Arabic Language: Nature and Challenges

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Outline

- Introduction
- Lexicons and Corpus Linguistics
- Morphology
- Syntactic Parsing
- Tokenization
- Multiword Expressions
- Statistical Parsing
- Which is better?
- Spelling Checking and Correction
- Integration with Applications





Living language are just ... living



Living language are just ... living

They grow



Living language are just ... living

Leaves fall off



 Living language are just ... living New leaves appear ... constantly changing They may grow old and die They maybe reborn New languages may appear

• Language reveals everything about us ...



- Language tell everything about us
- How rich or poor



- Language tell everything about us
- How well-educated

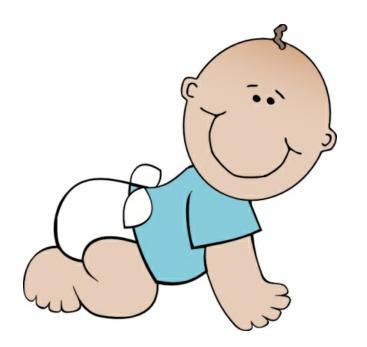




- Language tell everything about us
- what kind of work we do



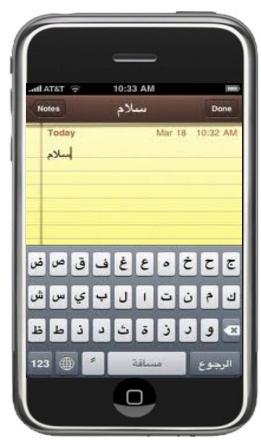
- Language tell everything about us
- our feelings and our sentiments





Language is the key to business expansion:
 Translation and localization

World translation business in 2011 = \$30 billion





And a repository of knowledge and information



Lexicons and Corpus Linguistics



مبادئ علم صناعة المعاجم

Definition of a dictionary

A description of the vocabulary (حصيلة لغوية) used by members of a speech community (مجتمع يتحدث نفس اللغة). A dictionary deals with:

• Conventions عرفی not idiosyncrasies

• norms سائد not rarities

م العام Probable واقع not possible م

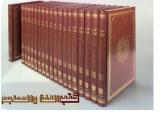
Lexical evidence

- Subjective evidence
 - Introspection الاستبطان
 - informant-testing استشارة أصحاب المعرفة
- Objective evidence
 - A corpus (ذخيرة النصوص) provides typifications (تصنيف للأنماط) of the language
 - A typical lexical entry means it is both "frequent" متكرر or "recurrent" دوري and "well-dispersed" موزع ومتفرق
 in a corpus.
 - A typical lexical entry belongs to the stable "core" of the language.



Corpora and Dictionaries

- Brown Corpus, 1 million words, 1960s,
 - → Citations for *American Heritage Dictionary*
- Birmingham corpus, 20 million words, 1980s
 - → Cobuild English Dictionary.
- British National Corpus (BNC), 100 million words, 1990s set the standard (balance, encoding)
- The Oxford English Corpus, one billion words, 2000s
 - → Oxford English Dictionary
- Longman Corpus Network, 330 million word
 - → Longman Dictionaries



Dictionaries before Corpora

- Citation banks مراجع اقتباسية استشهادية
 - A citation is a short extract providing evidence for a word usage or meaning in authentic use.
- Disadvantages
 - labour-intensive
 - instances of usage are authentic, but there is a big subjective element in their selection.
 - People tend to notice what is remarkable and ignore what is typical
 - bias towards the novel or idiosyncratic usages

- Characteristics of a reliable corpus (مواصفات ذخيرة)
 - The corpus does not favour high class language
 - The Corpus should be large and diverse
 - The corpus should be either synchronic or diachronic
 - The corpus should be well-balanced using "stratified sampling" (أخذ عينات بشكل نسبى)
 - The corpus should avoid skewing (الانحراف أو التحيز)

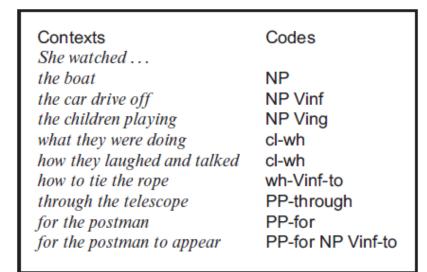
Lexical Profiling

- Word POSv, n, adj, adv, conj, det, interj, prep, pron
- Valency Information subcat frames, other obligatory or optional syntactic constructions
- Collocations
 commit a crime, sky blue, lame duck

