

ADVERSARIAL EXAMPLES AGAINST A BERT ABSA MODEL

FOOLING BERT WITH L33T, MISSPELLIGN, AND PUNCTUATION,

N. HOFER, P. SCHÖTTLE, A. RIETZLER, S. STABINGER AUGUST, 2021

Adversarial Examples Against a BERT ABSA Model



Bidirectional Encoder Representations from Transformers - BERT

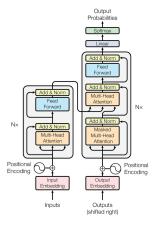


Figure: Transformer Model Architecture (Vaswani et al., 2017)

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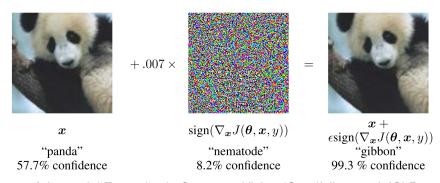


Figure: Adversarial Examples in Computer Vision (Goodfellow et al, ICLR 2015)



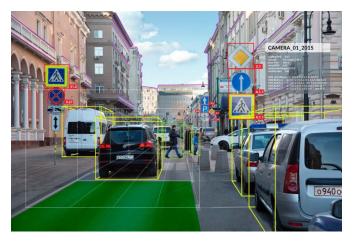


Figure: Object detection in autonomous driving (Source: becominghuman.ai)





Figure: Tweet containing misleading information regarding Covid-19, detected and labeled correctly





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Figure: Tweet containing misleading information regarding Covid-19. Potential problems due to the use of Leet Speak.

1. Fine-Tuning BERT base for ABSA

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Aspect-based Sentiment Analysis

Fine-Tuning BERT base for ABSA

Aspect-based Sentiment Analysis

Dataset: SemEval-2015 Task 12

- Labels contain a set of Entity Attribute -Sentiment
- 23 Entities 9 Attributes 3 Sentiments (POS, NEG, NEU)
- Entity: reviewd entity
- Attribute: particular attribute of an entity
- Sentiment: polarity towards the entity and its attribute

Entity Labels			
1. LAPTOP	13. BATTERY		
2. DISPLAY	14. GRAPHICS		
3. KEYBOARD	15. HARD DISK		
4. MOUSE	16. MULTIMEDIA DEVICES		
5. MOTHERBOARD	17. HARDWARE		
6. CPU 18. SOFTWARE			
7. FANS& COOLING	19. os		
8. PORTS	20. WARRANTY		
9. MEMORY	21. SHIPPING		
10. POWER SUPPLY	22. SUPPORT		
11. OPTICAL DRIVES	23. COMPANY		
Attribute Labels			
A. GENERAL	E. USABILITY		
B. PRICE	F. DESIGN& FEATURES		
C. QUALITY	G. PORTABILITY		
D. OPERATION&	H. CONNECTIVITY		
PERFORMANCE	I. MISCELLANEOUS		

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Aspect-based Sentiment Analysis

The computer is excellent for gaming but I think it is way too expensive!!

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Aspect: Gaming, Sentiment: POS Aspect: Price, Sentiment: NEG

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2. Identify Important Word

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2. Identify Important Word

Leave-One-Out Method

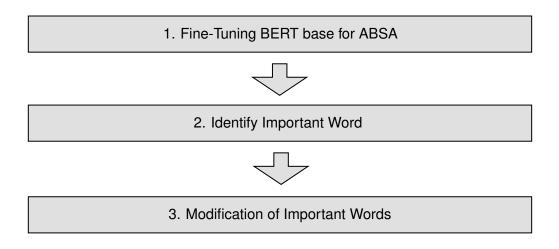
1. Fine-Tuning BERT base for ABSA

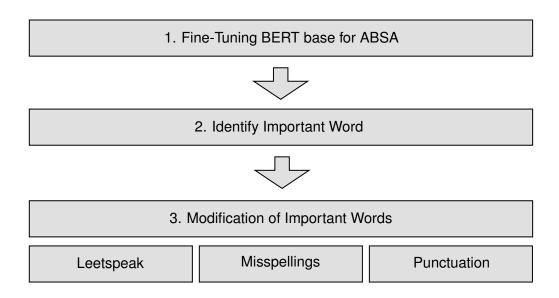


2. Identify Important Word

Leave-One-Out Method







Adversarial Attacks



Objectives

- Semantic Meaning
- Inconspicousness
- Relevance

1. Leetspeak

The computer is excellent for gaming but I think it is way too expensive!!

Aspect: Gaming, Sentiment: POS Aspect: Price, Sentiment: NEG

Original important word: **excellent**Modified important word: **excellent**

The computer is excellent for gaming but I think it is way too expensive!!

Aspect: Gaming, Sentiment: NEG
Aspect: Price, Sentiment: NEG

2. Misspellings

The computer is excellent for gaming but I think it is way too expensive!!

Aspect: Gaming, Sentiment: POS Aspect: Price, Sentiment: NEG

Original important word: **excellent** Modified important word: **ecxellent**

The computer is ecxellent for gaming but I think it is way too expensive!!

Aspect: Price, Sentiment: NEG

3. Punctuation

The computer is excellent for gaming but I think it is way too expensive!!

Aspect: Gaming, Sentiment: POS Aspect: Price, Sentiment: NEG

Original important word: **excellent** Modified important word: **excellent**,

The computer is excellent, for gaming but I think it is way too expensive!!

Aspect: Laptop (general), Sentiment: NEG
Aspect: Gaming, Sentiment: NEG
Aspect: Price, Sentiment: NEG



Perturbation Method	Leetspeak	Misspellings	Punctuation
Dataset A - # of original sentences	943	943	943



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Dataset E - # of changed predictions per sentence	790	259	253



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Dataset D - # of changed predictions total	1066	420	382
Dataset E - # of changed predictions per sentence	790	259	253
Overall Success Rate	47.76%	31.01%	14.95%
Distinct Success Rate	88.07%	70.19%	26.83%

Conclusion & Further Steps



Summary

- BERT can be fooled by input modofications
- Three attack methods:
 - Leetspeak
 - Misspellings
 - Falsly placed Punctuation

Next Steps

- Transferability between Transformer Models
- Using generated adversarial datasets for Adversarial Training

Thank you!



Adversarial Examples Against A BERT ABSA Model -

Fooling BERT with L 33T, Misspelli gn, and Punctuation,

Github: https://github.com/NoraH2004/adv-absa

Email: nora.hofer@uibk.ac.at



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