# Sound event detection using weakly labeled dataset with convolutional and recurrent neural network

### Sharath Adavanne, Tuomas Virtanen

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- Introduction
  - Sound tagging / Weak labels
  - Sound event detection / Strong labels
- Dataset
- Proposed neural network
- Results



Time -->







Time -->





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Sound event detection using sound tagging dataset with convolutional and recurrent neural network

## Dataset

- Subset of Google's AudioSet
- 51,172 for training, 488 for testing
- 10s clips (padded with zeros, if not 10s)



# Dataset

- Subset of Google's AudioSet
- 51,172 for training, 488 for testing
  - 10s clips (padded with zeros, if not 10s)

- Weak labeled
- 17 classes Car, Bus, Train, Truck etc..
- Single recording can have more than one sound source









Weak label Sound Tagging

### **Training procedure**



#### Input

Convolutional and recurrent neural network (CRNN) architecture

**Convolutional layers** 

**Recurrent layers** 

Fully connected layers

Output

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Convolutional and recurrent neural network (CRNN) architecture

**Convolutional layers** 

**Recurrent layers** 

Why?

Fully connected layers

Output

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Fully connected layers



Sound tagging

Output

Sound event detection



















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Input

Log mel-band energies of complete 10 second audio Saliency map for -jc0NAxK8M\_30.000\_40.000 recording

Gradients for strong label

Gradients for weak label

Log mel-band energy



		Sound tagging				Sound Event detection		
Strong Weight	Weak Weight	Precision	Recall	F-score		Error rate	F-score	
0.002	1	44.9	37.0	40.5		1.38	10.9	

Convolutional layers

**Recurrent layers** 

Fully connected layers

strong weight \* strong label loss for back propagation

Loss 1 : strong label loss

Fully connected layers

Loss 2 : weak label loss

weak weight \* weak label loss for back propagation

		Sound tagging			Sound Event detection			
Strong Weight	Weak Weight	Precision	Recall	F-score		Error rate	F-score	
0.002	1	44.9	37.0	40.5		1.38	10.9	
0.02	1	44.2	36.5	40.0		1.13	17.0	
0.2	1	47.5	39.6	43.2		0.84	38.1	
1	1	47.5	39.7	43.3		0.84	38.8	
1	0.2	47.3	39.5	43.0		0.84	38.6	
1	0.02	25.5	20.6	22.8		0.81	41.1	
1	0.002	20.5	16.5	18.3		0.81	42.4	

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1	1	47.5	39.7	43.3	45.5	0.84	38.8	0.81
1	0.2	47.3	39.5	43.0	44.5	0.84	38.6	0.82
1	0.02	25.5	20.6	22.8		0.81	41.1	
1	0.002	20.5	16.5	18.3	26.3	0.81	42.4	0.79

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					55.6			0.66

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- CRNN was shown to learn temporal information, given just weak labels
- Best result was against our intuition: equal scaling of strong and weak loss

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### • Future work

- More fine tuning
- Strong labels for high energy regions only
- Attention layers

# Thank you.

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#### 218,000 trainable weights