Fortunately, Discourse Markers Can Enhance Language Models for Sentiment Analysis

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Introduction



Introduction

- Pre trained language models 은 large data에 대해 trained 되고 downstream task 에 대해 작은 labeled data로 fine tuning
- 최근 모델은 annotation effort를 거의 들이지 않아도 넓은 범위의 target task에서 좋은 성능을 나타냄'
- 이 2단계 step을 더 정교하게 설계하면, 특정 interest 에 대한 particular task에서 좋은 starting point model이 됨
 - Domain specific data such as financial or legal documents is beneficial to tasks
 - Additional pretraining of BERT on dialogue data yields better results
 - Product review data에 대해 Sentiment aware pretraining을 더하면 sentiment analysis 성능 향상
 - SpanBERT, PEGASUS
 - → Question answering, summarization 목적에 맞게 여러 단어를 masking해 pretraining
 - → span-extraction: 주어진 context 및 관련 question을 이해하여, 지문과 관련된 질문에 적절한 답변을 지문속에서 도출

Introduction

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■ 본 논문에서는 intermediate training step(inter-training)을 추가

Aligned with a specific target task of interest

- Inter-training approach (similarity between the intermediate task and the target task)
 - 1. Weakly labeled data를 사용하고 완전 같음
 - 2. 살짝다른 task에 labeled data를 사용해 transfer learning
 - 3. No labeled data를 사용한 transfer learning

- Leverage the signal carried by DM to generate large amounts of weakly labeled data for sentiment analysis
- "Happily", "Sadly" convey a positive/negative sentiment
- Inter-training BERT on this data → significant performance
 - Especially on scarce labeled data, zero-shot scenario

Related Works

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- Learning with Discourse Markers
 - Discourse marker: Managing the flow and structure of discourse

Learning signal for the prediction of implicit discourse relations

- Sileo et al.(2020)
 - 처음으로 DM과 downstream task class의 연관관계를 분석
 - Semantic relation이 정해진 두 sentence pair 사이 plausible DM을 예측

Related Works

- Task-aware Language Models
 - For sentiment analysis,
 - Sentiment knowledge를 pretraining task에 포함시키려는 연구 다수
 - 주로 word-level sentiment prediction task에 focus
- This paper suggests a model that incorporates a sentence level sentiment prediction objective

Part 3, Model



- SenDM: A New Sentiment Language Model Training DM-based Sentiment Models
 - C: general corpus of news paper and journal articles
 - L:3 annotators, a list of 173 commonly used DMs
 - Lg: consists of 11 DMs, selected by 3 annotators

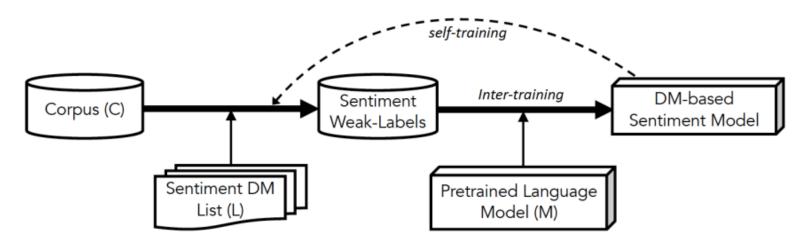


Figure 1: Overview of how DM-based sentiment models are trained.

- Inter-training Details
 - Weakly-labeled data를 사용해 BERT를 fine tuning하는 작업 포함
 - Obtain a total of 1,876,614 weakly labeled sentences

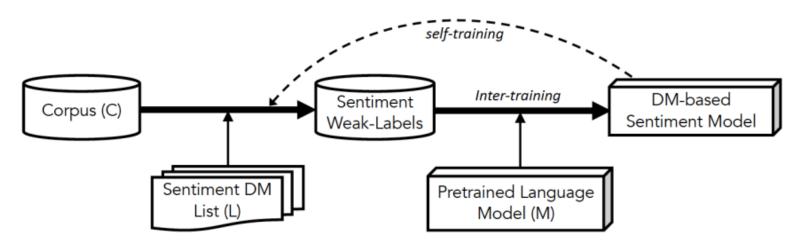


Figure 1: Overview of how DM-based sentiment models are trained.

-Experiments

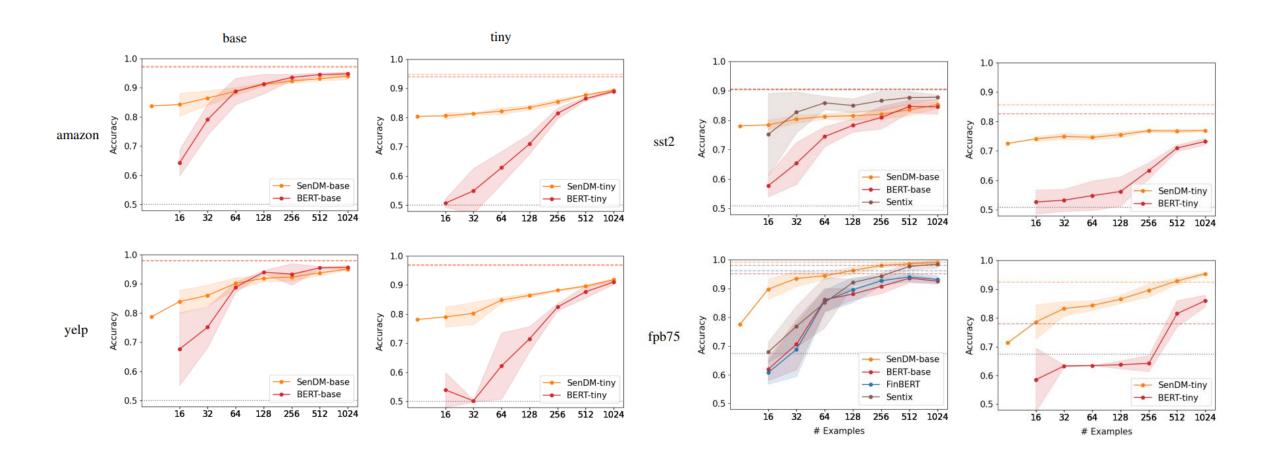
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Dataset

