

Aspect Based Sentiment Analysis with Gated Convolutional Networks

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Outline

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 - Problem Statement
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Problem Statement - Aspect Based Sentiment Analysis

Two subtasks:

- Aspect-Category Sentiment Analysis (ACSA)
- Aspect-Term Sentiment Analysis (ATSA)

Sentence

Average to good Thai food, but terrible delivery.

Average to good Thai food, but terrible delivery.

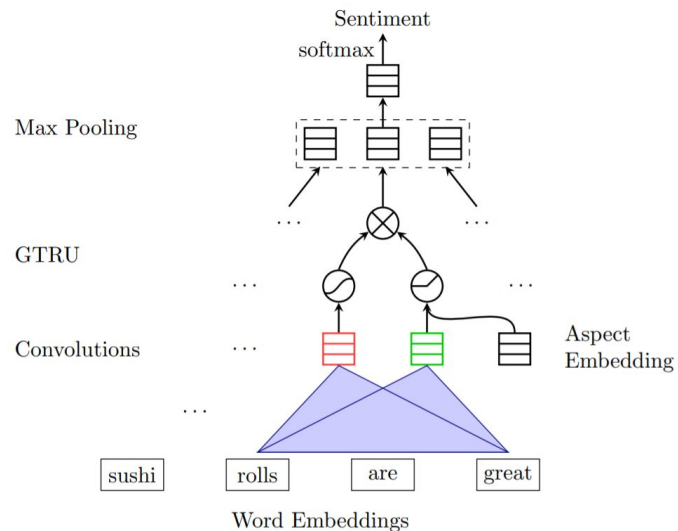
→ ATSA: Food, Delivery

→ ACSA: Service

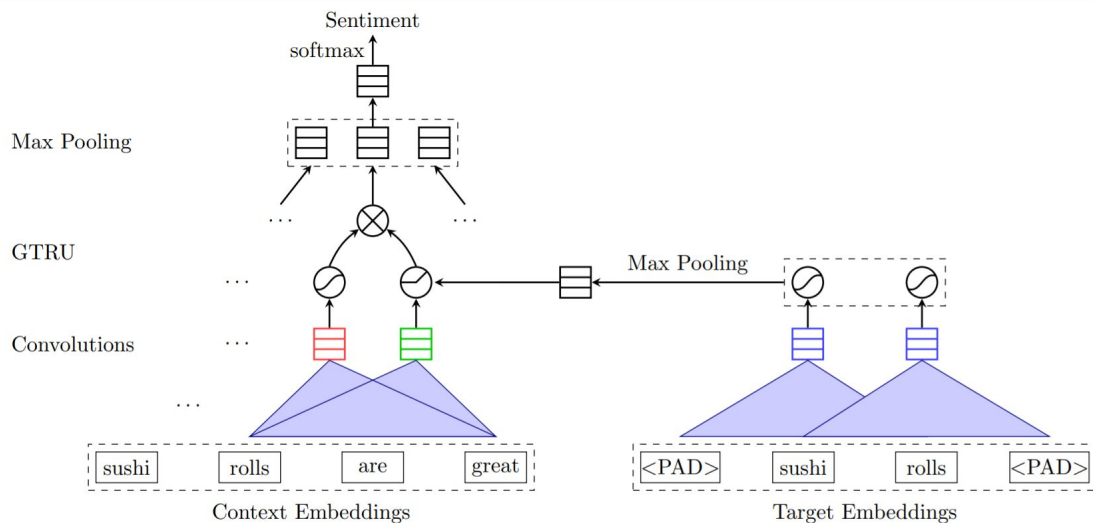
Proposed Approach

- CNN Model with Gating Mechanism
- Gated Tanh-ReLU selectively outputs sentiment features according to a given aspect or entity.
- Simpler than existing models (compared to models with attention)
- Can be trained in parallel - not time dependent unlike previous attention based and LSTM models

