#### **NLP Homework 3**

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#### The Options

- Develop a Semantic Role Labeling system
  - with various exciting extensions
- Do a top-secret homework, which:
  - o is **innovative** and might lead to a publication
  - a bit more time-intensive (but could be part of your thesis/excellence path/grant)
  - only for very motivated students
- Students who did not pass the Homeworks will have to do a bigger project (SRL + extension)
- You can achieve top marks with all homework types, it is just a matter of what interests you more.

#### Option 1

Semantic Role Labeling

#### Semantic Role Labeling

#### What is SRL?

- 1. Identify and disambiguate predicates
  - a. e.g. "He ate an apple" --> "He <u>ate</u>/eat.01 an apple"
- 2. Identify the arguments of a given predicate
  - a. "He <u>ate</u>/eat.01 an apple" --> "He <u>ate</u>/eat.01 an apple"
- 3. Label the roles of the arguments
  - a. "He ate/eat.01 an apple" --> "He\_A0 ate/eat.01 an apple\_A1"

These steps don't necessarily have to be split up and could also be performed jointly.

#### **CoNLL 2009 Dataset**

- Dataset for multilingual dependency-based SRL
- 6 languages: English, Catalan, Czech, German, Spanish and Chinese
- Verbal and nominal predicates
- Splits:
  - EN: 1334 dev sentences, 39279 training, 2399 test, 425 out-of-domain test,
  - other languages have similar splits, but they do not have an out-of-domain test set

#### What is provided in the Dataset?

- lemmatization
- POS tags
- dependency relations
- predicate and argument annotations

#### **CoNLL 2009 Format**

1 In	in	in	IN	IN	10	10 ADV	LOC		1	AM-DIS				7.0
2 any	any	any	DT	DT _	3	3 NMOD	NMOD		-	Ti.				70 Test
3 event	event	event	NN	NN	1	1 PMOD	PMOD							
4	0	0	0	0 0	10	10 P	Р					es eg	200	11 10011
5 Mr.	mr.	mr.	NNP	NNP	6	6 TITLE	TITLE				200	1908	202	1, 2011
6 Englund	englund	englund	NNP	NNP _	10	10 SBJ	SBJ	15 88	- MESS	A0	14020	103	888	200
7 and	and	and	CC	CC _	6	6 COORD	COORD		1		-		7.0	
8 many	many	many	DT	ງງ	9	9 NMOD	NMOD		.=2			1808	2000	8 276
9 others	others	others	NNS	NNS	7	7 CONJ	CONJ						- 100	1 2 2
10 say	say	say	VBP	VBP _	0	0 ROOT	ROOT	Y	say.01			1000		7. 002-
11 that	that	that	IN	IN _	10	10 OBJ	OBJ			A1		<u></u>		
12 the	the	the	DT	DT _	14	14 NMOD	NMOD					120	<u> </u>	11 14411
13 easy	easy	easy	JJ	ງງ	14	14 NMOD	NMOD				AM-MNR	1508	202	1 221
14 gains	gain	gain	NNS	NNS _	20	20 SBJ	SBJ	Y	gain.01	123	A2	100	888	A1
15 in	in	in	IN	IN _	14	14 NMOD	LOC		1-	1	A1		7.2	
16 narrowing	narrow	narrow	VBG	VBG	15	15 PMOD	PMOD	Y	narrow.01				769	A 2476
17 the	the	the	DT	DT _	19	19 NMOD	NMOD						1000	57 3-52 2
18 trade	trade	trade	NN	NN _	19	19 NMOD	NMOD			1000		1200	A1	7. 1000 m
19 gap	gap	gap	NN	NN	16	16 OBJ	OBJ	Y	gap.01			A1		
20 have	have	have	VBP	VBP _	11	11 SUB	SUB				<u>~</u>	1225		51 10015 8
21 already	already	already	RB	RB _	20	20 TMP	ADV			100	<u></u>	123	200	AM-TMP
22 been	be	be	VBN	VBN _	20	20 VC	VC	12 03	100	200	828	100	500	20
23 made	make	make	VBN	VBN _	22	22 VC	VC	Y	make.01				7.2	
24.					10	10 P	Р		=		200	1800	727	# 200

Word Lemmas POS Dependency Rel. Predicate Arguments and their roles

#### CoNLL 2004 and 2005 Dataset

Dataset for (span-based) Semantic Role Labeling

[A0 He] [AM-MOD would] [AM-NEG n't] [V accept] [A1 anything of value] from [A2 those he was writing about].

