

Morphological disambiguation

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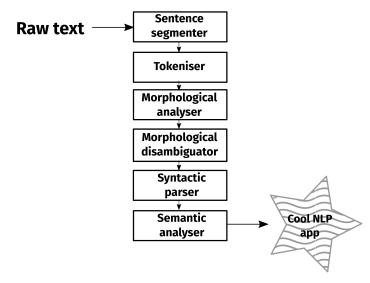
6 марта 2018 г.

Introduction

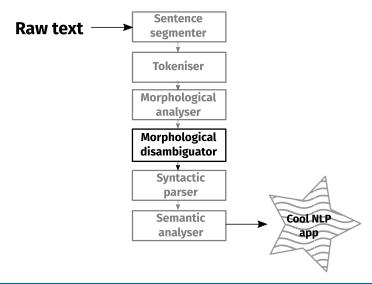


- Introduction, tagsets
- Approaches
 - Rule-based
 - HMM-based
 - Averaged perceptron
- Discussion









Motivating example



при

Motivating example



при:

- npu pr
- пря nfnn sg gen
- *пря* nfnn pl nom
- *пря* nfnn pl acc
- переть vblex impf tv imp p2 sg
- переть vblex impf iv imp p2 sg

Motivating example



Это я знал еще с 46-го года, когда начал писать, а может быть и раньше, – и факт этот не раз поражал меня и ставил меня в недоумение о полезности искусства **при** таком видимом его бессилии.

при:

- npu pr
- *пря* nfnn sg gen
- *пря* nfnn pl nom
- *пря* nfnn pl acc
- переть vblex impf tv imp p2 sg
- переть vblex impf iv imp p2 sg

Applications



Aside from being a stage in the pipeline, what can use POS tagging directly?

- Speech synthesis: How to pronounce a word in context, e.g. conduct
 - NOUN: /'kondukt/, VERB /kon'dukt/
- Disambiguation of meaning:
 - lie NOUN vs. lie VERB
- Corpus linguistics:
 - Find sequences of lexical categories
 - Limit searches for a wordform to a particular category

Terminology



Part-of-speech tagging:

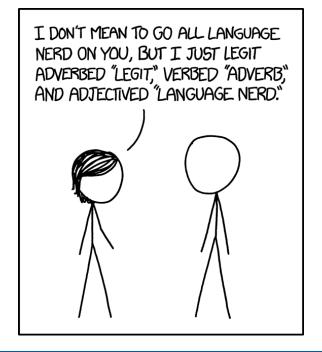
 Traditional term, based on approach(es) for English, finite-set of tags for all combinations of lexical category and morphology.

ln:	This	is	a	test
	This/PRON	is/VERB	a/DET	test/NOUN

Morphological disambiguation:

 More cross-linguistically applicable, conception is of disambiguating after morphological analysis.

ln:	This/DET/PRON	is/VERB	a/DET	test/VERB/NOUN
	This/PRON	is/VERB	a/DET	test/NOUN



What is a tagset?



- Lemmas generalise over sets of inflectional forms
- Part-of-speech tags generalise over sets of lexemes/lemmas that have similar syntactic distribution

Tagset design



Examples:

- Splitting: Participles from adjectives
- Merging: One class for all nominals

Questions:

- Can the ambiguity be resolved?
- Does the distinction help downstream applications?

MORPHOLOGY-BASED

NOMINAL VERBAL UNINFLECTED

> DET=PRON AUX=VERB SCONJ,CCONJ=CONJ

NOUN VERB ADJ ADV PRON DET AUX CCONJ SCONJ NUM ...

SYNTAX-BASED

Example tagsets



This/DT tagset/NNS contains/VBZ 48/CD unique/JJ tags/NNP

Penn Treebank

- 48 tags
- Tags are atomic
- Principles have been applied to other languages (Chinese, Bengali, ...)
- Extensible ?

Example tagsets



Table 2
The Penn Treebank POS tagset.

1. CC	Coordinating conjunction	25. TO	to
2. CD	Cardinal number	26. UH	Interjection
3. DT	Determiner	27. VB	Verb, base form
4. EX	Existential there	28. VBD	Verb, past tense
5. FW	Foreign word	29. VBG	Verb, gerund/present
6. IN	Preposition/subordinating		participle
	conjunction	30. VBN	Verb, past participle
7. JJ	Adjective	31. VBP	Verb, non-3rd ps. sing. present
8. JJR	Adjective, comparative	32. VBZ	Verb, 3rd ps. sing. present
9. JJS	Adjective, superlative	33. WDT	wh-determiner
10. LS	List item marker	34. WP	wh-pronoun
11. MD	Modal	35. WP\$	Possessive wh-pronoun
12. NN	Noun, singular or mass	36. WRB	wh-adverb
13. NNS	Noun, plural	37. #	Pound sign
14. NNP	Proper noun, singular	38. \$	Dollar sign
15. NNPS	Proper noun, plural	39	Sentence-final punctuation
16. PDT	Predeterminer	40. ,	Comma
17. POS	Possessive ending	41. :	Colon, semi-colon
18. PRP	Personal pronoun	42. (Left bracket character
19. PP\$	Possessive pronoun	43.)	Right bracket character
20. RB	Adverb	44. "	Straight double quote
21. RBR	Adverb, comparative	45. '	Left open single quote
22. RBS	Adverb, superlative	46. "	Left open double quote
23. RP	Particle	47.	Right close single quote
24. SYM	Symbol (mathematical or scientific)	48. "	Right close double quote
-1. 51141	of moor (maniematical or belefame)		rugin close double quote



