

Multi-Lingual Theme Prediction of Customer Reviews Using Deep Pre-Trained Embeddings

Team 06

Michael Sorg

08.05.2019





### Task

- Train a category prediction model on the Amazon product review dataset based on XLING embeddings per review
- Evaluate on German without training on German data
- Fine-tune and evaluate on our Organic Dataset for relevance, entity, and attribute classification



## Word embeddings

#### **BERT**

"BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding"

- Birectional transformer network
- Multilingual version exists

#### **XLING**

"Learning cross-lingual sentence representations via a multi-task dual-encoder model"

- Compute dense word vectors (512 dimensions) from sentences
- Embeddings can be fine-tuned (transfer learning)
- Multiple languages

#### **Datasets**



#### Kaggle Amazon Fine Food

(568454,	10)								
Id	ProductId	UserId	ProfileName	${\tt HelpfulnessNumerator}$	${\tt HelpfulnessDenominator}$	Score	Time	Summary	Text
0 1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	1	5	1303862400	Good Quality Dog Food	I have bought several of the Vitality canned d
1 2	B00813GRG4	A1D87F6ZCVE5NK	dli pa	0	0	1	1346976000	Not as Advertised	Product arrived labeled as Jumbo Salted Peanut

#### Amazon Review Multilingual

	8993, 15) marketpla	ce custo	omer_id	review_id	product_id	product_parent	product_title	product_category	star_rating	helpful_votes	total_votes	vine	verified_purchase	review_headline	review_body	review_date
0	[	DΕ	10133	RVOG49N0H1FB6	B004TACMZ8	569741360	Bosch GMS120 Ortungsgerät digital multi-Scanner	Home Improvement	5	0	0	N	Υ	Super	Delivery took a little bit more then i expecte	2014-08-01
1	[	DΕ	19612	RNCMD60LTP4HM	1846071224	785505948	The Wheels On The Bus: Favourite Nursery Rhyme	Books	5	1	1	N	Y	Great compilation	We enjoy listening to the song as preparation	2014-12-04

#### Organic dataset

(8	823, 12)											
	Author_ID	Author_name	Comment_number	Sentence_number	Domain_Relevance	Sentiment	Entity	Attribute	Sentence	Source_file	Annotator	Aspect
0	Justin-Ma	Justin Ma	521	1	0	NaN	NaN	NaN	Thanks for the thoughtful response.	quora.json	sumit	nan-nan
1	Justin-Ma	Justin Ma	521	2	0	NaN	NaN	NaN	I think we actually have a lot of common groun	quora.json	sumit	nan-nan
2	Justin-Ma	Justin Ma	521	3	0	NaN	NaN	NaN	All I want to emphasize are my main points: Pr	quora.json	sumit	nan-nan
3	Justin-Ma	Justin Ma	521	4	9	р	cg	pp	Industrialization is everything about producti	quora.json	sumit	cg-pp



### Questions

- Predict which categories / features ?
- Filter out English reviews in the German dataset?
- Also train XLING embeddings or leave them untouched?



### To-do until 22.05

Implement training and evaluation pipeline



Multi-Lingual Theme Prediction of Customer Reviews Using Deep Pre-Trained Embeddings

Team 06

Michael Sorg

22.05.2019



# Amazon Multilingual Dataset

ır:	1		_	-		
a:	1					
h-c	n:	1				
g:	2					
:0:	2					
v:	2					
r:	2					
s:	4					
t:	5					
·o:	8					
i:						
1:	15					
i:	17					
`u:	17					
1:	18					
u:	21					
1:	21					
k:	23					
ır:	24					
d:	24					
t:	32					
q:	33					
t:	54					
10:	81					
v:	81					

678993, 15)															CS:
marketplace	customer_i	d review_id	product_id	product_parent	product_title	product_category	star_rating	helpful_votes	total_votes	vine	verified_purchase	review_headline	review_body	review_date	lt ro
0 DE	1013	3 RVOG49N0H1FB6	B004TACMZ8	569741360	Bosch GMS120 Ortungsgerät digital multi-Scanner	Home Improvement	5	C	) (	N	Υ	Super	Delivery took a little bit more then i expecte	2014-08-01	vi
1 DE	1961:	2 RNCMD6OLTP4HM	1846071224	785505948	The Wheels On The Bus: Favourite Nursery Rhyme	Books	5	1	1	N	Y	Great compilation	We enjoy listening to the song as preparation	2014-12-04	ru
2 DE	1961:	2 R4AUOBI8YC0R8	0375851569	516548029	Dr. Seuss's Beginner Book Collection	Books	5	C	) C	N	Υ	Great Collection	Very great compilation. Interesting story and	2014-12-04	pl
3 DE	1967	7 R1VSHIJ1RHIBTE	B0060SVG54	302116447	Zwei an einem Tag	Video DVD	5	C	) (	N	Υ	Guter Verfilmung	Den Film habe ich bereits vor lesen des Buches	2015-07-16	id:
4 DE	1999	R3JBLVALWSLCZD	B00EYQ6CVC	368843515	Dr. House - Die komplette Serie, Season 1-8 (L	Video DVD	5	g	) 14	N	Y	Kauft diese Box!	Die Box ist super verarbeitet, sieht gut aus b	2014-02-08	pt:
															da: cy: af:

nl: 179 es: 180 fr: 307 en: 48660 de: 628461

it: 150 ca: 158



#### Review

- For each dataset create tensorflow dataset loader (using dataset api)
- Amazon Reviews dataset does not fit on disk and into ram → change runtime in colab
- Todo: look into TFRecord