Here, the roles for the predicate **accept** (that is, the *roleset* of the predicate) are defined in the PropBank Frames scheme as:

V: verb

A0: acceptor

A1: thing accepted A2: accepted-from

A3: attribute

AM-MOD: modal AM-NEG: negation

Train: 39832 sentences, Dev: 1346, Test: 2837

#### CoNLL 2004/5 Format

The - (A0* \$ - * 1.4 - *	(A0* * *
1.4 *	*
	*
billion *	
robot - *	*
spacecraft - *)	*)
faces face (V*)	*
a - (A1*	*
six-year - *	*
journey - *	*
to *	*
explore explore *	(V*)
Jupiter - *	(A1*
and - *	*
its - *	*
16 *	*
known - *	*
moons - *)	*)
*	*

#### What is PropBank?

- PropBank is originally a corpus (Penn Treebank) that has been annotated with information about predicate-argument relations.
- A verb lemma, for example "develop", can be associated with various **frames**: a frame denotes a specific semantic usage of a lemma.
  - e.g. **develop.01** (come about, arisen), **develop.02** (create, built up)
- These propbank framesets come with various roles:

ARGO	agent	ARG3	starting point, benefactive, attribute
ARG1	patient	ARG4	ending point
ARG2	instrument, benefactive, attribute	ARGM	modifier

Table 1.1: List of arguments in PropBank

Table 1 Subtypes of the ArgM modifier tag.

LOC: location CAU: cause EXT: extent DIS: discourse connectives ADV: general purpose NEG: negation marker MOD: modal verb

TMP: time PNC: purpose MNR: manner DIR: direction

## An Example of two PropBank frames with their roles:

Frameset cut.01 "slice"

Arg0: cutter

Arg1: thing cut

Arg2: medium, source

Arg3: instrument

Ex:  $[A_{Arg0}]$  Longer production runs  $[A_{ArgM-MOD}]$  would  $[A_{Arg1}]$  inefficiencies from adjusting machinery between production cycles  $[A_{Arg1}]$ . (wsj\_0317)

Frameset cut.04 "cut off = slice"

Arg0: cutter

Arg1: thing cut (off)

Arg2: medium, source

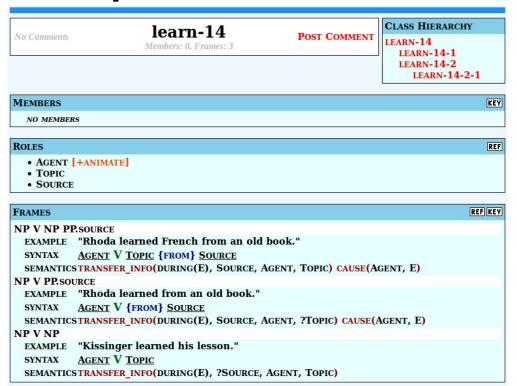
Arg3: instrument

Ex:  $[_{Arg0}$  The seed companies] *cut off*  $[_{Arg1}$  the tassels of each plant]. (wsj\_0209)

#### What is VerbNet?

- an online, hierarchical verb lexicon, containing verb classes
  - Verb Classes are clusters of verbs which share a common argument structure
- while PropBank is part of a corpus, this resource is only a lexicon without a corpus
- the verbs and roles differ from PropBank, but are linked through SemLink
- a verb class contains
  - syntactic frames:
    - Agent V patient
  - o thematic roles:
    - Agent, Theme, Location etc.
    - used across classes
  - selectional restrictions (define the thematic roles):
    - Patient [+comestible & +solid]

#### An example of a VerbNet class



#### What you <u>have</u> to do for Option 1:

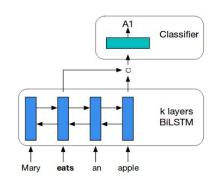


Figure 2: Predicting an argument and its label with an LSTM encoder.