Gated Convolutional Network with Aspect Embedding (GCAE)



ACSA



ATSA

Datasets

- SemEval Workshops - customer reviews: Restaurant & Laptop
- **Hard Datasets:** Sentences having multiple aspect labels associated with multiple sentiments

Sentence	aspect category/term	sentiment label
Average to good Thai food, but terrible delivery.	food	positive
Average to good Thai food, but terrible delivery.	delivery	negative

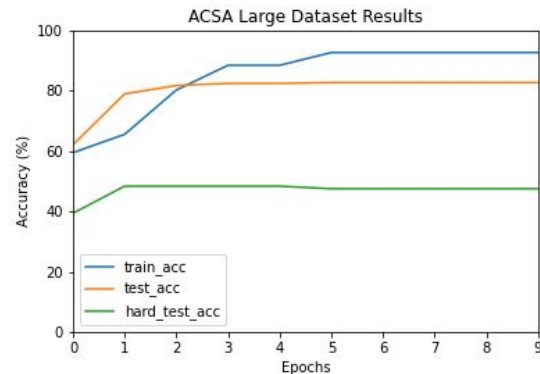
Table 1: Two example sentences in one hard test set of restaurant review dataset of SemEval 2014.

Implementation Details

- 300-dimension GloVe vectors pre-trained on unlabeled data of 840 billion tokens
- Random initialization - uniform distribution $U(-0.25, 0.25)$
- Adagrad
- Batch size: 32 instances
- Learning Rate: $1e-2$
- Maximal epochs: 30
- 5-fold cross validation

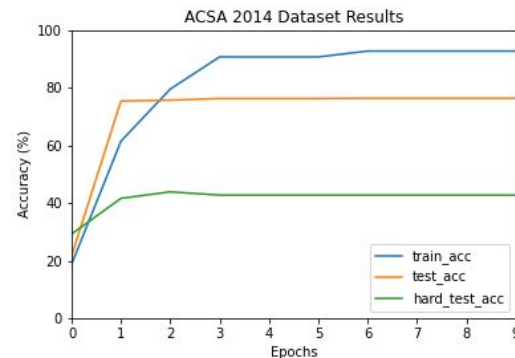
Results - ACSA

Models	Restaurant-Large		Restaurant 2014	
	Test	Hard Test	Test	Hard Test
SVM*	-	-	75.32	-
SVM + lexicons*	-	-	82.93	-
ATAE-LSTM	83.91 \pm 0.49	66.32 \pm 2.28	78.29 \pm 0.68	45.62 \pm 0.90
CNN	84.28 \pm 0.15	50.43 \pm 0.38	79.47 \pm 0.32	44.94 \pm 0.01
GCN	84.48 \pm 0.06	50.08 \pm 0.31	79.67\pm0.35	44.49 \pm 1.52
GCAE	85.92\pm0.27	70.75\pm1.19	79.35 \pm 0.34	50.55\pm1.83



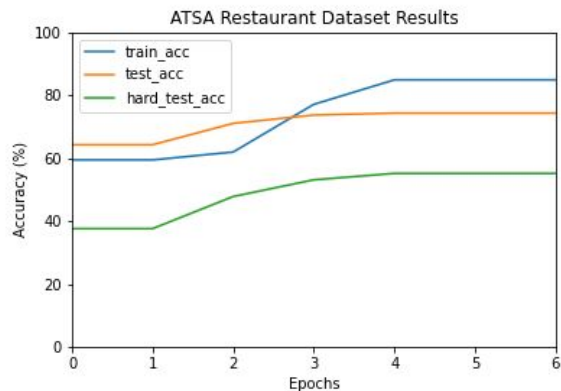
Reproduced Results

GCAE	Restaurant-Large		Restaurant 2014	
	Test	Hard-Test	Test	Hard-Test
	82.60	47.43	76.29	42.69