Positional tags

```
<s id="Osl.1.2.3.4">
  <w lemma="Winston" ana="Npmsn">Winston</w>
  <w lemma="se" ana="Px-----y">se</w>
  <w lemma="biti" ana="Vcip3s--n">je</w>
  <w lemma="napotiti" ana="Vmps-sma">napoti1</w>
  <w lemma="proti" ana="Spsd">proti</w>
  <w lemma="stopnica" ana="Ncfpd">stopnicam</w>
  <c>.</c>
  </s>
```

- + Compact
- Hard to read
- No support for derivational morphology

Example tagsets



Mnemonic tags

```
Sápmelaččas [sápmelaš] N Sg Loc
leai [leat] V IV Ind Prt Sg3
dakkár [dakkár] Pron Dem Attr
luondu [luondu] N Sg Nom
, [,] CLB
ahte [ahte] CS
son [son] Pron Pers Sg3 Nom
háliidišgodii [háliidit] V TV Der/goahti Ind Prt Sg3
gottiid [goddi] N Pl Acc
. [.] CLB
```

- + Easily handle derivations
- + Implicit morphological structure
- Number of tags can explode
- Modelling derivation is less language-independent

Example tagsets



Feature/value pairs

```
      1
      Польша
      _ PROPN
      _ Animacy=Inan|Case=Nom|Gender=Fem|Number=Sing
      _

      2
      является
      _ VERB
      _ Aspect=Imp|Number=Sing|Person=3|Tense=Pres
      _

      3
      безъядерной
      _ ADJ
      _ Animacy=Inan|Case=Ins|Gender=Fem|Number=Sing
      _

      4
      страной
      _ NOUN
      _ Animacy=Inan|Case=Ins|Gender=Fem|Number=Sing
      _

      5
      _ PUNCT
```

- + Easy to read
- No support for derivational morphology
- No implicit morphological structure
- Takes up a lot of space

Limitations



қорагыркиплыткогъат

- қора-гырки-плыткогъат
- reindeer-catch-finish-3PL
- They finished catching reindeer.
- Does VERB really capture what this is?
- We can represent morphology easily, maybe even derivation
- But what about the incorporation?

Scale of the problem



- UD corpora
- Percentage of tokens and types that receive more than one analysis
- Underestimation, e.g. Turkish için:
 - + for POST
 - + inside.GFN
 - inside.2SG.NOM
 - drink.IMP.2PL

Language	Tokens	/type	/token
Turkish	58k	4.29	17.44
Finnish	201k	3.46	18.09
Kurmanji	10k	9.35	36.72
Basque	121k	11.47	38.47
Russian	1.1M	13.50	40.94
Erzya	2k	9.73	41.37
Norwegian	301k	8.28	43.78
Czech	1.5M	18.09	47.17
English	254k	14.20	52.34
German	292k	20.17	56.52
Portuguese	227k	13.19	64.51
Catalan	531k	8.31	66.49
Hebrew	161k	15.56	71.62
Hindi	351k	36.28	86.84

12/52

Types of ambiguity



Type	all tokens	ambig. tokens
Intraparadigm.	59.0%	90.9%
Incongruent	27.7%	42.7%
Congruent	1.2%	1.8%

- Intraparadigmatic:
 - 'тела (SG.GEN) vs. тел'а (PL.NOM)
- Morphosyntactically incongruent:
 - до'рога (NOUN) vs. дорог'а (ADJ)
- Morphosyntactically congruent:
 - 'замок (SG.NOM) vs. зам'ок (SG.NOM)

Approaches



- Rule-based
- HMM-based
- Averaged perceptron

Rule-based

Constraint Grammar



- Developed by Fred Karlsson¹ in the late 1980s
- Does not aim at producing a full "parse tree"
- Describes what is ungrammatical, not what is grammatical
- Linguists formalise "constraints" which describe language **impossibilities**
 - e.g. "No noun can be in prepositional case without a preposition which governs the prepositional case."
- No "encapsulation", all parts of the analysis (surface form \rightarrow semantics) are always available
- Input is all possible analyses, output is only possible analyses

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¹The same Fred Karlsson that wrote "Finsk grammatik".