- Develop a Semantic Role Labeling model that takes as input the disambiguated predicate and its sentence and finds and labels its arguments.
  - a. Luca loves\_love.01 Anna --> Luca\_A0, Anna\_A1

A model for example could look like the one to the left.

(Marcheggiani, Diego, Anton Frolov, and Ivan Titov. "A Simple and Accurate Syntax-Agnostic Neural Model for Dependency-based Semantic Role Labeling." *Proceedings of the 21st Conference on Computational Natural Language Learning (CoNLL 2017)*. 2017.)

#### **Extension Idea 1.1**

- Disambiguate the dataset (extra points if you successfully use the system you built for Homework 2)
  - use the semantic information to improve your SRL classifier
  - AND
  - successfully include sense or synset embeddings in your system
     (<a href="http://lcl.uniroma1.it/sensembed/">http://lcl.uniroma1.it/nasari</a>)
- Develop a neural architecture that goes beyond a standard LSTM:
  - use an autoencoder
  - OR use Attention and biLSTMs
  - OR use multi-task learning
  - OR graph-convolutional NN (<a href="https://arxiv.org/pdf/1703.04826.pdf">https://arxiv.org/pdf/1703.04826.pdf</a>)(attention, very hard)

#### Extension Idea 1.2

- Implement a good predicate identification and disambiguation system:
  - Luca loves Anna. --> love.01

#### How:

- o disambiguate the English CoNLL training data (extra points if you use your HW2 system) and create an alignment between PropBank predicate sense and BabelNet synset
- Given this alignment, train a predicate disambiguator on SemCor
- Test your predicate identifier and disambiguator on the English
   CoNLL test data and see whether it improves when you add the
   additional SemCor training data, instead of only training on CoNLL

#### Extension Idea 1.3

- Automatically define the selectional preference of PropBank predicates and their arguments
- Take the CoNLL corpus, disambiguate it and collect all lemmas and their synsets for a given argument of a given predicate:
  - e.g.: eat.01 A1: [apple/bn:00005054n, pizza/bn:00062694n, cucumber/bn:00024294n]
- Given these sets of predicates-arguments-roles:
  - o find ways to generalize:
    - e.g. clustering, centroids, hypernyms, distributional methods
- With these generalizations you have semantically defined the selectional preferences of PropBank
  - In the end you want to have an output similar to the Selectional Preferences of VerbNet:
    - **e.g.** eat.01 A1 : food (you can choose what set of selectional preferences you want to define)

#### Language: up to you!

- you can choose one of the following languages in the CoNLL dataset:
  - English
  - Spanish
  - Chinese
  - Czech

#### **Evaluation**

- You can evaluate your system on the CoNLL 2009 dev split.
- We will give you the unlabeled test split which you have to label and hand in.
- Please mention: F1, Accuracy, Precision, Recall
- Include a confusion matrix
- The following numbers can give you an idea about current state-of-the-art (you do not have to achieve SOTA to get full marks, don't worry!)

Language	Precision	Recall	F1		
ES	81.4	79.3	80.3		
EN	88.7	86.8	87.7		
ZH	83.4	79.1	81.2		
CS	86.6	85.4	86.0		

## What you have to do to get which grade:

- If you develop the mandatory task: maximum grade 27
  - o Report 50%
  - Novelty 20%
  - Performance 20%
  - Code 10%
- If you develop the mandatory task and implement one of the extensions proposed: maximum grade 35
  - Report 50%
  - Novelty 20%
  - Performance 20%
  - Code 10%

#### What you have to submit:

#### A report:

- up to 4 pages (+infinite pages for references, images, tables, graphs, etc.)
- including: a brief introduction to the project problem, a brief state
  of the art, an illustration of the methods/approach/techniques (min.
  1 page), a quantitative (and ideally a small qualitative) evaluation of
  the system, some analysis of the results.
- Your source code via a gitlab project (developer permission to us)
- The output of your system on the test data.
- You will also have to present your work in a presentation
- Submission form <u>HERE</u>

#### Test data output format

- We will provide you with the test data, without the labels
- You have to label the test data in the CoNLL format:

 $\circ$  You have to fill in the all the columns after the predicate column nr.