### Review

☑ Full training pipeline with dummy architecture



## Roadmap

- 1. network architecture search (train/test on amazon multilingual en/de)
  - w/o training xling embeddings
- 2. Additionally use full amazon reviews dataset
- 3. Test and fine-tune on organic dataset



### Questions

Docker instances → Google account required

(5684	154,	10)								_
1	[d	ProductId	UserId	ProfileName	${\tt HelpfulnessNumerator}$	${\it HelpfulnessDenominator}$	Score	e Time	Summary	Text
0	1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	1	5	5 1303862400	Good Quality Dog Food	I have bought several of the Vitality canned d
1	2	B00813GRG4	A1D87F6ZCVE5NK	dli pa	0	0		1 1346976000	Not as Advertised	Product arrived labeled as Jumbo Salted Peanut
2	3	B000LQOCH0	ABXLMWJIXXAIN	Natalia Corres "Natalia Corres"	1	1	4	4 1219017600	"Delight" says it all	This is a confection that has been around a fe
3	4	B000UA0QIQ	A395BORC6FGVXV	Karl	3	3	2	2 1307923200	Cough Medicine	If you are looking for the secret ingredient i
4	5	B006K2ZZ7K	A1UQRSCLF8GW1T	Michael D. Bigham "M. Wassir"	0	0		5 1350777600	Great taffy	Great taffy at a great price. There was a wid

#### De: 600k, UK: 1,7m, US: 7M

	marketpla	ice custo	mer_id	review_id	product_id	product_parent	product_title	product_category	star_rating	helpful_votes	total_votes	vine	verified_purchase	review_headline	review_body	review_date
0	ı	DE	10133	RVOG49N0H1FB6	B004TACMZ8	569741360	Bosch GMS120 Ortungsgerät digital multi-Scanner	Home Improvement	5	0	0	N	Y	Super	Delivery took a little bit more then i expecte	2014-08-01
1	ı	DE	19612	RNCMD6OLTP4HM	1846071224	785505948	The Wheels On The Bus: Favourite Nursery Rhyme	Books	5	1	1	N	Y	Great compilation	We enjoy listening to the song as preparation	2014-12-04

(88	323, 12)											
	Author_ID	Author_name	Comment_number	Sentence_number	Domain_Relevance	Sentiment	Entity	Attribute	Sentence	Source_file	Annotator	Aspect
0	Justin-Ma	Justin Ma	521	1	0	NaN	NaN	NaN	Thanks for the thoughtful response.	quora.json	sumit	nan-nan
1	Justin-Ma	Justin Ma	521	2	0	NaN	NaN	NaN	I think we actually have a lot of common groun	quora.json	sumit	nan-nan
2	Justin-Ma	Justin Ma	521	3	0	NaN	NaN	NaN	All I want to emphasize are my main points: $\mbox{\rm Pr}$	quora.json	sumit	nan-nan
3	Justin-Ma	Justin Ma	521	4	9	р	cg	pp	Industrialization is everything about producti	quora.json	sumit	cg-pp



Multi-Lingual Theme Prediction of Customer Reviews Using Deep Pre-Trained Embeddings

Team 06

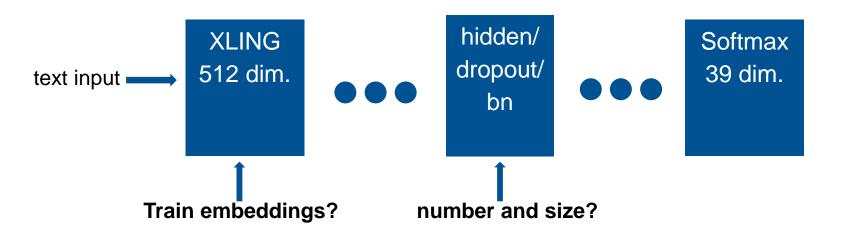
Michael Sorg

05.06.2019



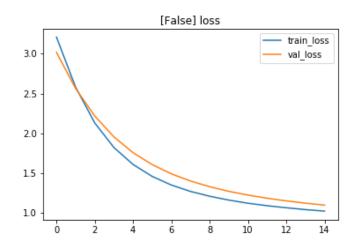


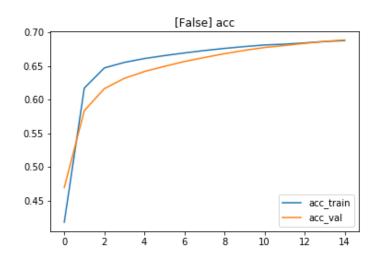
#### Network search





- Amazon Multilingual UK with 100k examples
- One hidden layer (only xling + softmax)