Formalism



Input:

```
«Польша»"

"Польша"np top f sg nom

«является»"

"являться"v impf iv pres p3 sg
«безъядерный"adj f an sg gen
"безъядерный"adj f an sg dat
"безъядерный"adj f an sg prp
"безъядерный"adj f an sg ins
«страной»"

"страна"n f nn sg ins
«.>"
"sent
```

Operators:

- select: Discard all readings except the reading matching a condition
- remove: Discard a single reading matching a condition

Context conditions:

- (-1 pres) → previous token has the tag PRES
- (1C ins) → following token *only* has the tag INS
- (NOT -1* pr) \rightarrow no token to the left has the tag PR



Input:

```
«Польша>"
"Польша"np top f sg nom
«яляется>"
«яляется>"
«безъядерной>"
"безъядерный"adj f an sg gen
"безъядерный"adj f an sg prp
"безъядерный"adj f an sg prp
"безъядерный"adj f an sg prp
"безъядерный"adj f an sg ins
«страной>"
"страна"n f nn sg ins
«.>"
"."sent
```



Input:

```
«Польша»"
 "Польша"np top f sq nom
 "являться"v impf iv pres p3 sg
 «безъядерной»"
 "безъядерный"adj f an sg gen
 "безъядерный"adj f an sg prp
 "безъядерный"adj f an sg prp
 "безъядерный"adj f an sg ins
 "сезъядерный"adj f an sg ins
 "страной»"
 "страна"n f nn sg ins
 «.>"
 "."sent
```

1 REMOVE prp IF (not -1* pr)



Input:

```
«Польша>"
    "Польша"пр top f sq nom
«является>"
    "являться" v impf iv pres p3 sq
«безъядерной>"
    "безъядерный"adj f an sq dat
    "безъядерный"adj f an sg ins
«страной>"
    "crpaha"n f nn sg ins
```

- 1 REMOVE prp if (not -1* pr)
- 2 REMOVE gen IF (-1 pres) (OC adj) (not 1 gen)

" "sent



Input:

```
«Польша»"

"Польша"np top f sg nom

«являтся»"

"являться"v impf iv pres p3 sg
«безъядерный"adj f an sg gen
"безъядерный"adj f an sg dat
"безъядерный"adj f an sg prp
"безъядерный"adj f an sg ins
«страной»"

"страна"n f nn sg ins
«.>"
""sent
```

- 1 REMOVE prp IF (not -1* pr)
- 2 REMOVE gen IF (-1 pres) (OC adj) (not 1 gen)

Exercise: Can we safely remove the dative reading?

Less basic example



« Для соседних с Руандой государств руандийские события апреля – июля 1994 года вылились в огромное число прибывших беженцев . »

Standard trigram taggers



Для

A=pl,gen,plen соседних PR

c Руандой S,f,inan=sq,ins S,n,inan=pl,gen государств

A=pl,acc,inan,plen руандийские события S,n,inan=pl,acc S,m,inan=sq,gen

апреля

S,m,inan=sg,gen июля 1994 NUM=ciph S,m,inan=sg,gen года

V,pf,intr,med=pl,praet,indic выпипись

PR

огромное A=n,sq,acc,inan,plen чиспо

S,n,inan=sg,acc

прибывших V,pf,intr,act=partcp,pl,gen,praet,plen

беженцев S,m,anim=pl,gen

Standard trigram taggers



Для

соседних A=pl,gen,plen S,f,inan=sg,ins Руандой

S,n,inan=pl,gen государств руандийские A=pl,acc,inan,plen события S,n,inan=pl,acc

S,m,inan=sq,gen апреля

июпя S.m.inan=sq.gen 1994 NUM=ciph S,m,inan=sq,gen года

V,pf,intr,med=pl,praet,indic вылились

A=n,sq,acc,inan,plen огромное

S,n,inan=sg,acc число

V,pf,intr,act=partcp,pl,gen,praet,plen прибывших

беженцев S,m,anim=pl,gen

2/19 = 89.5% accuracy



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
   "соседний"adj mfn an pl prp
   "соседний"adj mfn aa pl acc
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
   "руандийский" adj mfn nn pl acc
«события>"
   "событие"n nt nn sq gen
   "событие"n nt nn pl nom
   "событие"n nt nn pl асс
«апреля>"
   "апрель"n m nn sg gen
«->"
  "-"guio
«июпя>"
   "июль"n m nn sq qen
```

```
«1994>"
                                        rule: -
   "1994"num
«гола>"
   "год"n m nn sq gen
«выпились>"
   "вылиться" v perf iv past mfn pl
"×8>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg nom
   "огромный" adj nt an sg acc
«число>"
   "число"n nt nn sg acc
   "число"n nt nn sg nom
«прибывших>"
   "прибыть" v perf iv pp actv mfn an pl acc
   "прибыть" v perf iv pp actv mfn an pl prp
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "беженец"n m aa pl acc
   "."sent
```



```
«Лля>"
«соседних>"
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
   "руандийский" adj mfn nn pl acc
«события>"
   "событие"n nt nn sq gen
   "событие"n nt nn pl nom
   "событие"n nt nn pl асс
«апреля>"
   "апрель"n m nn sg gen
«->"
  "-"guio
«июпя>"
   "июль"n m nn sq qen
```