14

DF	ORM	LEMMA	PLEMMA	POS	PPOS	FEAT	PFEAT	HEAD	PHEAD	DEPREL	PDEPREL	FILLPRED	PRED	APRED1	APRED2	APRED3	APRED4	APRED
ı w		w.			NNP			3	3	NAME	NAME							
2 E		ed			NNP	_	_	3	3	NAME	NAME	_	_	_	_	_	_	
_	_	tyler		_	NNP	_	_	18	18	SBJ	SBJ	_	_	_	_	A1	_	_
4 .	,					_	_	3	3	Р	Р	_	_	_	_		_	
5 3	7	37	37	CD	CD	_	_	6	6	NMOD	NMOD	_	_	_	_	_	_	_
6 ye	ears	year	year	NNS	NNS	_	_	7	7	AMOD	AMOD	_	_	_	_	_	_	_
7 ol		old	,	JJ	JJ	_	_	3	3	NMOD	NMOD	_	_	_	_	_	_	_
8 .							_	3	3	Р	Р		_				_	
9 a		a	a	DT	DT	_	_	12	12	NMOD	NMOD		_		_	_	_	
10 se	enior	senior	senior	JJ	JJ	_		12	12	NMOD	NMOD			A3			_	
11 vi	ice	vice	vice	NN	NN			12	12	NMOD	NMOD			A3				
12 pr	resident	president	president	NN	NN			3	3	APPO	APPO	Y	president.01	A0				
13 at	t	at	at	IN	IN			12	12	LOC	LOC			A2				
14 th	nis	this	this	DT	DT			16	16	NMOD	NMOD	_		_		_		
15 pr	rinting	print	print	VBG	VBG	_	_	16	16	NMOD	NMOD		_		A1			
16 cc	oncern	concern	concern	NN	NN	_	_	13	13	PMOD	PMOD	Υ	concern.02	_	A0	_	_	_
17,		,	,	,	,		_	3	3	Р	Р				_		_	
18 w	as	be	be	VBD	VBD	_	_	0	0	ROOT	ROOT	_	_	_	_	_	_	_
19 el	lected	elect	elect	VBN	VBN	_	_	18	18	VC	VC	Y	elect.01		_			
20 pr	resident	president	president	NN	NN	_	_	19	19	OPRD	OPRD	Υ	president.01			A2	A0	
21 of	f	of	of	IN	IN	_	_	20	20	NMOD	NMOD		_	_	_		A2	
22 its	S	its	its	PRP\$	PRP\$	_	_	24	24	NMOD	NMOD	_	_	_	_			A2
23 te	chnology	technology	technology	NN	NN	_	_	24	24	NMOD	NMOD	_		_	_		_	A1
	roup	group	group	NN	NN	_	_		21	PMOD	PMOD	Y	group.01	_	_	_	_	
25,		,	,	,	,	_	_		20	Р	P	_	_	_	_		_	_
26 a		a		DT	DT	_	-		28	NMOD	NMOD		_		_		_	_
27 ne	ew	new	new	JJ	JJ	_	_	28	28	NMOD	NMOD	_	_	_	_	_	_	_ =
	osition	position	position	NN	NN	_			20	APPO	APPO	Υ	position.02		_	_	_	_
29 .								18	18	Р	Р							

#### Option 2

Innovative Homework 3

#### General Idea

- You will receive a newly created multilingual predicate inventory
- There is no available training data for this inventory (like for example CoNLL for PropBank)
- You will have to create labeled sentences by:
  - leveraging example sentences, the inventory and the selectional preferences/types from the inventory, using CoNLL
- You train a supervised system on the new training data that you created
- You will evaluate on a test set of manually labeled sentences provided by us

