
Learnable pooling with context Gating for video classification

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Video data classification



Video data classification(WILLOW)



- Feature extraction : Given
- Feature aggregation : Use existing methods
 - LSTM, GRU, DBoF, VLAD, Fisher Vector encoding
- Classification : MoE

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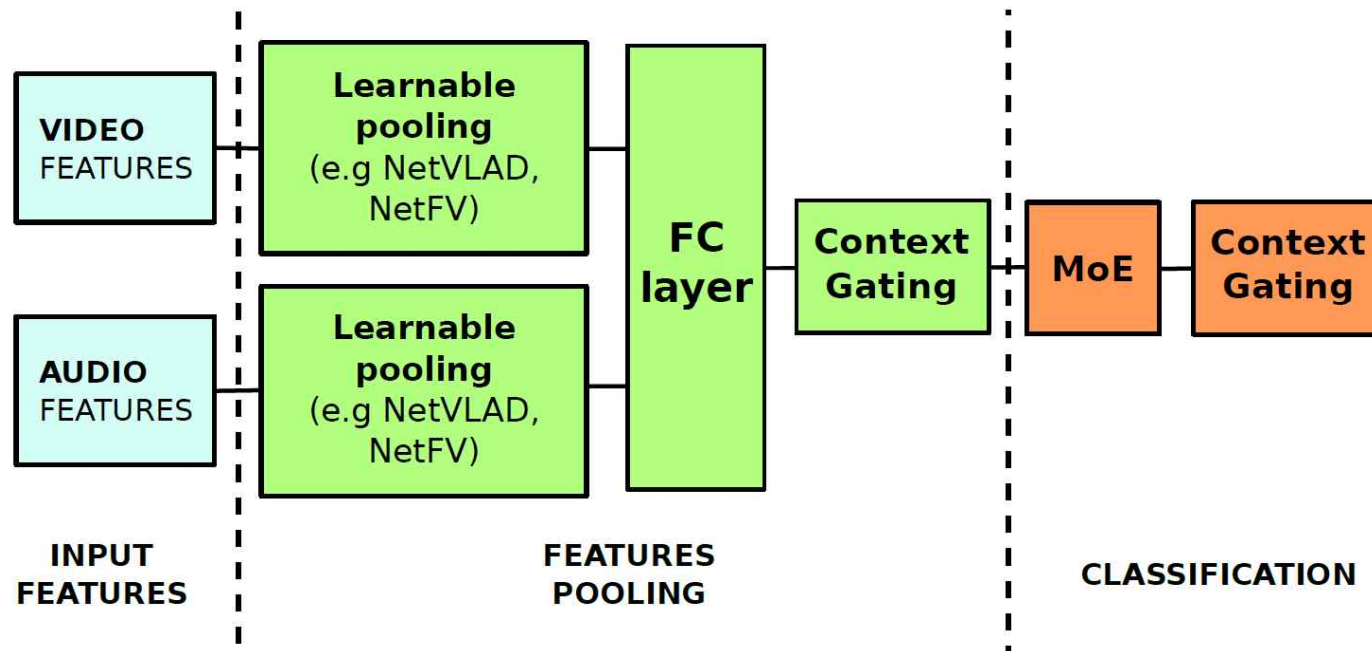


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Context
Gating

Overview



Context Gating

$$CG(X) = \sigma(W X + b) \cdot X$$

- Motivation

1. Wish to introduce non-linear interactions among activations of the input representation
2. Wish to recalibrate the strengths of different activations of the input representation through a self-gating mechanism

- Aim

1. Capture dependencies among features
2. Capture prior structure of output space

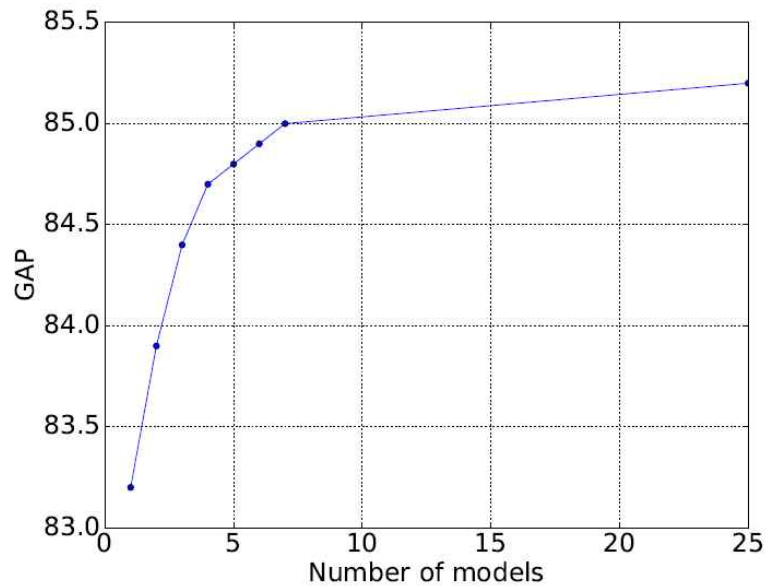
Experiments

Method	GAP
Baseline 1 (Average pooling + Logistic Regression)	71.4%
Baseline 2 (Average pooling + MoE + CG)	74.1%
LSTM (2 Layers)	81.7%
GRU (2 Layers)	82.0%
Soft-DBoW (4096 Clusters)	81.6%
NetFV (128 Clusters)	82.2%
NetVLAD (256 Clusters)	82.4%
Gated Soft-DBoW (4096 Clusters)	82.0%
Gated NetFV (128 Clusters)	83.0%
Gated NetRVLAD (256 Clusters)	83.1%
Gated NetVLAD (256 Clusters)	83.2%

Method	GAP
NetVLAD	82.2%
NetVLAD + CG after pooling	82.7%
NetVLAD + GLU after pooling, CG after MoE	82.7%
NetVLAD + CG after pooling and MoE	83.0%

Method	Early Concat	Late Concat
NetVLAD	81.9%	82.4%
NetFV	81.2%	82.2%
GRU	82.2%	82.1%
LSTM	81.7%	81.1%

Ensemble



- The ensemble did not bring much when combining best but similar models
- Simple greedy approach
- The first seven models
 - Gated NetVLAD
 - Gated NetFV
 - Gated Soft-DBoW
 - Soft DBoW
 - Gated NetRVLAD
 - GRU
 - LSTM