```
«1994>"
                                        rule: 1
   "1994"num
«гола>"
   "год"n m nn sq gen
«выпились>"
   "вылиться" v perf iv past mfn pl
"×8>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg nom
   "огромный" adj nt an sg acc
«число>"
   "число"n nt nn sg acc
   "число"n nt nn sg nom
«прибывших>"
   "прибыть" v perf iv pp actv mfn an pl acc
   "прибыть" v perf iv pp actv mfn an pl prp
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "беженец"n m aa pl acc
   "."sent
```



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
   "руандийский" adj mfn nn pl acc
«события>"
   "событие"n nt nn sq gen
   "событие"n nt nn pl nom
   "событие"n nt nn pl асс
«апреля>"
   "апрель"n m nn sg gen
«->"
  "-"guio
«июля>"
   "июль"n m nn sq qen
```

```
«1994>"
                                        rule: 2
   "1994"num
«гола>"
   "год"n m nn sq gen
«выпились>"
   "вылиться" v perf iv past mfn pl
"×8>"
«огромное>"
«число>"
«прибывших>"
   "прибыть" v perf iv pp actv mfn an pl acc
   "прибыть" v perf iv pp actv mfn an pl prp
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "беженец"n m aa pl acc
   "."sent
```



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
   "руандийский" adj mfn nn pl acc
«события>"
   "событие"n nt nn sq gen
   "событие"n nt nn pl nom
«апреля>"
   "апрель"n m nn sg gen
«->"
  "-"guio
«июпя>"
   "июль"n m nn sq qen
```

```
«1994>"
                                         rule: 3
   "1994"num
«гола>"
   "год"n m nn sq gen
«вылились>"
«B>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg acc
«число>"
   "число"n nt nn sg acc
«прибывших>"
   "прибыть" v perf iv pp actv mfn an pl prp
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "."sent
```



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
«события>"
«апреля>"
   "апрель"n m nn sq gen
«->"
  "-"guio
«июпя>"
   "июль"n m nn sq qen
```

```
«1994>"
                                         rule: 4
   "1994"num
«гола>"
   "год"n m nn sq gen
«вылились>"
«B>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg acc
«число>"
   "число"n nt nn sg acc
«прибывших>"
   "прибыть" v perf iv pp actv mfn an pl prp
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "."sent
```

Input: Morphological analysis



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
«события>"
   "событие"n nt nn pl nom
«апреля>"
   "апрель"n m nn sq gen
«->"
  "-"guio
«июля>"
   "июль"n m nn sq qen
```

```
«1994>"
                                          rule: 5
   "1994"num
«гола>"
   "год"n m nn sq gen
«выпились>"
   "вылиться" v perf iv past mfn pl
"×8>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg acc
«число>"
«прибывших>"
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "."sent
```

Input: Morphological analysis



```
«Лля>"
   "для"рг
«соседних>"
   "соседний"adj mfn an pl gen
«c>"
   "c"pr
«Руандой>"
   "Руанда"np top f sq ins
«государств>"
   "государство"n nt nn pl gen
«руандийские>"
   "руандийский"adj mfn an pl nom
«события>"
   "событие"n nt nn pl nom
«апреля>"
   "апрель"n m nn sq gen
«->"
  "-"guio
«июля>"
   "июль"n m nn sq qen
```

```
«1994>"
                                          rule: 5
   "1994"num
«гола>"
   "год"n m nn sq gen
«выпились>"
   "вылиться" v perf iv past mfn pl
"×8>"
   "B"pr
«огромное>"
   "огромный" adj nt an sg acc
«число>"
   "число"n nt nn sg acc
«прибывших>"
   "прибыть" v perf iv pp actv mfn aa pl gen
«беженцев>"
   "беженец"n m aa pl gen
   "."sent
```

Proposed rule (I)



1 Immediately after "для" remove any reading which is in a case other than genitive.

Exceptions:

• None?

```
LIST Gen = gen ;
SET NGDAIP = nom OR gen OR dat OR acc OR ins OR prp ;
REMOVE NGDAIP - Gen IF (-1C ("для")) ;
```

Proposed rule (II)



2 After "B" remove any reading which is in nominative

Exceptions:

Joining an organisation?

```
LIST Nom = nom ; REMOVE Nom IF (-1C ("B")) ;
```



3 In a sentence with a single intransitive finite verb, remove any reading in accusative which is not immediately governed by a preposition

Exceptions:

- There is a trans. part. form having an acc. arg.
- Some adverbial forms... *Мы проехали километр*.

```
LIST IV = iv;
LIST TV = tv;
LIST Acc = acc;
LIST Pr = pr;
REMOVE Acc IF (0 Acc LINK NOT -1* Pr) ((-1* IV) OR (1* IV)) (0 Acc LINK NOT -1* TV);
```



- 4 Select nominative if there is an intransitive verb which agrees with a nominative noun in the sentence for number (and/or gender)
 - and is preceded by an adj. that can only be nom.
 - and there is no other nom. head in the sentence.

Exceptions:

Appositions, titles, parentheticals? Non-canonical agreement?