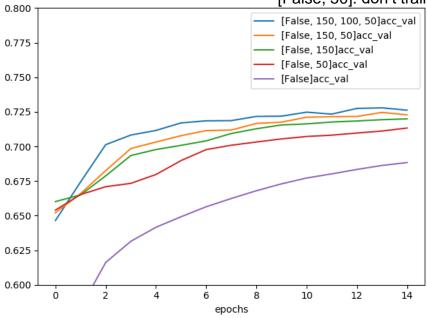






Amazon Multilingual UK with 100k examples

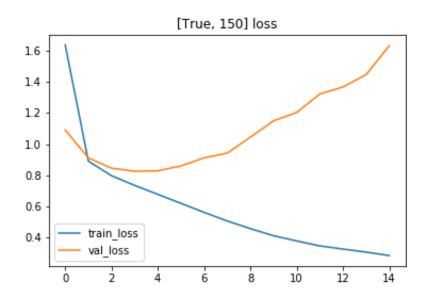
[False, 50]: don't train embeddings, fc with 50 units + softmax

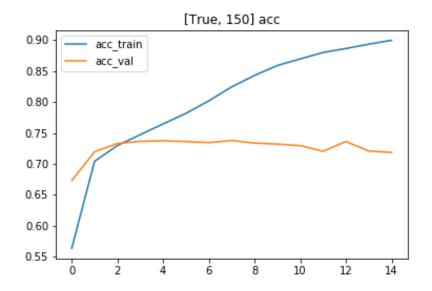




self.train\_op = tf.train.AdamOptimizer(epsilon=0.1).minimize(self.loss)

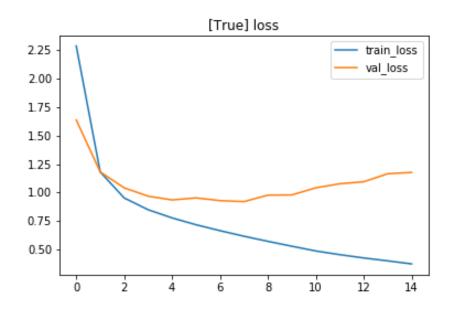
Train also xling embeddings (100k samples)

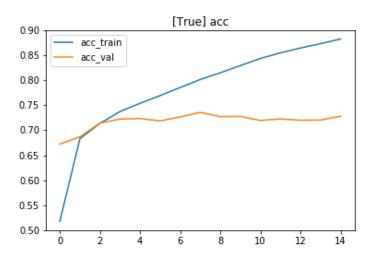






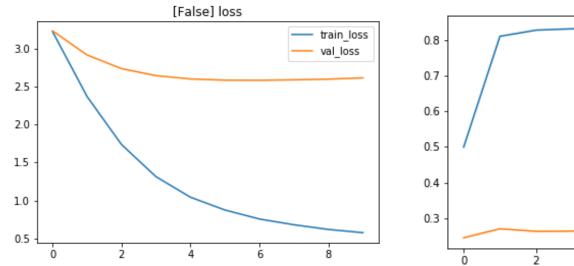
- Train also xling embeddings
- Add dropout to prevent overfitting

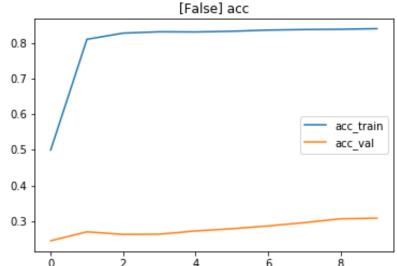






Amazon Multilingual US with 100k examples









O	de.product_category.value_co	ounts()	C÷	Mobile_Apps	1467128
	Video DVD Music Books Mobile_Apps Digital_Video_Download Digital_Music_Purchase Toys Digital_Ebook_Purchase PC Camera Wireless Electronics Video Sports Video Games Watches Shoes Home Musical Instruments Baby Home Improvement Home Entertainment Office Products Personal_Care_Appliances Automotive Lawn and Garden Luggage Kitchen Furniture	279068 160588 63784 54709 25124 21554 18602 12872 12250 5421 4441 4035 2927 2034 1706 1575 1517 1454 1094 810 672 605 412 411 410 397 247 120 93	D	Digital_Ebook_Purchase Video DVD Digital_Video_Download Books Music Digital_Music_Purchase Toys PC Video Home Entertainment Wireless Camera Video Games Electronics Musical Instruments Watches Tools Shoes Baby Sports Home Improvement Outdoors Office Products Home Kitchen Lawn and Garden Health & Personal Care Automotive Mobile_Electronics Apparel	1242173 1093612 1051622 836136 776810 107461 57465 56817 46697 36298 22637 16911 15398 11421 10914 10537 7475 7342 5868 4123 3698 3184 2306 1998 1834 1204 1066 223 184 121
	Health & Personal Care	37		Luggage	78
	Software	19		Beauty	52
	Pet Products	2		Software	52
	Grocery	2		Grocery	18
	Beauty	1		Personal_Care_Appliances	9
	Name: product_category, dt	ype: int64		Furniture	8
				Pet Products	5
				2012-12-22	1

f = us[0:100000]
f.product\_category.value\_counts()