#### Why is this interesting to do?

- You will work on a novel approach which might lead to a publication!
- You aim to resolve the knowledge-acquisition bottleneck by:
  - creating new training data!
  - working on a language-independent inventory
  - have access to a knowledge base (BabelNet)

# Introduction to the Predicate Inventory by Andrea Di Fabio

### The predicate inventory and its innovation

- It is completely multilingual/language-independent!
  - All the inventories we previously presented (PropBank, VerbNet, FrameNet etc.) are language-specific
  - This allows: multilingual learning and knowledge interchange between languages
- It is synset-based and therefore connected to BabelNet
  - Does not stand isolated
  - You can make use of this additional knowledge-base

#### **Predicate Inventory**

- multilingual verb clusters, containing:
  - Argument structure of the clusters containing synsets with similar semantic
  - Semantic roles <u>inspired</u> to VerbNet which can occur within the instances of verb synsets in the cluster
  - Selectional preferences to identify the most probable lexical arguments connected through BabelNet IDs to the WordNet hierarchy
  - Internal arguments tagged with BabelNet IDs to explicit the implicit lexical information of the synset

#### Example of the EAT cluster

IPatient (meal, bn:00053993n)

agent (animal, bn:00004222n)

patient (bn:00035182n)

iPatient (food, bn:00035649n)

he devoured three sandwiches

The boy chomped his sandwich

This dog doesn't eat certain kinds of meat

The children crunched the celery sticks

She polished off the remaining potatoes

Swallow the raw fish--it won't kill you!

the old man had no teeth left and mumbled his food

There was so much food at the party that we quickly ? 1

The food was placed on the table and the children pit. 1

We did not eat until 10 P.M. because there were so m 1

She was eating a banana

He jawed his bubble gum

gnaw an old cracker

NO EXAMPLE

chaw tobacco

NO EXAMPLE

NO EXAMPLE

NO EXAMPLE

He filled up on turkey

iPatient (food, bn:00035649n), agent (an) the herd was grazing

iPatient (food, bn:00035649n), agent (an) The animals forage in the woods

	A	В	С	D	E	F G
1	SYNSET	Nome cluster	n° o	Struttura argomentale	iARG	Esempi
1250	un agent mang	i <u>a un</u> patient [in <u>una</u> location]		1) Agent (person   animal),	patient (food), [location ()]; 2) Agent (pe	erson   animal), patient (physical_object), [location ()]
1251	bn:00083503v	EAT, BITE	51		iPatient (food, bn:00035649n)	She stuffed herself at the dinner 1
1252	bn:00083752v	EAT, BITE	51		39199999	NO EXAMPLE 1
1253	bn:00083753v	EAT, BITE	51			Don't bolt your food!
254	bn:00083755v	EAT, BITE	51			The children gobbled down most of the birthday cake 1
1255	bn:00085688v	EAT, BITE	51			Some people can down a pound of meat in the cours 1
1256	bn:00085689v	EAT, BITE	51			Have another bowl of chicken soup!