```
LIST Head = np n prn;

SET NUM = (sg) OR (pl);

SELECT Nom + \$NUM IF (-1C A + Nom) (NOT -1* Head + Nom)

(NOT 1* Head + Nom) ((-1* V + \$NUM) OR (1* V + \$NUM));
```

Proposed rule (V)



5 If there is a prepositional case reading, remove it if you see a noun which is only in a case other than prepositional without any preceding transitive participle form

Exceptions:

• ..

```
LIST Prp = prp ;

LIST N = n ;

REMOVE Prp IF (-1* N + NGDAIP - Prp) ;
```

Output: Morphologically-disambiguated text



```
«Для>"
   "для"рг
                                              «1994>"
«соседних>"
                                                 "1994"num
   "соседний"adj mfn an pl gen
                                              «гола>"
                                                 "год"n m nn sq gen
                                              «вылились>"
«c>"
                                                 "вылиться" v perf iv past mfn pl
   "c"pr
                                             «B>"
«Руандой>"
                                                 "B"pr
   "Pyaндa"np top f sq ins
                                              «огромное>"
«государств>"
   "государство"n nt nn pl gen
                                                 "огромный" adj nt an sq acc
«руандийские>"
                                              «число>"
   "руандийский"adi mfn an pl nom
                                                 "число"n nt nn sq acc
«события>"
                                              «прибывших>"
   "событие"n nt nn pl nom
                                                 "прибыть" v perf iv pp actv mfn aa pl gen
                                              «беженцев>"
«апреля>"
                                                 "беженец"n m aa pl gen
   "апрель"n m nn sq qen
«->"
   "-"quio
                                             «.>"
«июля>"
                                                 "."sent
   "июль"n m nn sq qen
```

Examples



Languages with constraint grammars:

- Finnish
- North Sámi, Lule Sámi, South Sámi
- Norwegian (Nynorsk, Bokmål)
- Faroese
- Udmurt
- Breton

HMM-based

Predict hidden states from observed events

- hidden states = sequence of part of speech tags
- observed events = ambiguity classes or surface forms

$$M = (A, B, \pi)$$

- *A* = transition probabilities
- B = emission probabilities
- π = initial probabilities

Visible events



The visible events can be either:

- Surface forms: In many traditional HMM-based taggers, the visible events are surface forms
- Ambiguity classes: Generalisation over types of ambiguity e.g. NOUN/VERB, DET/PRON

Example:

```
Surface forms: This is a test .