43422 Music Books 23948 Video 17702 Video DVD 14911 Toys 10 Office Products Tools Name: product\_category, dtype: int64

f2 = de[0:100000]	
<pre>f2.product_category.value_counts()</pre>	

Ľ→

Video DVD	406	560
Music	179	961
Mobile_Apps	119	905
Digital_Video_Download	63	387
Books	57	727
Toys	36	533
Digital_Music_Purchase	36	509
Digital_Ebook_Purchase		579
PC		396
Wireless		376
Camera		310
Electronics		779
Video Games		104
Watches		324
Shoes		301
Sports		267
Musical Instruments		263
Home		244
Baby		177
Home Entertainment		133
Home Improvement	1	124
Automotive		80
Office Products		78
Lawn and Garden		61
Personal_Care_Appliances		61
Video		55
Luggage		51
Furniture		28
Kitchen		16
Health & Personal Care		9
Software		2
Name: product_category, o	itype:	int6



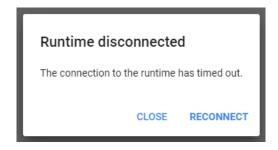
#### How to find architecture?

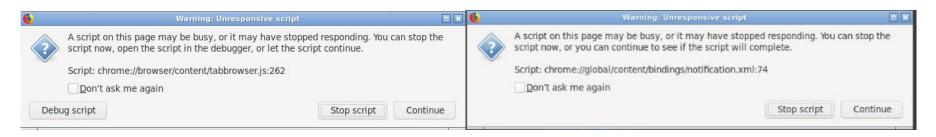
- Use more advanced search techniques (Progressive Neural Architecture Search (PNAS), Efficient Neural Architecture Search (ENAS), Reinforcement learning)
- Tradeoff between time and size of training set?



#### Infrastructure Issues

Training for more than 30 min. fails in most cases





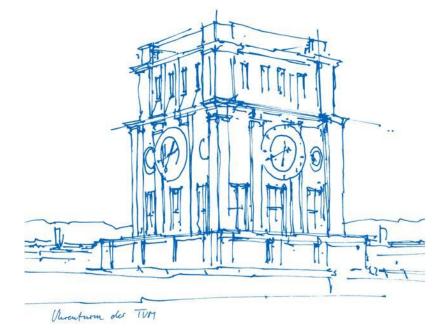


Multi-Lingual Theme Prediction of Customer Reviews Using Deep Pre-Trained Embeddings

Team 06

Michael Sorg

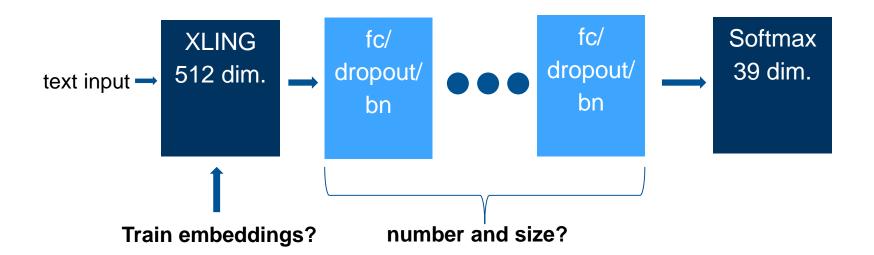
19.06.2019





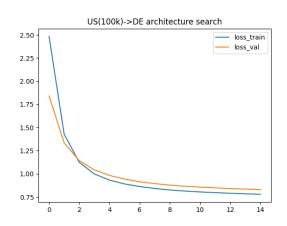
#### Network search

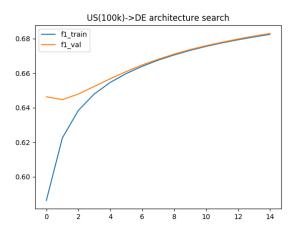
Task: train on english data only – test on German data

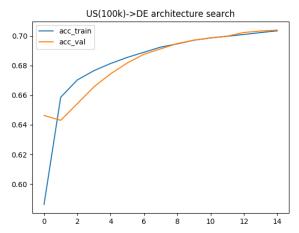




One hidden layer (only xling + softmax)





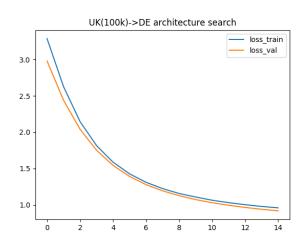


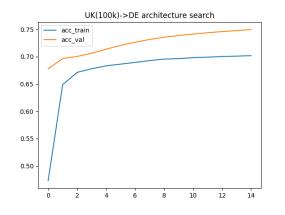


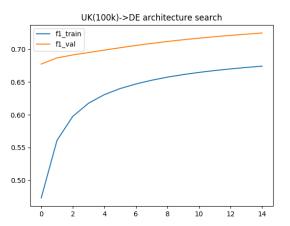
Test data DE (100k)

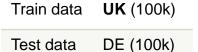


One hidden layer (only xling + softmax)