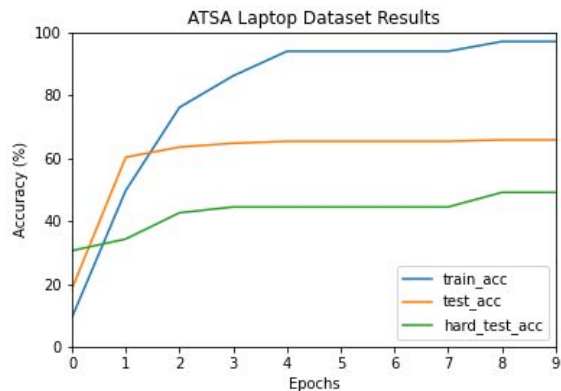
Results - ATSA

Models	Restaurant		Laptop	
	Test	Hard Test	Test	Hard Test
SVM*	77.13	-	63.61	-
SVM + lexicons*	80.16	-	70.49	-
TD-LSTM	73.44 \pm 1.17	56.48 \pm 2.46	62.23 \pm 0.92	46.11 \pm 1.89
ATAE-LSTM	73.74 \pm 3.01	50.98 \pm 2.27	64.38 \pm 4.52	40.39 \pm 1.30
IAN	76.34 \pm 0.27	55.16 \pm 1.97	68.49 \pm 0.57	44.51 \pm 0.48
RAM	76.97 \pm 0.64	55.85 \pm 1.60	68.48 \pm 0.85	45.37 \pm 2.03
GCAE	77.28\pm0.32	56.73\pm0.56	69.14\pm0.32	47.06\pm2.45



Reproduced Results

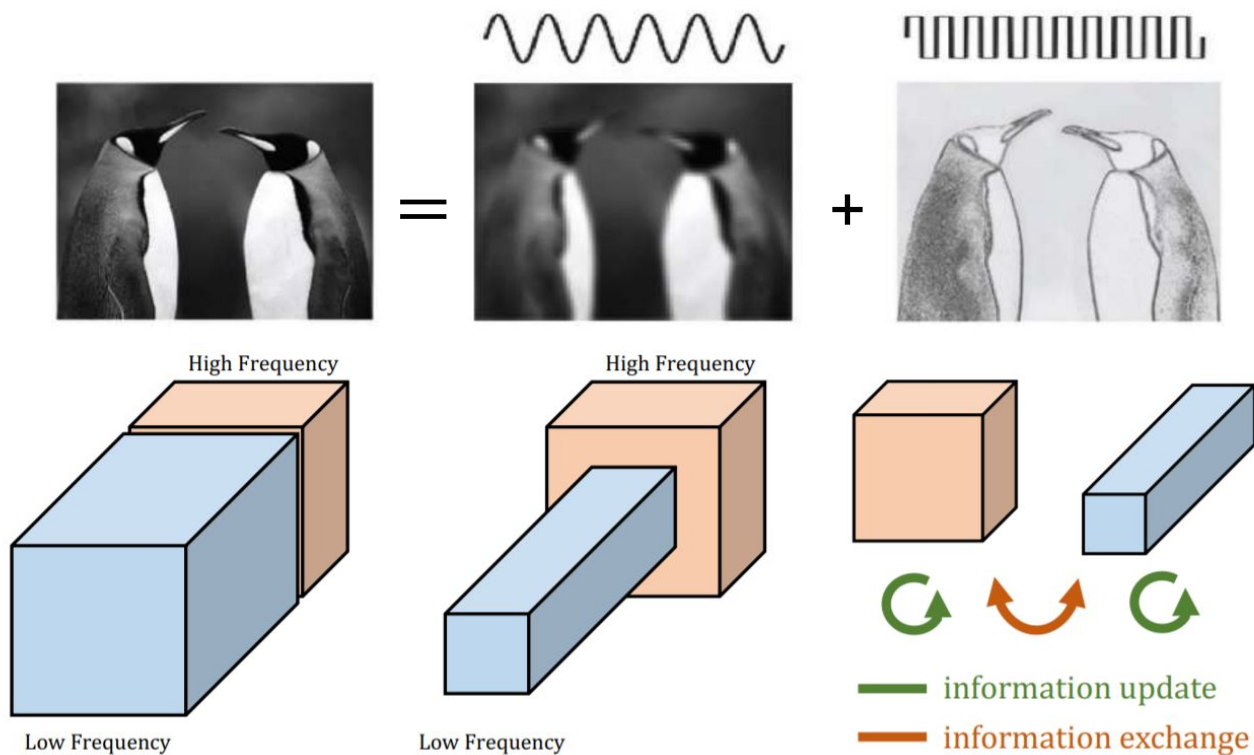
GCAE	Restaurant		Laptop	
	Test	Hard-Test	Test	Hard-Test
	74.25	55.10	65.74	49.07