Dataset	Domain	Test set size
amazon	Product reviews	2K
yelp	Business reviews	20K
sst2	Movie reviews	1821
fpb75	Financial news	691

3 scenarios

- Zero-shot: simply use the classification layer obtained from inter-training
- Few-shot: fine-tune inter-trained model with a small sample of n
- Full-data: all training examples are used



- Adapting SenDM to a new domain, finance domain
 - Financial sentiment analysis is a challenging task
 - Import ant task for potential users

- Training Approach
 - Cd: domain specific text corpus
 - Ld: domain specific DM list

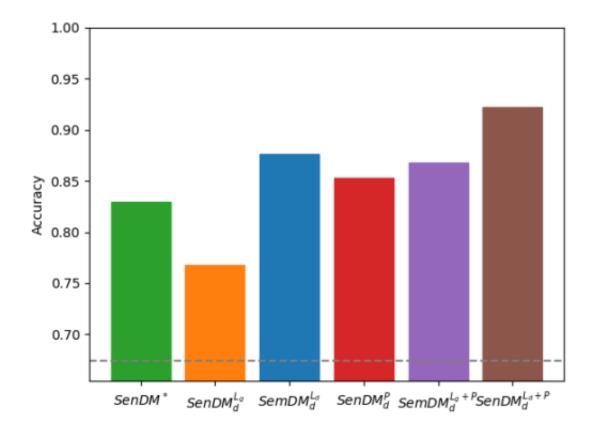
- Training Approach
 - Cd: domain specific text corpus
 - Ld: domain specific DM list

Model name	С	L	M	With self- training
SenDM	C_g	$ig L_g$	BERT	NA
$SenDM_d^{L_g}$	C_d	L_g	SenDM	No
$Sem DM_d^{L_d}$	C_d	L_d	SenDM	No
$SenDM_d^P$	C_d	NA	SenDM	Yes
$SenDM_d^{L_g+P}$	C_d	L_g	SenDM	Yes
$SemDM_d^{L_d+P}$	C_d	L_d	SenDM	Yes

- Domain specific sentiment related DMs
 - 1. Identifying a list of candidate DMs
 - Grouping using NER (e.g., instead of multiple bigrams of the type "on Sept 9th", "on 10/2/2020",... we generate one bigram "on DATE")
 - 2. Using SenDM to select the domain specific DMs out of the candidate list
 - Analyse the sentences to start with the DMs in the candidate list
 - Sample sentences to start with the DM
 - SenDM에 의해 높은 신뢰도로 scored 된 sentence의 경우 감정을 부여

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Results



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Domain specific DMs

Domain	Associated with Posi-	Associated with Neg-
	tive Sentiment	ative Sentiment
General	'fortunately', 'hap-	'admittedly', 'curi-
	pily', 'hopefully',	ously', 'inevitably',
	'ideally', 'luckily',	'sadly', 'unfortu-
	'thankfully'	nately'
Finance	'as ORG', 'at the	according to police',
	event', 'fortunately',	'sadly', 'the problem',
	'hopefully', 'ide-	'the problem is', 'un-
	ally', 'in future',	fortunately', 'worse'
	'in other business',	
	'luckily', 'once com-	
	pleted','ORG CEO',	
	'starting DATE',	
	'thankfully', 'the pro-	
	gram', 'this way', 'to	
	achieve this', 'under	
	his leadership', 'with	
	ORG'	

Sports	'beginning DATE',	'admittedly', 'alas',
	'fortunately', 'in the	'granted', 'ironi-
	future', 'luckily',	cally', 'sadly', 'true',
	'thankfully', 'that	'unfortunately', 'un-
	way'	fortunately for ORG'
Science	'established in DATE',	'admittedly', 'at
	'if necessary', 'if	ORDINAL glance',
	possible', 'if success-	'at times', 'curiously',
	ful', 'luckily', 'that	'even then', 'even
	way', 'to address	worse', 'in part',
	this', 'when possible',	'inevitably', 'paradox-
	'whenever possible',	ically', 'predictably',
	'where possible',	'regrettably', 'the
	'with this approach'	problem', 'there was',
		'too often', 'unsurpris-
		ingly', 'without it',
		'women'

Conclusion

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Contribution

- 1. A novel approach that leverages sentiment signals of discourse markers for creating sentiment-aware language models that significantly outperform prior models.
- 2. A new method for enhancing domain-specific sentiment classification, based on statistical analysis of discourse markers in a domain-specific corpus.
- 3. A large dataset of weakly labeled sentences from Wikipedia, and a code for generating weakly labeled data from a given text corpus

Conclusion

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Conclusion

- This approach can be easily adapted to other languages
- Leveraging DM to create task-specific language models can be potentially applied to other tasks
- This approach shows how to enhance zero-shot learning performance