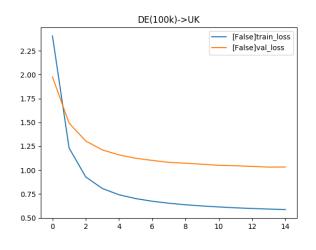


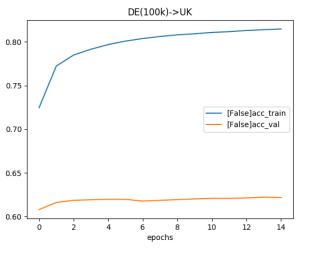


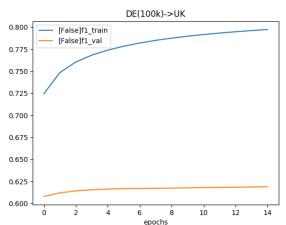




One hidden layer (only xling + softmax)







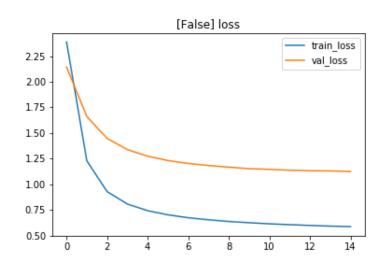


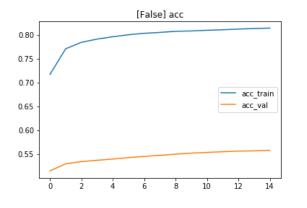


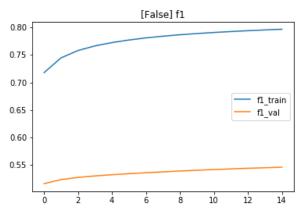
Test data UK (100k)

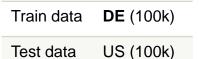


One hidden layer (only xling + softmax)









product_category.value_co.	ounts()	uk.product_category.value_
ideo DVD	41048	Video DVD
usic	23890	Music
Books	9353	Digital_Ebook_Purchase
4obile_Apps	7998	Books
igital_Video_Download	3768	Mobile Apps
igital_Music_Purchase	3116	Digital Video Download
ovs	2729	Digital Music Purchase
igital_Ebook_Purchase	1870	Toys
C	1782	PC
amera	835	Camera
reless 654		Wireless
lectronics 566		Electronics
'ideo	411	Baby
ports	306	Video
'ideo Games	247	Video Games
latches	238	Watches
ome	218	Home
hoes	202	Musical Instruments
usical Instruments	164	Sports
Baby	121	Shoes
łome Improvement	103	Home Improvement
Home Entertainment	82	Office Products
Automotive	70	Automotive
Lawn and Garden	57	Lawn and Garden
Office Products	52	Health & Personal Care
Personal_Care_Appliances	49	Home Entertainment
Luggage	27	Software
Kitchen	20	Personal Care Appliances
Furniture	15	Kitchen

us.product	_category.	value_c	ounts()

	( )
Mobile Apps	21056
Digital Ebook Purchase	18173
/ideo DVD	15949
Digital_Video_Download	15427
Books	12097
Music	11148
Digital_Music_Purchase	1488
Toys	820
PC	766
/ideo	666
Home Entertainment	512
vireless	304
Camera	272
/ideo Games	226
Musical Instruments	167
Electronics	160
Natches	151
Γools	119
Shoes	111
Baby	100
Sports	63
Outdoors	52
Home Improvement	51
Home	35
Office Products	26
Kitchen	21
Health & Personal Care	15
Lawn and Garden	10
Mobile_Electronics	5
Automotive	4
Luggage	1
Personal_Care_Appliances	1
Grocery	1
Apparel	1
Software	1
Beauty	1

Name: product\_category, dtype: int64



Health & Personal Care

Name: product\_category, dtype: int64

Software

Kitchen

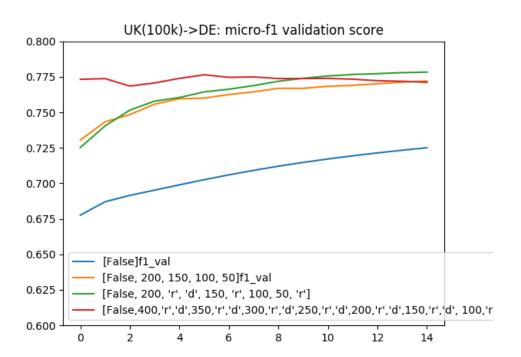
Luggage

Pet Products

Name: product\_category, dtype: int64



### Deeper architectures



Train data **UK** (100k)

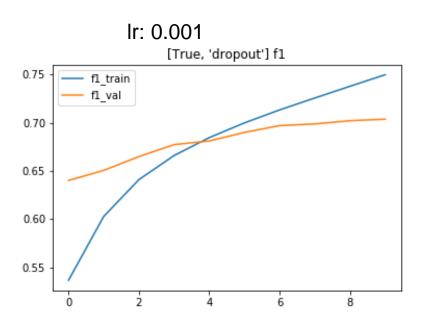
Test data DE (100k)