Ambig. classes: PRON/DET VERB/AUX DET VERB/NOUN PUNCT
```

A tagged corpus



Analysed:

Vino/NOUN/VERB a/ADP la/DET/PRON playa/NOUN ./PUNCT
Voy/VERB a/ADP la/DET/PRON casa/NOUN/VERB ./PUNCT
Bebe/VERB vino/NOUN/VERB en/ADP casa/NOUN/VERB ./PUNCT
La/DET/PRON casa/NOUN/VERB es/VERB grande/ADJ ./PUNCT
Es/VERB una/DET/PRON/VERB ciudad/NOUN grande/ADJ ./PUNCT

Tagged:

Vino/VERB a/ADP la/DET playa/NOUN ./PUNCT
Voy/VERB a/ADP la/DET casa/NOUN ./PUNCT
Bebe/VERB vino/NOUN en/ADP casa/NOUN ./PUNCT
La/DET casa/NOUN es/VERB grande/ADJ ./PUNCT
Es/VERB una/DET ciudad/NOUN grande/ADJ ./PUNCT



We calculate the transition probabilities, A from a matrix of transition counts:

Second tag									
VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT			
0	1	1	0	2	1	0			
1	0	0	0	1	1	3			
0	4	0	0	0	0	0			
0	0	0	0	0	0	0			
0	1	2	0	0	0	0			
0	0	0	0	0	0	2			
3	0	1	0	0	0	0			
	0 1 0 0 0	0 1 1 0 0 4 0 0 0 1 0 0	VERB NOUN DET 0 1 1 1 0 0 0 4 0 0 0 0 0 1 2 0 0 0	VERB NOUN DET PRON 0 1 1 0 1 0 0 0 0 4 0 0 0 0 0 0 0 1 2 0 0 0 0 0	VERB NOUN DET PRON ADP 0 1 1 0 2 1 0 0 1 0 1 0 4 0 0 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0	VERB NOUN DET PRON ADP ADJ 0 1 1 0 2 1 1 0 0 0 1 1 0 4 0 0 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0 0 0			



We calculate the transition probabilities, A from a matrix of transition counts:

Second tag									
VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT			
0	1	1	0	2	1	0			
1	0	0	0	1	1	3			
0	4	0	0	0	0	0			
0	0	0	0	0	0	0			
0	1	2	0	0	0	0			
0	0	0	0	0	0	2			
3	0	1	0	0	0	0			
	0 1 0 0 0	0 1 1 0 0 4 0 0 0 1 0 0	VERB NOUN DET 0 1 1 1 0 0 0 4 0 0 0 0 0 1 2 0 0 0	VERB NOUN DET PRON 0 1 1 0 1 0 0 0 0 4 0 0 0 0 0 0 0 1 2 0 0 0 0 0	VERB NOUN DET PRON ADP 0 1 1 0 2 1 0 0 1 0 1 0 4 0 0 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0	VERB NOUN DET PRON ADP ADJ 0 1 1 0 2 1 1 0 0 0 1 1 0 4 0 0 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0 0 0			



We calculate the transition probabilities, A from a matrix of transition counts:

		S	econd ta	g		
VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT
0	1	1	0	2	1	0
1	0	0	0	1	1	3
0	4	0	0	0	0	0
0	0	0	0	0	0	0
0	1	2	0	0	0	0
0	0	0	0	0	0	2
3	0	1	0	0	0	0
	0 1 0 0 0	0 1 1 0 0 4 0 0 0 1 0 0	VERB NOUN DET 0 1 1 1 0 0 0 4 0 0 0 0 0 1 2 0 0 0	VERB NOUN DET PRON 0 1 1 0 1 0 0 0 0 4 0 0 0 0 0 0 0 1 2 0 0 0 0 0	0 1 1 0 2 1 0 0 0 1 0 4 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0	VERB NOUN DET PRON ADP ADJ 0 1 1 0 2 1 1 0 0 0 1 1 0 4 0 0 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0 0 0 0 0 0



We calculate the transition probabilities, A from a matrix of transition counts:

			S	econd ta	g		
	VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT
VERB	0	0.2	0.2	0	0.4	0.2	0
NOUN	0.16	0	0	0	0.16	0.16	0.5
DET	0	1	0	0	0	0	0
PRON	0	0	0	0	0	0	0
ADP	0	0.3	0.6	0	0	0	0
ADJ	0	0	0	0	0	0	1
PUNCT [†]	0.75	0	0.25	0	0	0	0

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[†] This row represents the initial probabilities, π of the model.



We calculate the transition probabilities, A from a matrix of transition counts:

			S	econd ta	g		
	VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT
VERB	0	0.2	0.2	0	0.4	0.2	0
NOUN	0.16	0	0	0	0.16	0.16	0.5
DET	0	1	0	0	0	0	0
PRON	0	0	0	0	0	0	0
ADP	0	0.3	0.6	0	0	0	0
ADJ	0	0	0	0	0	0	1
PUNCT [†]	0.75	0	0.25	0	0	0	0

† This row represents the initial probabilities, π of the model.

33 / 52

A tagged corpus



Analysed:

Vino/NOUN/VERB a/ADP la/DET/PRON playa/NOUN ./PUNCT Voy/VERB a/ADP la/DET/PRON casa/NOUN/VERB ./PUNCT Bebe/VERB vino/NOUN/VERB en/ADP casa/NOUN/VERB ./PUNCT La/DET/PRON casa/NOUN/VERB es/VERB grande/ADJ ./PUNCT Es/VERB una/DET/PRON/VERB ciudad/NOUN grande/ADJ ./PUNCT

Tagged:

Vino/VERB a/ADP la/DET playa/NOUN ./PUNCT Voy/VERB a/ADP la/DET casa/NOUN ./PUNCT Bebe/VERB vino/NOUN en/ADP casa/NOUN ./PUNCT La/DET casa/NOUN es/VERB grande/ADJ ./PUNCT Es/VERB una/DET ciudad/NOUN grande/ADJ ./PUNCT

Emission probabilities



The probability of seeing an ambiguity class given a tag, B.

	VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT
ADJ	0	0	0	0	0	0	2
DET/PRON	0	0	3	0	0	0	0
DET/PRON/VERB	0	0	1	0	0	0	0
NOUN	0	2	0	0	0	0	0
NOUN/VERB	1	4	0	0	0	0	0
ADP	0	0	0	0	3	0	0
PUNCT	0	0	0	0	0	0	5
VERB	4	0	0	0	0	0	0
Total:	5	6	4	0	3	2	5

Высшая школа экономики

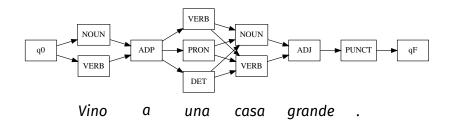
Emission probabilities



The probability of seeing an ambiguity class given a tag, B.

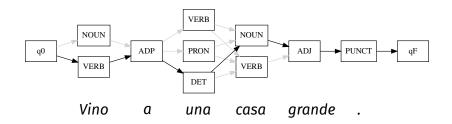
	VERB	NOUN	DET	PRON	ADP	ADJ	PUNCT
ADJ	0	0	0	0	0	1.0	0
DET/PRON	0	0	0.75	0	0	0	0
DET/PRON/VERB	0	0	0.25	0	0	0	0
NOUN	0	0.33	0	0	0	0	0
NOUN/VERB	0.2	0.67	0	0	0	0	0
ADP	0	0	0	0	1.0	0	0
PUNCT	0	0	0	0	0	0	1.0