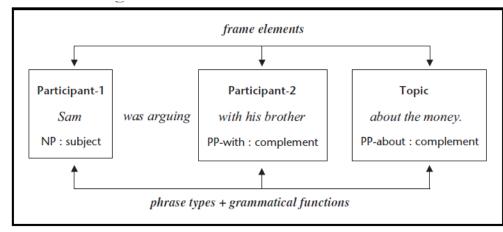
ظلام دامس، ارتكب جريمة، فعل فعلة

 Colligational preferences was acquitted, trials (difficult experiences)

لم يأبه

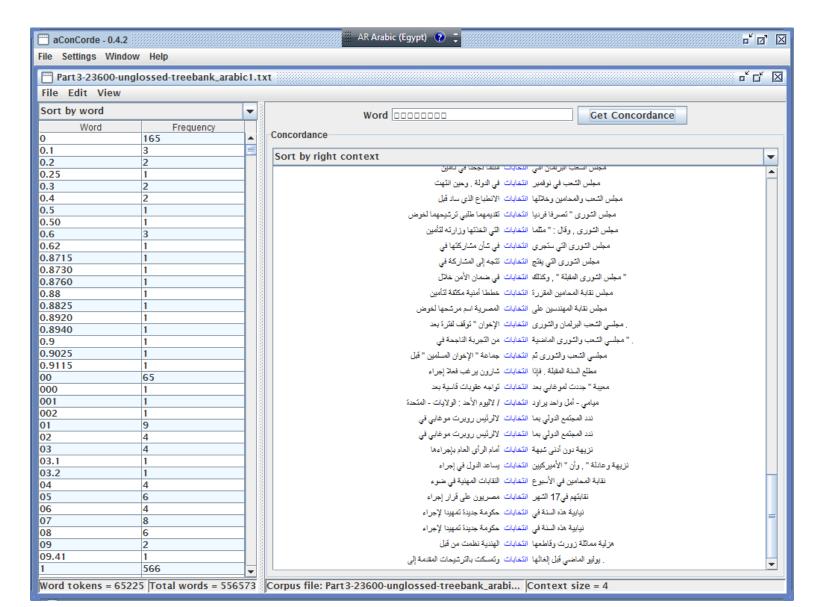


Some constructions for the verb watch



- Lexical Profiling Software
- Concordancers
- Word Sketch (Sketch Engine) Adam Kilgarriff

Concordancer



Sketch Engine

object_of	<u>264</u>	2.7	a_modifier	251	2.0
strike	<u>61</u>	43.38	hard	23	25.99
drive	26 27.56		real	20	23.43
get	27	16.38 best		14	19.31
seal	5	14.82	good	19	18.01
make	26 27 5 26 8	13.6	bad	8	15.31
find	8	7.81	better	8	14.4
modifies	221	0.9	n_modifier	115	1.1
basement	22	38.62	plea	26	40.62
hunter	22	37.23	wage	6	16.8
price	54	33.65	credit	6	14.68
bookshop	11	26.73	sale	5	10.47

Part of the Word Sketch for the noun bargain

1973 مستويات العربية المعاصرة للسعيد محمد بدوي

- فصحى التراث
- فصحى العصر
- عامية المثقفين
- عامية المتنورين
- عامية الأميين

Modern Standard Arabic vs. Classical Arabic vs. Colloquial Arabic

Modern Standard Arabic

 The language of modern writing, prepared speeches and the language of the news

Classical Arabic

- The language of Arabia before Islam and after Islam until the Medieval Times
- Present religious teaching, poetry and scholarly literature.

Colloquial Arabic

 Variety of Arabic spoken regionally and which differs from one country or area to another. They are to a certain extent mutually intelligible.

Code Shifting – Code Switching – Diglossia – multi-layered diglossia

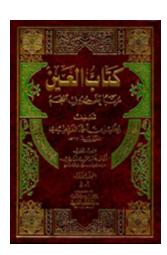
Modern Standard Arabic vs. Classical Arabic vs. Colloquial Arabic

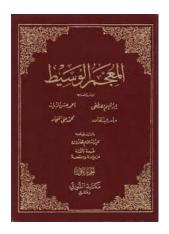
- Modern Standard Arabic
 - Tendency for simplification
 - Some CA structures to die out
 - Structures marginal in CA started to have more salience
 - no strict abidance by case ending rules
 - A subset of the full range of structures, inflections and derivations available in CA
 - MSA conforms to the general rules of CA
 - How "big" or how "small" the difference (on morphological, lexical or syntactic levels) need more research and investigation