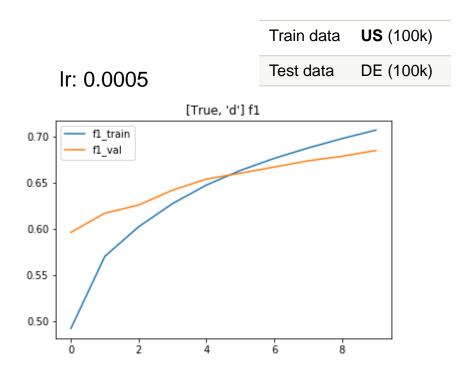




### Train Embeddings

- Quickly overfits
- Even with dropout and reduced learning rate







### **Progress**

- Colab issues -> switched to google cloud compute engine
- Training 100k examples takes 5 min per epoch (15 epochs ~45 min.)
- Training 1 Million examples takes 30 min per epoch (15 epochs ~7-8 hours)
- Problem: balance between training size and computation time



## Roadmap

- Continue with architecture search
- Balanced training
- Run baseline architecture on organic dataset
- Fine-tune on organic dataset



# Multi-Lingual Theme Prediction of Customer Reviews Using

Deep Pre-Trained Embeddings

Team 06

Michael Sorg

03.07.2019

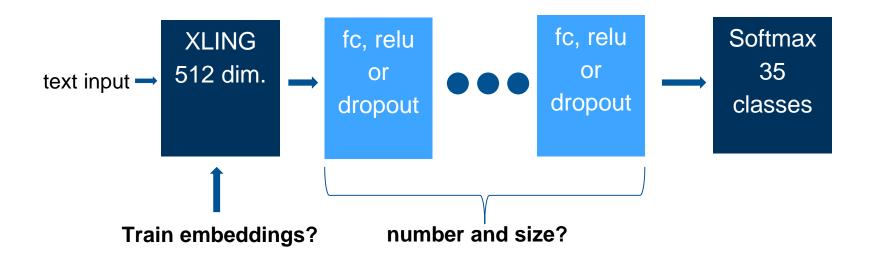




#### Network search

Task: category prediction on english reviews

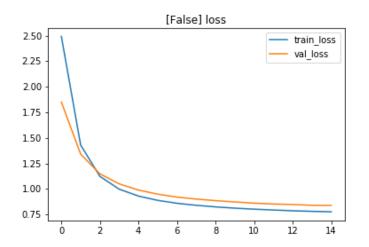
– test on German

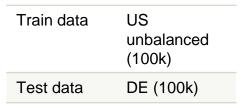


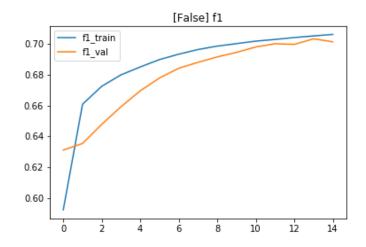


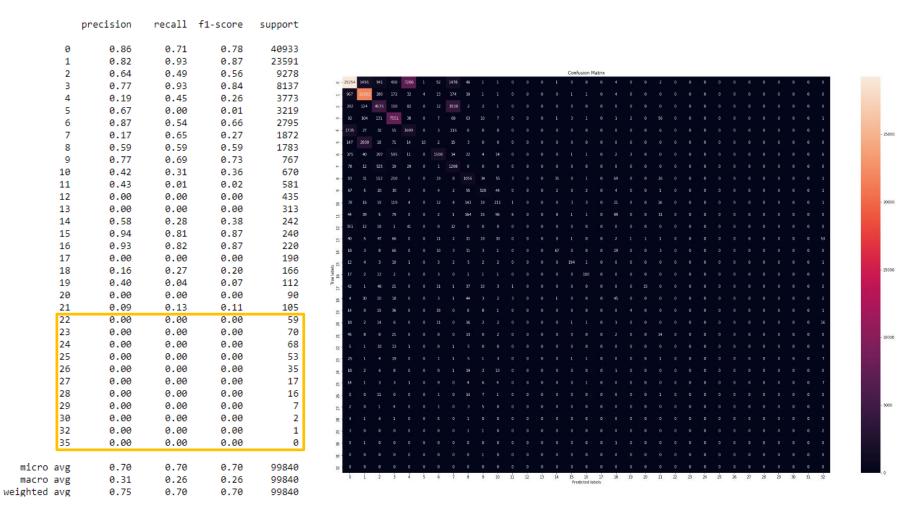
# Baseline experiments

- One hidden layer (only xling + softmax)
- Unbalanced training







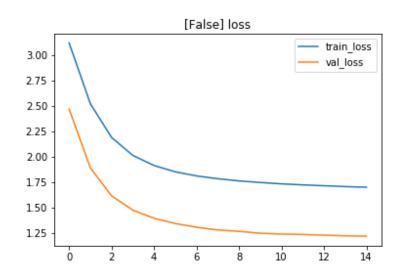


NLP for Opinion Mining | Group 6 | Mult

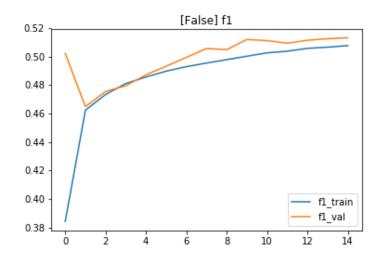


# Baseline experiments

- One hidden layer (only xling + softmax)
- balanced training



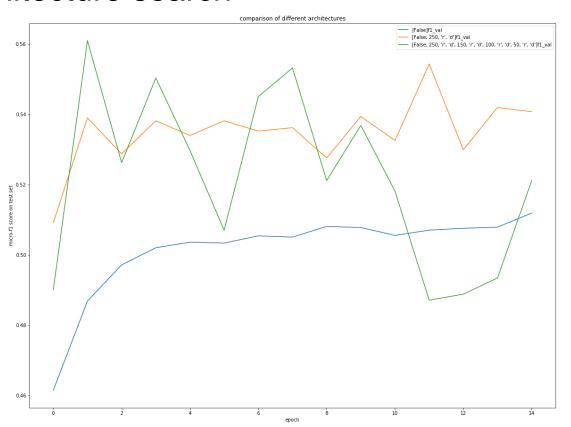
Train data	US balanced (100k)
Test data	DE (100k)