51

51

51

51

51

51

51

51

51

51

51

51

51

51

51

51

51

51

51 -patient

51 -patient

bn:00086765v EAT, BITE

bn:00087460v EAT, BITE

bn:00087461v EAT, BITE

bn:00087462v EAT, BITE

bn:00084764v EAT, BITE

1262 bn:00084765v EAT. BITE

bn:00084843v EAT, BITE

bn:00084888v EAT, BITE

1265 bn:00086112v EAT. BITE

1266 bn:00088911v EAT. BITE

bn:00089118v EAT, BITE

bn:00084145v EAT, BITE

bn:00084147v EAT, BITE

bn:00084464v **EAT, BITE** bn:00086815v **EAT, BITE** 

bn:00087468v EAT, BITE

bn:00088039v EAT. BITE

bn:00088181v EAT, BITE

bn:00088449v EAT, BITE

1276 bn:00088574v EAT, BITE

1277 bn:00088759v EAT. BITE

#### **Example of the EAT cluster**

Lemmi

rson   animal), patient (physical_object), [location (	)]	
She stuffed herself at the dinner	1 Overeat or eat immodestly; make a pig of oneself	gorge · engorge · glut · <u>ingurgitate</u> · overeat
NO EXAMPLE	1 Swallow hastily	bolt
Don't bolt your food!	1 Eat hastily without proper chewing	gobble · bolt · gorge · devour ·engorge
The children gobbled down most of the birthday cake	1 Eat a large amount of food quickly	bolt down · garbage down · gobble up ·shovel in
<sup>1255</sup> Some people can down a pound of meat in the cours <b></b>	1 Eat immoderately	consume · devour · down · go through
Have another bowl of chicken soup!	Serve oneself to, or consume regularly	ingest take consume have take in
he devoured three sandwiches	1 Eat greedily	devour · guttle · pig · raven
She was eating a banana	1 Take in solid food	eat · consume
We did not eat until 10 P.M. because there were so m	1 Eat a meal; take a meal	eat · consume
This dog doesn't eat certain kinds of meat	1 Take in food; used of animals only	feed · eat
NO EXAMPLE	1 Chafe at the bit, like horses	champ
The boy chomped his sandwich	1 Chew noisily	champ · chomp
chaw tobacco	2 Chew without swallowing	chaw
He jawed his bubble gum	1 Chew (food); to bite and grind with the teeth	chew · masticate · jaw · manducate
The children crunched the celery sticks	1 Chew noisily	crunch · munch
gnaw an old cracker	1 Bite or chew on with the teeth	gnaw
the old man had no teeth left and mumbled his food	1 Grind with the gums; chew without teeth and with gr	gum · mumble
the herd was grazing	1 Feed as in a meadow or pasture	browse · graze · crop · pasture · range
There was so much food at the party that we quickly 🤌	1 Eat lightly, try different dishes	browse · graze
NO EXAMPLE	1 Eat human flesh	cannibalise · cannibalize
1271 The food was placed on the table and the children pit▶	1 Eat heartily	dig in · pitch in
She polished off the remaining potatoes	1 Finish eating all the food on one's plate or on the tab	eat up · finish · polish off
NO EXAMPLE	1 Eat well	fare
He filled up on turkey	1 Eat until one is sated	fill · fill up
The animals forage in the woods	1 Wander and feed	forage
NO EXAMPLE	1 Eat a lot and without restraint	fress · gluttonize · gluttonise
Swallow the raw fishit won't kill you!	1 Pass through the esophagus as part of eating or dri	swallow get down

P) Glossa

Esempi

#### Selectional preferences

Selectional preferences	SYNSET				
absorbent	bn:00000467n				
activity	bn:00001172n				
animal	bn:00004222n				
animate_being	bn:00004222n				
art	<u>bn</u> :00005926n				
artifact	bn:00005956n				
authority	bn:00007300n				
belief	<u>bn</u> :00009716n				
body_process	bn:00001173n				
communication	bn:00021222n				
<sup>2</sup> conduit	<u>bn</u> :00062568n				
container	bn:00022129n				
determination	bn:00026651n				
event	bn:00032021n				
human	bn:00044576n				
instructions	bn:00027356n				
knowledge	<u>bn</u> :00020452n				

#### BabelNet to WordNet hierarchy

#### WordNet Search - 3.1 - WordNet home page - Glossary - Help Word to search for: scientist Search WordNet Display Options: (Select option to change) Change Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations Display options for sense: (gloss) "an example sentence" Noun • <u>S:</u> (n) **scientist** (a person with advanced knowledge of one or more sciences) direct hyponym I full hyponym has instance o direct hypernym I inherited hypernym I sister term • S: (n) person, individual, someone, somebody, mortal, soul (a human being) "there was too much for one person to do" • S: (n) organism, being (a living thing that has (or can develop) the ability to act or function independently) • S: (n) living thing, animate thing (a living (or once living) entity) • S: (n) whole, unit (an assemblage of parts that is regarded as a single entity) "how big is that part compared to the whole?"; "the team is a unit" • S: (n) object, physical object (a tangible and visible entity; an entity that can cast a shadow) "it was full of rackets. balls and other objects" • S: (n) physical entity (an entity that has physical existence) • S: (n) entity (that which is perceived or known or