Kitab al-'Ain by al-Khalil bin Ahmed al-Farahidi (died 789)
 (refinement/expansion/organizational Improvement)

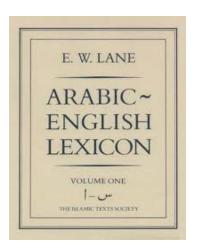


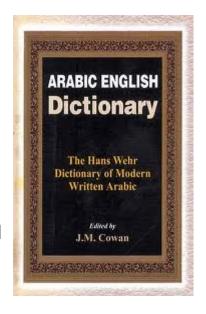
- Tahzib al-Lughah by Abu Mansour al-Azhari (died 980)
- *al-Muheet* by al-Sahib bin 'Abbad (died 995)
- Lisan al-'Arab by ibn Manzour (died 1311)
- al-Qamous al-Muheet by al-Fairouzabadi (died 1414)
- Taj al-Arous by Muhammad Murtada al-Zabidi (died 1791)
- Muheet al-Muheet (1869) by Butrus al-Bustani
- al-Mu'jam al-Waseet (1960)





- Bilingual Dictionaries
 - Edward William Lane's Arabic-English Lexicon (1876) indebted to Taj al-Arous by al-Zabidi
 - Hans Wehr's Dictionary of Modern Written Arabic (1961)
 - Size: 45,000 entries
 - Aim: Using scientific descriptive principles to describe present-day vocabulary through wide reading in literature of every kind
 - Application
 - Selection of works by high flying poets and literary critics such as Taha Husain, Taufiq al-Hakim, Mahmoud Taimur, al-Manfalauti, Jubran Khalil Jubran
 - Use of secondary sources (dictionaries) for expansion
 - Inclusion of rarities and classicisms that no longer formed a part of the living lexicon





Bilingual Dictionaries

- Landau and Brill (1959) A Word Count of Modern Arabic
 Prose
 - A word count based on 270,000 words based on: 136,000 from the news (Moshe Brill, 1940) and 136,000 from 60 contemporary books on:
 - fiction, literary criticism, history, biography, political science, religion, social studies, economics, travels and historical novels
 - 6,000 words in the news
 - 11,000 words in literature
 - 12,400 words in the combined list (does not include proper nouns)

- Bilingual Dictionaries
 - Van Mol's (2000) Arabic-Dutch learner's dictionary
 - COBUILD-style, Corpus-based (3 million words)
 - Manually constructed
 - Covers the whole range of the actual vocabulary in the corpus with 17,000 entries compared to 45,000 entries in Hans Wehr
 - 5% of frequent new words not found in Hans Wehr

- Buckwalter Arabic Morphological Analyzer (2002)
 - Size: 40,222 lemmas (including 2,034 proper nouns)
 - Includes many obsolete lexical items (But how many?)

#	Meaning	Classical	Google	MSA Word	Google
		Word			
1	sully	qalʻat قلعط	8	laṭṭaḫa لطخ	29,600
2	caulk	qalfaṭ قلفط	9	afsada' أفسد	205,000
3	wear	istakadda' استکد	4	anhaka' أنهك	37,100
4	fickle	ġamlağ غملج	7	mutaqallib متقلب	189,000
5	erosion	i'tikāl' ائتكال	7	taʾākul تأكل	1,700,000

Google score for Classical vs. MSA entries



Largest corpus of modern Arabic to date Arabic gigaword 1,200,000,000

= 16,000 large books

= 800 meters of bookshelves

Burj Khalifah is 830m

Avr reader reads 200 wpm With 60% comprehension.

You will need 11 years 24/7 to read the Gigaword corpus



Buckwalter obsolete words: 8,400 obsolete words

رمل: هَيكان وَعْس مِيعاس عِثْيَر

صحراء: فَيْفاء فَدْفَد قَواء مَوْماة مَتْلَف سَبْسَب

سرج: حِداجَة مَخْلُوفَة

حِمْل: طَعِينَة حِدْج طَعُون وِقْر

الجام: فدام كَعَم كِعام أَرْنَبَة شَكَم غِمامة

راكب: حُدّاء

جمل: هَجِينَة

رداء: دفِّيَّة بشْتَة

حذاء: مِبْذَل بَشْمَق زَرْبُول زَرْبُون



Not in Dictionaries: about 10,000 need to be added

سياسة: أمننة شرعنة أفروعربية إثني إقصائي تسييس محاصصة جبهوي جمهوعسكرية العصبوية شخصنة أمركة

تكنولوجيا:

شيفرة

رقمنة، أتمتة، مَكْنَنة
فيس بوك، تويتر، تغريدة
هاتف \ جوال \ تليفون محمول
لاب توب
الهواتف الذكية
حوسبة
بريد إلكتروني
آي فون، دي في دي، سي دي
سبام، فيروس
ملتي ميديا
كمبيوتر لوحي، شاشة لمسية



اقتصاد: خصخصة ريعي يورو بورصة تعويم داو_جونز تضخم أسهم قيمة_دفترية مليار ترليون تجارة_إلكترونية

Morphological Lexicon AraComLex

- How our lexical database will be different from Buckwalter's.
 We include
 - only entries attested in a corpus
 - subcategorization frames
 - +/-human semantic information for nouns
 - Information on allowing passive and imperative inflection for verbs
 - Information on diptotes
 - detailed information about derived nouns/adjectives (active or passive participle or a verbal noun, masdar)
 - multi-word expressions
 - classification of proper nouns: person, place, organization, etc.
 - Frequency information
 - Citation in real examples

Corpus-based Lexicon

Lexicons as a truthful representation of the language as evidenced in a corpus

The Arabic Gigaword corpus

AraComLex

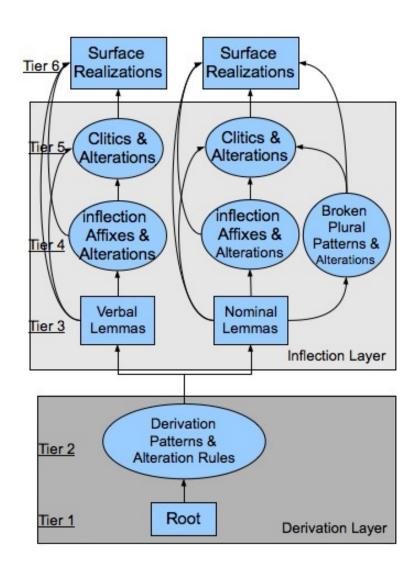
Our morphological analyser – based on a lexical database automatically derived from the Arabic Gigaword Corpus



Arabic Morphology

Arabic Morphotactics

Root	درس			
	m drs			
Template	$R_1aR_2aR_3a$	$R_1aR_2R_2aR_3a$	$R_1\bar{a}R_2iR_3$	$muR_1aR_2R_2iR_3$
POS	V	V	N	N
Stem	darasa	darrasa	dāris	mudarris
	'study'	'teach'	'student'	'teacher'

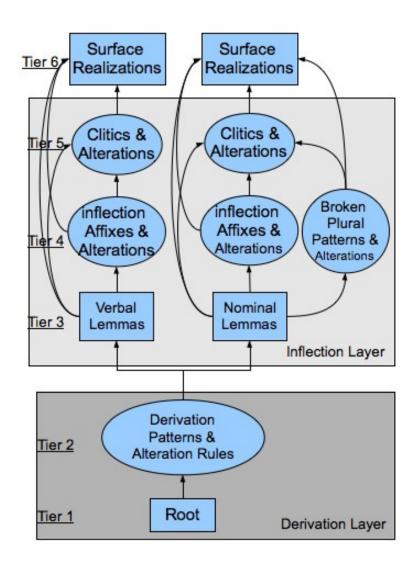


Arabic Morphology

- Design Approach: Three approaches
- Root-based Morphology Xerox Arabic FTM
- Stem-based morphology Buckwalter