VERB	8.0	0	0	0	0	0	0





- Dynamic programming algorithm
- Find the most likely sequence of hidden states given observed sequence
- e.g. Find POS tag sequence given words or ambiguity classes





- Dynamic programming algorithm
- Find the most likely sequence of hidden states given observed sequence
- e.g. Find POS tag sequence given words or ambiguity classes



Vino a una casa grande.

 q_F VERB NOUN DET PRON ADP ADJ PUNCT VERB/NOUN ADP DET/PRON/VERB NOUN/VERB ADJ PUNCT Vino grande а casa una



			\rightarrow				
q_F							
VERB	0.15, q ₀						
NOUN	0.0, q ₀						П
DET							П
PRON							П
ADP							П
ADJ							П
PUNCT							П
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	П
	Vino	а	una	casa	grande		П

- = P(VERB,PUNCT) * P(VERB, VERB/NOUN) = 0.75 * 0.2 = 0.15
- = P(NOUN, PUNCT) * P(NOUN, VERB/NOUN) = 0.0 * 0.67 = 0.0



			\rightarrow				
q_F							
VERB	0.15, q ₀						
NOUN	0.0, q ₀						
DET							
PRON							
ADP		0.06, VERB					
ADJ							
PUNCT							
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	arande		П

- = P(ADP, VERB) * P(ADP, ADP) * P(PATH) = 0.4 * 1.0 * 0.15 = 0.06
- = P(ADP,NOUN) * P(ADP, ADP) * P(PATH) = 0.16 * 1.0 * 0.0 = 0



_ (

q_F							
VERB	0.15, q ₀		0.0, ADP				
NOUN	0.0, q ₀						
DET			0.009, ADP				
PRON			0.0, ADP				П
ADP		0.06, VERB					П
ADJ							
PUNCT							П
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	П
	Vino	а	una	casa	grande		

- P(DET,ADP) * P(DET, DET/PRON/VERB) * P(PATH) = 0.6 * 0.25 * 0.06 = 0.009
- = P(PRON,ADP) * P(PRON, DET/PRON/VERB) * P(PATH) = 0.0 * 0.0 * 0.06 = 0.0
- = P(VERB,ADP) * P(VERB, DET/PRON/VERB) * P(PATH) = 0.0 * 0.0 * 0.06 = 0.0



q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ							
PUNCT							
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

- = P(NOUN, VERB) * P(NOUN, NOUN/VERB) * P(PATH) = 0.2 * 0.67 * 0.009 = 0.001
- = P(VERB, VERB) * P(VERB, NOUN/VERB) * P(PATH) = 0.0 * 0.2 * 0.009 = 0.0
- P(NOUN,DET) * P(NOUN, NOUN/VERB) * P(PATH) = 1.0 * 0.67 * 0.009 = 0.006
- = P(VERB,DET) * P(VERB, NOUN/VERB) * P(PATH) = 0.0 * 0.2 * 0.009 = 0.0
- P(NOUN,PRON) * P(NOUN, NOUN/VERB) * P(PATH) = 0.0 * 0.67 * 0.009 = 0.0
- = P(VERB.PRON) * P(VERB. NOUN/VERB) * P(PATH) = 0.0 * 0.67 * 0.009 = 0.0



__

q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				П
ADP		0.06, VERB					П
ADJ					0.001, NOUN		П
PUNCT							П
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	П
	Vino	а	una	casa	grande		

- = P(ADJ,NOUN) * P(ADJ, ADJ) * P(PATH) = 0.16 * 1.0 * 0.006 = 0.00096
- = P(ADJ,VERB) * P(ADJ, ADJ) * P(PATH) = 0.2 * 1.0 * 0.0 = 0.0



_

q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			Г
DET			0.009, ADP				Г
PRON			0.0, ADP				
ADP		0.06, VERB					Г
ADJ					0.001, NOUN		Г
PUNCT						0.001, ADJ	Г
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

 ⁼ P(PUNCT,ADJ) * P(PUNCT, PUNCT) * P(PATH) = 1.0 * 1.0 * 0.001 = 0.001



_

q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ					0.001, NOUN		
PUNCT						0.001, ADJ	
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

PUNCT



q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ					0.001, NOUN		Г
PUNCT						0.001, ADJ	Г
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	Г
	Vino	а	una	casa	grande		

ADJ PUNCT



_

q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	$0.0, q_0$			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ					0.001, NOUN		
PUNCT						0.001, ADJ	
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

NOUN ADJ PUNCT



_

q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				П
PRON			0.0, ADP				П
ADP		0.06, VERB					П
ADJ					0.001, NOUN		
PUNCT						0.001, ADJ	
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

DET NOUN ADJ PUNCT



q _F							П
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			П
NOUN	0.0, q ₀			0.006, DET			П
DET			0.009, ADP				П
PRON			0.0, ADP				П
ADP		0.06, VERB					П
ADJ					0.001, NOUN		П
PUNCT						0.001, ADJ	П
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

ADP DET NOUN ADJ PUNCT



q_F							
VERB	0.15, q ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ					0.001, NOUN		
PUNCT						0.001, ADJ	
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

VERB ADP DET NOUN ADJ PUNCT

Decoding



q_F							
VERB	0.15, <i>q</i> ₀		0.0, ADP	0.0, DET*			
NOUN	0.0, q ₀			0.006, DET			
DET			0.009, ADP				
PRON			0.0, ADP				
ADP		0.06, VERB					
ADJ					0.001, NOUN		
PUNCT						0.001, ADJ	
	VERB/NOUN	ADP	DET/PRON/VERB	NOUN/VERB	ADJ	PUNCT	
	Vino	а	una	casa	grande		

VERB ADP DET NOUN ADJ PUNCT

Implementation



Extensions



- Trigrams
- Backoff
- Unknown words
- Capitalisation

Averaged perceptron

Training