or nonliving))

inferred to have its own distinct existence (living

## What you have to do for Option 2

#### Option 2.1

- Take SemCor, which is **annotated** with senses
- Extract all sentences which contain a verb sense from the predicate inventory clusters
- Label these sentences with the roles pertaining to the specific verb cluster:
  - HOW? Up to you.. Some ideas:
    - Exploit the selectional preferences
    - Use dependency parsing
    - Exploit the WordNet hierarchy

#### Option 2.2

- Use CoNLL 2009, which is already annotated with predicates and roles, but from a different inventory
- By disambiguating CoNLL 2009, learn a mapping from PropBank predicate to BabelNet predicate (from the clusters) and also learn a mapping from PropBank roles to cluster roles
- Use this mapping to train a system on the CoNLL sentences, labeled with the predicates and arguments from the new predicate inventory

#### **Evaluation**

- You can evaluate your system on gold standard sentences provided by us
- Please mention: F1, Accuracy, Precision, Recall
- Include a confusion matrix

# **Group work**

• if you want you can work in groups of 2 (or alone)

# How we will grade

- up to 30 if you come up with a working system and labeled sentences
  - Report 50%
  - Novelty 20%
  - Performance 20%
  - Code 10%
- 30+ if you have a novel approach and produce good quality labeled sentences.

# NLP Project for non-attending students

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#### **Your Task**

- You implement:
  - A predicate disambiguation system
  - A semantic role labeling system (see extra task 1.2)
- And do 1 of the two other extra tasks (extra task 1.1 and 1.3)
- Refer to the slides in Option 1 to see how you have to do this.

# How we will grade

- If you implement the two systems and one extension you can get up to 30.
  - Report 50%
  - Novelty 20%
  - Performance 20%
  - Code 10%
- If you add something extra on top 30+

# Administrative Stuff

#### **Submission**

- Fill out the GOOGLE FORM:
   https://docs.google.com/forms/d/14jzy1jf3BfMHY82l8T
   3jhWsKx1Ws9AXJTHupl3DA-08/edit
- The google form contains instructions on how to submit your report and your source code.

#### Source code submission

- Register to <u>GitLab.com</u> and create a new project
- Name the project
  - firstname\_lastname\_matricola\_nlp18hw3
- Share the project with (project setting -> members):
  - o <u>pyatkin@di.uniroma1.it</u>
  - o <u>federico.scozzafava@gmail.com</u>
  - o <u>navigli@di.uniroma1.it</u>
- Give us developer permissions
- Upload your code on the repository
- Get the HTTP URL

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  \*\*SSH\*\*

  \*\*HTTPS\*\*

  Create a personal access token on your account to pull or push via HTTPS.

#### **Deadlines**

- Please check
   <a href="http://naviglinlp.blogspot.it/p/natural-language-processi">http://naviglinlp.blogspot.it/p/natural-language-processi</a>
   <a href="http://naviglinlp.blogspot.it/p/natural-language-processi">ng-basic.html</a> for submission and presentation dates.
- For those doing the HW:
  - there are 3 possible submission dates
  - 1 week after you submission you will have to present your HW with a presentation
- For those doing the project:
  - you have 5 possible submission dates
  - 1 week after you submission you will have to present your project with a presentation

# The presentation

- Presentation will be done in classroom with slides.
- Each student will have a maximum time of 10 minutes for presentation (strict) + 10 minutes for questions from the Professor
- You should present a summary of your work on the slides
  - make a presentation out of your report

#### FINAL GRADE

- Your homework 3 will be graded by us and after the presentation, Prof. Navigli will set your final grade
- If you did all the Homeworks your final grade will be the average of the three
- For those doing the big project: this will be your only and final grade, determined by the submission and the presentation.

# **Good Luck!**