		precision	recall	f1-score	support
	0	0.90	0.38	0.54	40928
	1	0.84	0.84	0.84	23762
	2	0.73	0.31	0.44	9446
	3	0.92	0.60	0.73	7895
	4	0.18	0.32	0.23	3748
	5	0.29	0.29	0.29	3158
	6	0.75	0.60	0.66	2687
	7	0.15	0.71	0.25	1858
	8	0.58	0.49	0.53	1843
	9	0.58	0.72	0.64	801
	10	0.38	0.35	0.37	666
	11	0.27	0.35	0.30	553
	12	0.02	0.61	0.03	474
	13	0.15	0.12	0.13	286
	14	0.07	0.47	0.13	250
	15	0.60	0.92	0.73	218
	16	0.69	0.89	0.78	237
	17	0.02	0.01	0.01	246
	18	0.11	0.42	0.17	176
	19	0.10	0.21	0.13	114
	20	0.03	0.07	0.04	88
	21	0.03	0.50	0.06	86
	22	0.05	0.37	0.09	68
	23	0.13	0.36	0.19	64
	24	0.07	0.51	0.13	53
	25	0.16	0.59	0.25	51
	26	0.07	0.63	0.13	38
	27	0.03	0.35	0.05	23
	28	0.00	0.00	0.00	17
	29	0.00	0.00	0.00	5
	30	0.00	0.00	0.00	1
	31	0.00	0.00	0.00	0
	32	0.00	0.00	0.00	0
	33	0.00	0.00	0.00	0
accur				0.51	99840
macro	_	0.26	0.38	0.26	99840
weighted	avg	0.77	0.51	0.58	99840



### Architecture search



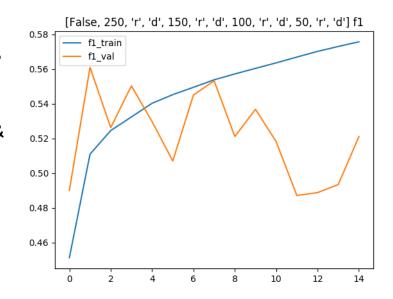
Train data	US balanced (300k)
Test data	DE (300k)

NLP for Opinion Mining | Group 6 | Multi-Lingual Theme Prediction | Michael Sorg



### **Problems**

- Not much learning f1 score doesn't improve much after 2-3 epochs
- zig-zagging f1 score
- Overfitting when using deeper networks (even with dropout & regularization)





## **Progress**

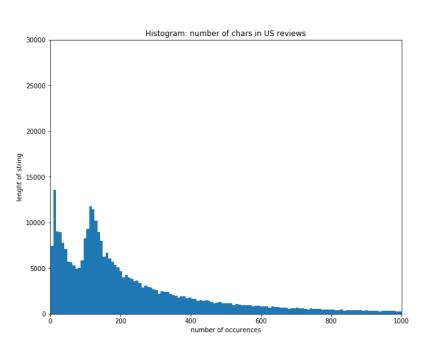
- Tf-idf approach with KNN/SVM (val\_f1: 55%, test\_f1: 40%)
- size of training set doesn't improve performance

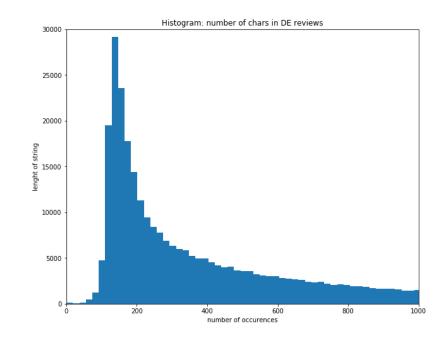


# **Progress**

IIS madian: 160 A

• German reviews are longer







# Roadmap

- continue with Architecture search
- relevance, entity, and attribute classification on organic dataset
- Fine-tune on organic dataset



Multi-Lingual Theme Prediction of Customer Reviews Using

Deep Pre-Trained Embeddings

Team 06

Michael Sorg

17.07.2019





### Outline

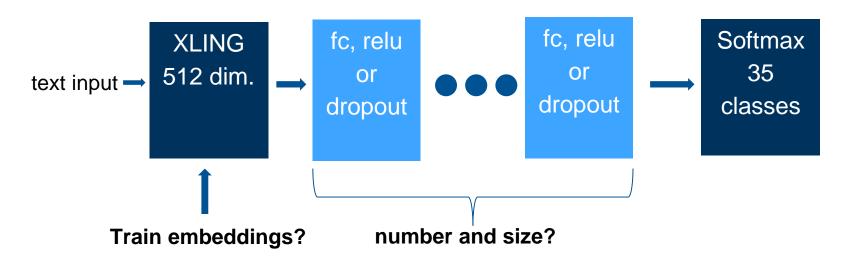
Task description

- Amazon Reviews
  - Unbalanced training
  - Balanced training
  - Architecture search
- Organic Dataset
  - w/o fine-tuning