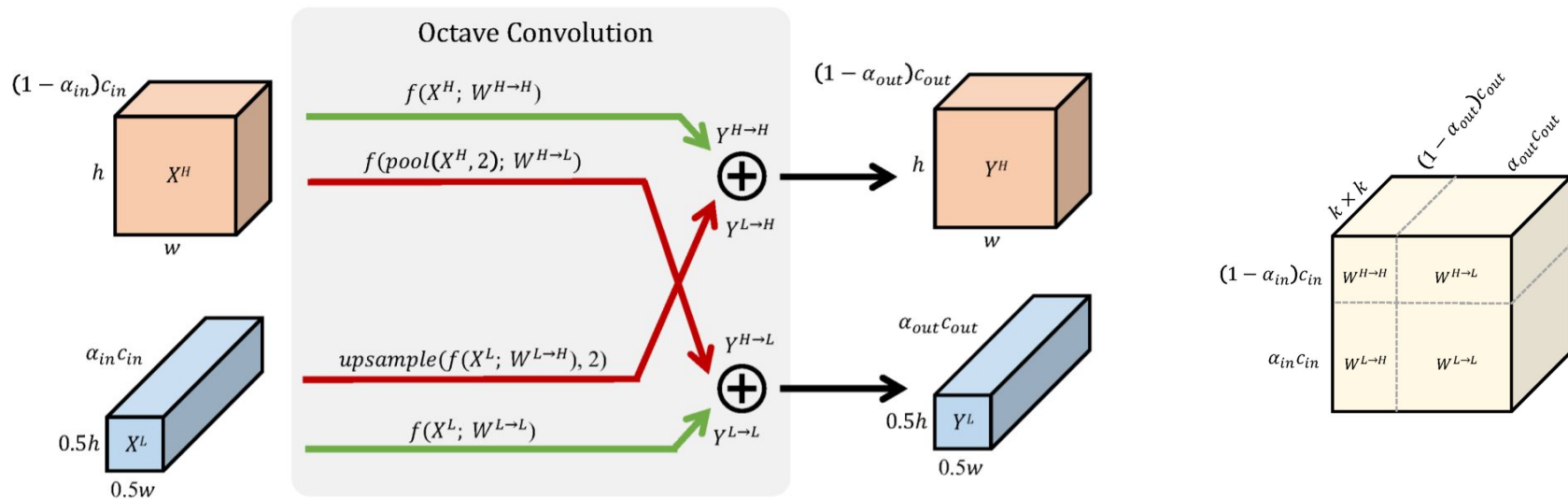
Novel Ideas that can be implemented

- Octave Convolutions
- Squeeze & Excitation Block

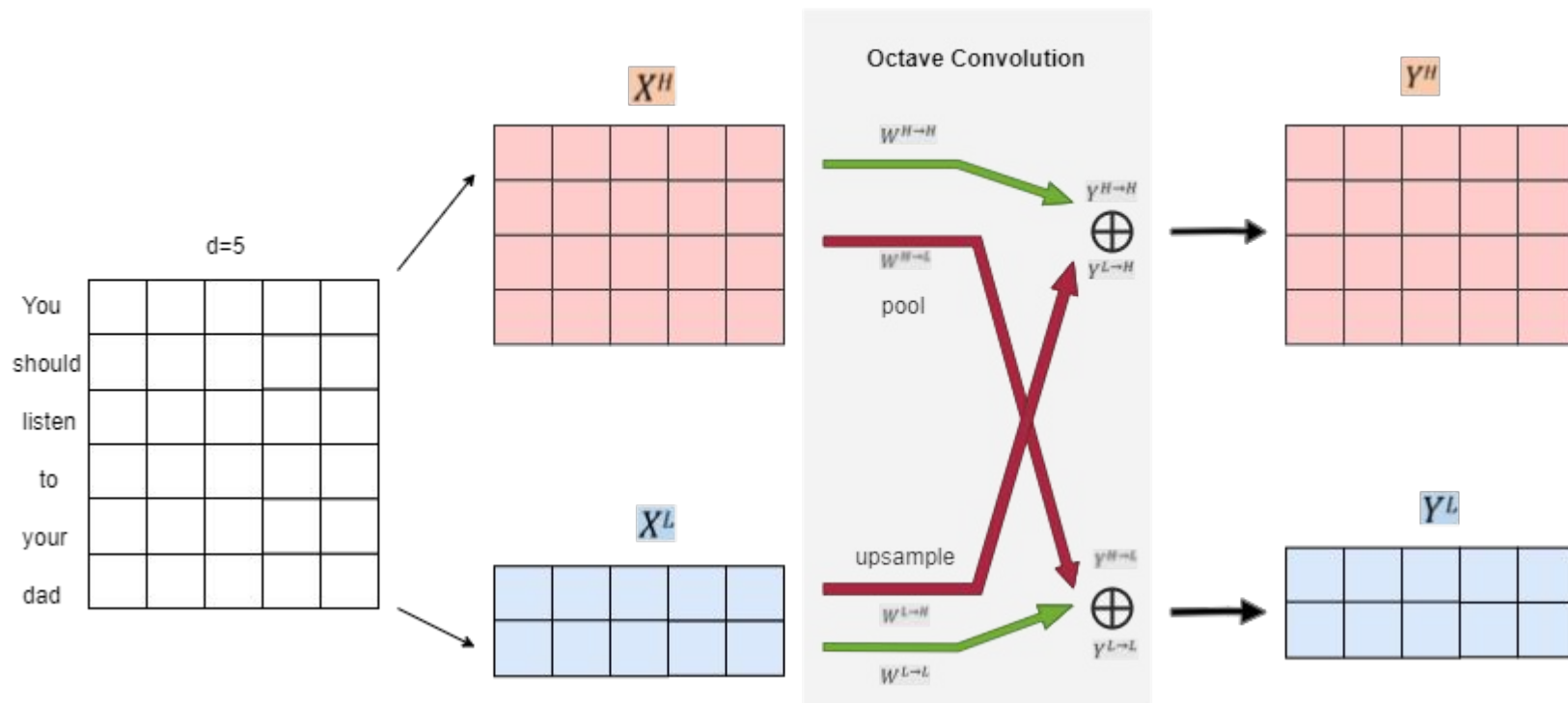
Octave Convolutions



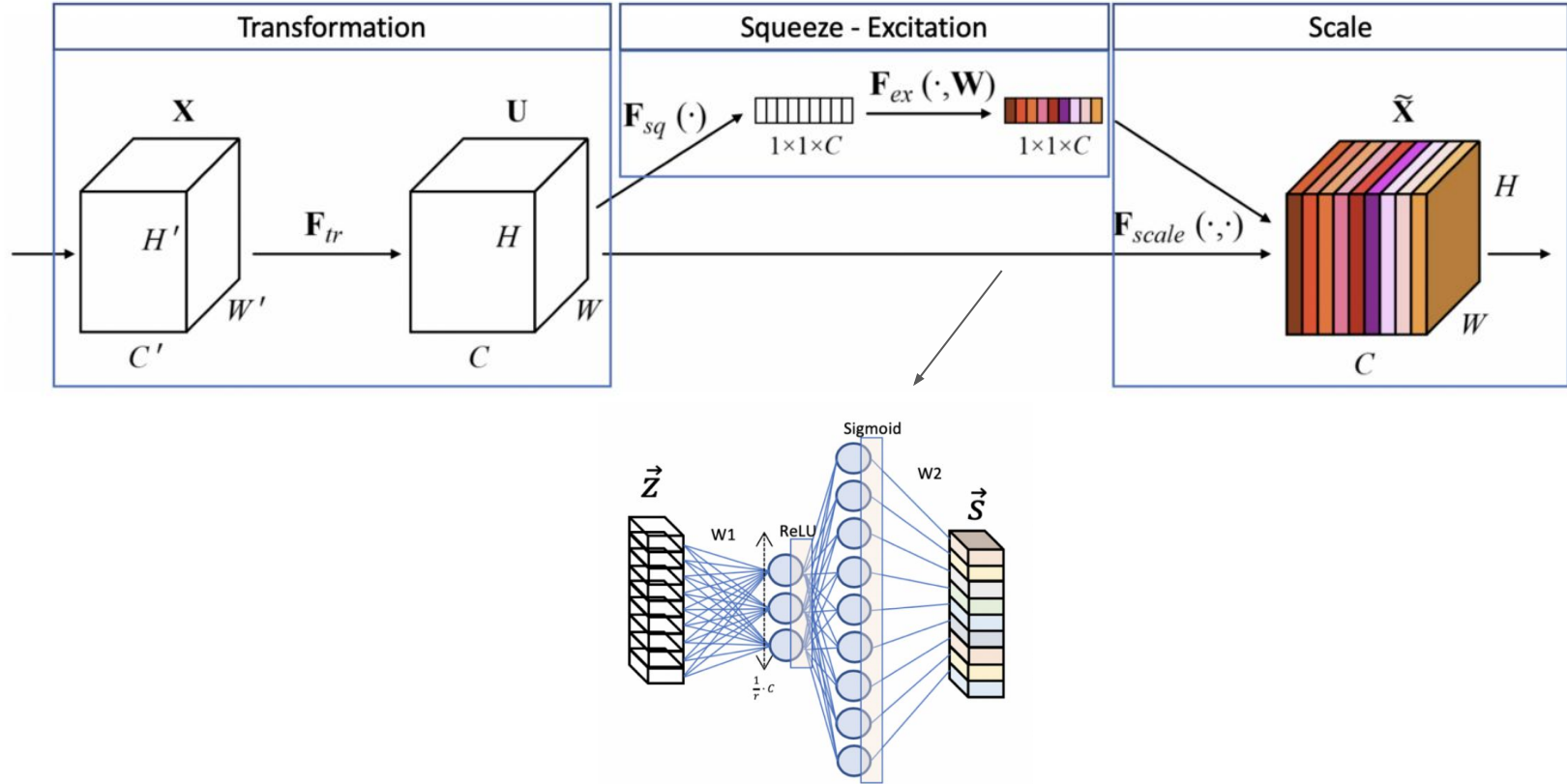
Detailed Design



Octave convolutions on word embeddings



Squeeze & Excitation Block



Conclusion

- An efficient CNN with gating mechanisms for ACSA and ATSA tasks.
- Reproduced results similar to paper findings.
- Octave convolutions and SE Block to improve accuracy and enhance performance or original architecture

References

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- Hu, J., Shen, L., Albanie, S., Sun, G., & Wu, E. (2020). Squeeze-and-Excitation Networks. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 42, 2011-2023.
- <https://github.com/wxue004cs/GCAE>

Thank You