; and Ortiz, G. 2020. Integrating complex event processing and machine learning: An intelligent architecture for detecting iot security attacks. *Expert Systems with Applications* 149:113251.
- [2] Roig Vilamala, M.; Hiley, L.; Hicks, Y.; Preece, A.; and Cerutti, F. 2019. A pilot study on detecting violence in videos fusing proxy models. In *2019 22th International Conference on Information Fusion (FUSION)*, 1–8.
- [3] Mishra, S.; Jain, M.; Siva Naga Sasank, B.; and Hota, C. 2018. An ingestion based analytics framework for complex event processing engine in internet of things. In *Mondal, A.; Gupta, H.; Srivastava, J.; Reddy, P. K.; and Somayajulu, D., eds., Big Data Analytics*, 266–281. Cham: Springer International Publishing.
- [4] Liu, K.; Liu, W.; Gan, C.; Tan, M.; and Ma, H. 2018. T-c3d: Temporal convolutional 3d network for real-time action recognition.
- [5] Xing, T.; Garcia, L.; Vilamala, M. R.; Cerutti, F.; Kaplan, L.; Preece, A.; and Srivastava, M. 2020. Neuroplex: Learning to Detect Complex Events in Sensor Networks through Knowledge Injection. New York, NY, USA: Association for Computing Machinery. 489–502.

Results

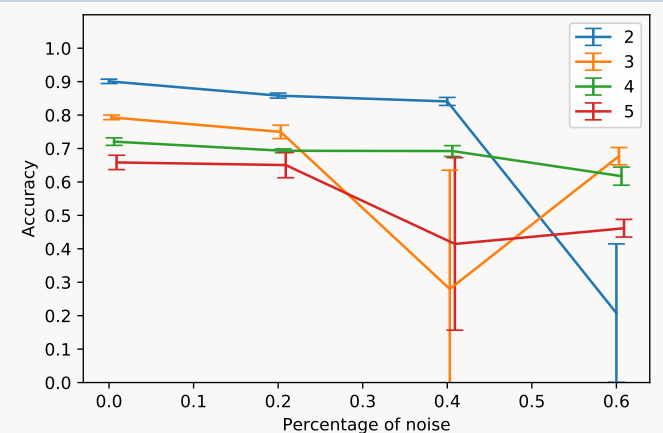


Figure: Evaluation of system's performance with the different percentages of noise from the random noise datasets. Each colour represents a different window size. Error bars show the standard deviation over 3 executions.

- ▶ As shown in the figure above, our approach performs fairly well with the base dataset (percentage of noise of 0.0).
- ▶ Bigger window sizes perform worse than the smaller ones. This is due to the added complexity of bigger window sizes.
- ▶ A percentage of noise of 20% has an almost imperceptible effect on the accuracy of the system.
- ▶ Bigger percentages of noise lead to unreliable results, as the system is not able to train correctly in a consistent manner.

Conclusions and future work

- ▶ We have designed a system that fulfills all four objectives.
- ▶ As future work, we want to consider other types of subsymbolic data.
- ▶ Improvements to the time efficiency of our approach would also be a good area of research, as that is a downside of our approach.