```
1. def train(self, nr iter, examples):
        ''' Update the feature weights according to guesses '''
2.
3.
        for i in range(nr iter):
            for features, true tag in examples:
                guess = self.predict(features)
 5.
6.
                if guess != true tag:
7.
                    for f in features:
8.
                        self.weights[f][true tag] += 1
                        self.weights[f][guess] -= 1
9.
            random.shuffle(examples)
10.
11.
```

Features



Vino a una casa grande .
$$i-2$$
 $i-1$ i $i+1$ $i+2$ $i+3$

```
Trigram suffix
                     una
       Unigram prefix
                     u
i-1
                     ADP
       Tag
i-2
                     VERB
       Tag
       Word
                     una
i-1, i Tag, Word
                     ADP + una
i-1 Word
                     а
i-1 Trigram suffix
                     а
i-2 Word
                     Vino
i+1 Word
                     casa
i+1 Trigram suffix
                     asa
i+2
       Word
                     grande
```

Averaging



Dictionary



Prediction



```
1. def predict(self, features):
 2.
        '''Dot-product the features and current weights and return the best class.'''
        scores = defaultdict(float)
 3.
        for feat in features:
 4.
 5.
            if feat not in self.weights:
                continue
 6.
 7.
            weights = self.weights[feat]
            for clas, weight in weights.items():
 8.
 9.
                scores[clas] += weight
10.
        # Do a secondary alphabetic sort, for stability
11.
        return max(self.classes, kev=lambda clas: (scores[clas], clas))
12.
```

Discussion

Comparison of approaches



+

CG Start from scratch

Model distribution

Easy to incorporate feats

Tagset not learnt

Hard to incorporate feats

No *n*-best

All techniques can reach 97% token accuracy.

Great ... but 57% full-sentence accuracy.

Perceptron



Table 4. Frequency of different POS tagging error types.

Class	Frequency
1. Lexicon gap	4.5%
2. Unknown word	4.5%
3. Could plausibly get right	16.0%
4. Difficult linguistics	19.5%
5. Underspecified/unclear	12.0%
6. Inconsistent/no standard	28.0%
7. Gold standard wrong	15.5%

Chris Manning (2011) "Part-of-Speech Tagging from 97% to 100%: Is It Time for Some Linguistics?"

How much time and effort?



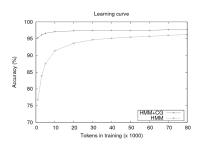
Annotation time vs. rule-writing time

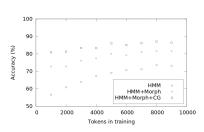
- Hand annotation: 8000–10000 tokens/month
 - 50–100k tokens = 6–12 months
- Rule-based:
 - Morphological analyser: 3–6 months
 - Constraint grammar: 3–6 months

Approaching a new language, depends on what you like doing more.

Tagger combination







- Voting systems
- Combine systems that make complementary errors

Some taggers



Russian:

- pymorphy2
- mystem3

Trainable:

- HunPos (HMM, OCaml)
- UDPipe/MorphoDiTa (Av. Perceptron, C++)

Высшая школа экономики

- MarMot (CRF, Java)
- NLTK (various, Python)

Practicals



```
https://ftyers.github.io/2017-КЛ_МКЛ/practicals/disambiguation.html
```

- Tagger comparison:
 - Compare three taggers on a language/domain of your choice
- Constraint grammar:
 - Select a small text (one paragraph) in a language of your choice
 - Analyse it with a morphological analyser
 - Resolve as much of the ambiguity as you can
- Perceptron tagger:
 - Download https://github.com/ftyers/ conllu-perceptron-tagger
 - Run it on a language from Universal Dependencies
 - Improve it so that you get better performance
 - Add support for morphological features
 - Try tweaking other features