#### Task

- category prediction on amazon reviews based on XLING embeddings
- Evaluate on german reviews without training on german data
- Fine-tune on organic dataset for relevance, attribute and entity classification





## Baseline experiment

- no hidden layer (only xling + softmax)
- Unbalanced training
- Some classes are never predicted
- Micro-f1 score on validation set: ~ 70%

		precision	recall	f1-score	support
	0	0.86	0.71	0.78	40933
	1	0.82	0.93	0.87	23591
	2	0.64	0.49	0.56	9278
	3	0.77	0.93	0.84	8137
	4	0.19	0.45	0.26	3773
	5	0.67	0.00	0.01	3219
	6	0.87	0.54	0.66	2795
	7	0.17	0.65	0.27	1872
	8	0.59	0.59	0.59	1783
	9	0.77	0.69	0.73	767
	10	0.42	0.31	0.36	670
	11	0.43	0.01	0.02	581
	12	0.00	0.00	0.00	435
	13	0.00	0.00	0.00	313
	14	0.58	0.28	0.38	242
	15	0.94	0.81	0.87	240
	16	0.93	0.82	0.87	220
	17	0.00	0.00	0.00	190
	18	0.16	0.27	0.20	166
	19	0.40	0.04	0.07	112
	20	0.00	0.00	0.00	90
	21	0.09	0.13	0.11	105
	22	0.00	0.00	0.00	59
	23	0.00	0.00	0.00	70
	24	0.00	0.00	0.00	68
	25	0.00	0.00	0.00	53
	26	0.00	0.00	0.00	35
	27	0.00	0.00	0.00	17
	28	0.00	0.00	0.00	16
	29	0.00	0.00	0.00	7
	30	0.00	0.00	0.00	2
	32	0.00	0.00	0.00	1
	35	0.00	0.00	0.00	0
_		0.70	0.70	0.70	00040
	avg	0.70	0.70	0.70	99840 99840
o d	avg	0.31 0.75	0.26 0.70	0.26 0.70	99840
u.	avg	0.75	0.70	0.70	22040

micro	avg	0.70	0.70	0.70	99840
macro	avg	0.31	0.26	0.26	99840
weighted	avg	0.75	0.70	0.70	99840



## Baseline experiment

- no hidden layer (only xling + softmax)
- balanced training
- Predictions for some classes are very bad
- Micro-f1 score on validation set: ~ 51%

		precision	recall	f1-score	support
	0	0.90	0.38	0.54	40928
	1	0.84	0.84	0.84	23762
	2	0.73	0.31	0.44	9446
	3	0.92	0.60	0.73	7895
	4	0.18	0.32	0.23	3748
	5	0.29	0.29	0.29	3158
	6	0.75	0.60	0.66	2687
	7	0.15	0.71	0.25	1858
	8	0.58	0.49	0.53	1843
	9	0.58	0.72	0.64	801
	10	0.38	0.35	0.37	666
	11	0.27	0.35	0.30	553
	12	0.02	0.61	0.03	474
	13	0.15	0.12	0.13	286
	14	0.07	0.47	0.13	250
	15	0.60	0.92	0.73	218
	16	0.69	0.89	0.78	237
	17	0.02	0.01	0.01	246
	18	0.11	0.42	0.17	176
	19	0.10	0.21	0.13	114
	20	0.03	0.07	0.04	88
	21	0.03	0.50	0.06	86
	22	0.05	0.37	0.09	68
	23	0.13	0.36	0.19	64
	24	0.07	0.51	0.13	53
	25	0.16	0.59	0.25	51
	26	0.07	0.63	0.13	38
	27	0.03	0.35	0.05	23
	28	0.00	0.00	0.00	17
	29	0.00	0.00	0.00	5
	30	0.00	0.00	0.00	1
	31	0.00	0.00	0.00	0
	32	0.00	0.00	0.00	0
	33	0.00	0.00	0.00	0
accui	racv			0.51	99840
macro		0.26	0.38	0.26	99840
ghted	_	0.77	0.51	0.58	99840
S	2.0			2.30	22210

macro weighted



#### Architecture search

- Training xling embeddings leads to overfitting
- Problems when going deeper:
  - f1 score doesn't improve much after 2-3 epochs
  - Overfitting when using deeper networks (more than 2-3 layers)

	micro-f1 on test set
Tf-idf + svm	44 %
Baseline model	51 %
Best model (xling,150,relu,dropout,softmax)	57 %



#### Results

- Increased data set size doesn't lead to better performance
- Including review headline has no effect
- Filtering out English reviews from the German test set has no effect



# Organic dataset

• Fine-tuning increases performance

	f1 score	F1 score with fine-tuning
relevance	74 %	77 %
entity	51 %	57 %
attribute	44 %	50 %