Pre-trained 1-	Date. ()
	inguege models 1)
	sparse representation
embeddings	Dense representation
······································	NN (simple) like word Evec ()
	The Company like volla area
	NN (complex) - Contratuction 1 (7)
	NN (complex) - Gontactualized (2)
(D) not	continuing Contest information (a unique word only has one ve
	g contest substruction (a unique word only has one ve
in 6	
<u> </u>	), each word in each contact has a rector
* ELMO	(from RNN tamily)
	words to CSIAN International
Forward	and backward
* BERT	and backward home separate informed are concatinated.  The end
multi-hers	d self attention, bidirectional transformer  ( hering for a
	Diametianal 1
	hewing forward and backy 1 . P
2.1	Same Lines
- C (wyer	(hewing forward and bachward into at the Same Line)  r (2 headed) bidinectionally connected them transformer
	J transforma
- train	Scenerio 8
	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \
	(Pre-tran) (hinetune)
25Lp	(fre-tran) (fine-time)
having only	y a few labeled data, we train and on dif I
having only	

ect:	
Month. Date. ()	
. In NLP, test data is kind of a supervised (latted) since we	e hery's
hove 4. nest word 1 " 1 2.	0
have the rest word - also called Semi-supervised	
SERT Base -> 12 layers BERT, ->24 levers	
BERTBase > 12 layers BERT - 524 lengers encoder encoder	
We give 2	5
we give 2 sentences to the model SHS2 8 (US) SI (Sep) S (back to back) Semi supervised part	9
Semi supervised part	
1) marked language undel	
I) masked language model : remains some percentege of words to	, be
F	
(15% of input words usually)	10
@ neat sentence prediction: given	-10
and So we want the	e ccls)
(151. of input words usually)  2 neat sentence prediction: giving S, and S, we want the token to output it S2 is the next s  5	unhera al
8	
<u>C C U</u>	
Supervised part	
have tunning the weightes I covered in the previous part for asp  Dentance poir classification - relation between 2.1	15
The tunning the weightes leavned in the previous part for ass	with track
Dentence pair classification relation	
Jentence poir classification school between 2 to tasks @ single sentence Classification text classifice from @ Queenstion Answering	4
(3) On 1: 1	<b>Y</b>
avenstion Answering	
A single centure sequence labeling	
	20
Disput 2 sentences output in als taken adout	
input a sentinces but put the als talken alput	
paraphrase identification, answer retrival, testual entailment	
(ilumented is in ) (interpretation is	
paraphrase identification, answer retrival, testual entailment	
Dentence classification input the buen and light genture - subject of als token	25
Input ils buen and list gentuce - subjut of cls tokan	
HEL	

Subject:	
Year, Month, Date, ()	
3) Rueskon Answering -> wit own in Oleviso  who are concil enter oury  output. Oleviso (span of answer from 160)	4
مواسوال مام درمن ما دس ح	
was and jento any	
nutput. Ole use (span of answer from 140)	
(4) sequence labeling	5
X Cle by	
* Cls token can output the representation for the life text	
* ISERT can be used only for representation ( no first husing)	
* BERT can be used only for representation (no fine tunning)	
Problem	
problem voing BERT in mobiles (large model)	10
(Parsing 1)	
/	
ر بران الماري من وراي هاي سر وراي سر وراي هاي سر وراي	
طعه برسادی اسافهای سی واقع های در عمل	
Cool: 1	15
applications: spell cheeling, speech recognition, translation, LM	-
J , LM	
At constituency parsing 1)	
- phrase structure (each with a head)	
Loop Golde in the water of the star con in lad letter con	
	20
مان ما بعام معلق معلی معلق با بلد جامع المرن اما باده ما معلی معلق با بلد جامع المرن اما باده ما معلق می معلق	
to park a surface we need a grammer Contest of	
The Market Market of the Marke	Mari
G	
cc / wan-terminal (N) phrases	
1 Start Symbol (5)	25
SAHEL	***************************************
	-