\$kr \$akar PV thank;give thanks \$kr \$okurIV thank;give thanks

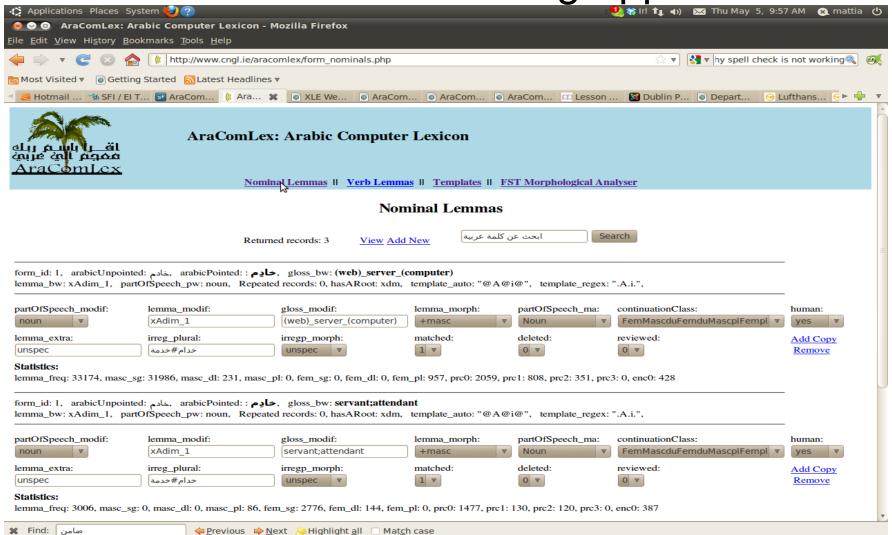
3. Lemma-based morphology





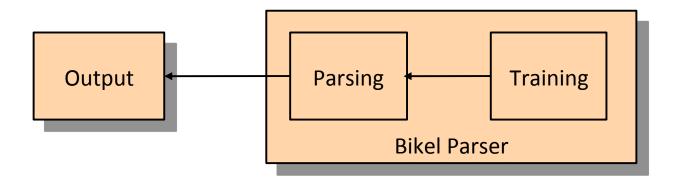
Morphological Lexicon - AraComLex

AraComLex Lexicon Writing Application

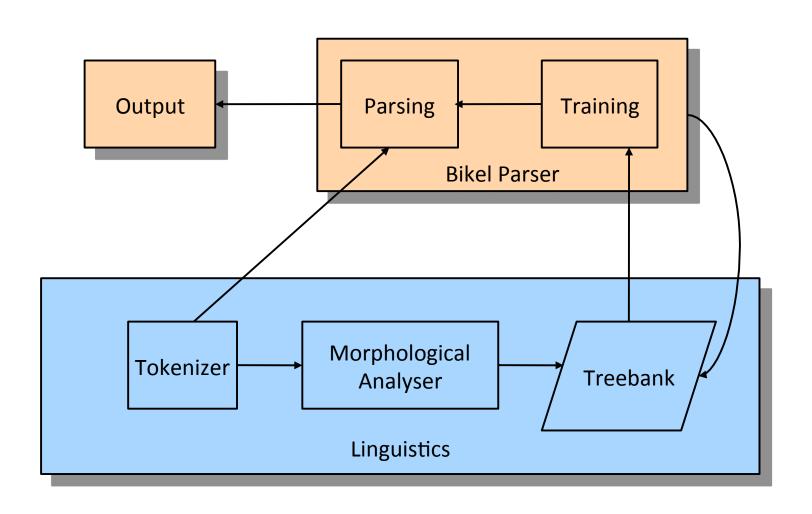


Syntactic Parsing

Algorithms and Data Structure



Algorithms and Data Structure



Why Linguistics

- Linguistic Data is a naughty blackbox:
 - You get non-deterministic answers
 - You can get wrong answers
 - For the same question, you can get a set of inconsistent answers
- We need to make the algorithms suite the data structure, and we also need to make sure that the data is structured properly.

Sentence

```
ساعدت الهبئة الفلسطينيين
sāʿadat al-haiʾatu al-filisṭīniyyīn/ al-filisṭīniyyain
helped the-agency the-Palestinian.pl/ the-Palestinian.dual
'The agency helped the Palestinians/ the two Palestinians.'
```

Tokenization

helped@the@agency@the@Palestinians

Morphological analysis

```
ساعدت
       pers+ساعدverb+past+active+
         -verb+past+active+ساعد3pers+sg+fem
helped
         +verb+past+active+ساعد2pers+sg+fem
         +verb+past+active+ساعد2pers+sg+masc
 ائــ
         +defArt-1
the
         fem+sg+هیئة+fem+sg
agency
 masc+dual+accgen+فلسطيني +ad
masc+pl+accgen+فلسطینی Palestinians-ad
         +noun+human-فلسطینیmasc+dual+accgen
         +noun+human-فلسطيني masc+pl+accgen
```

Lexicon (Lexical properties/subcategorization frames)

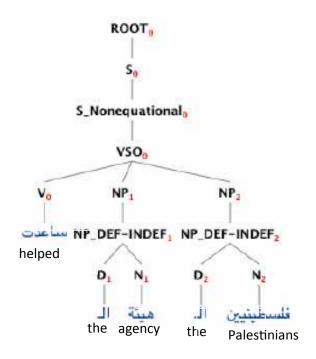
```
"V XLE (^ GLOSS)=help "This verb has three different subcat frames"
 helped
          { (^ PRED)='% stem<(^ SUBJ)(^ OBJ)(^ COMP)>'
            (^ COMP COMP-FORM)=c أن (^ COMP COMP-TYPE)=c verbal
           على (^ PRED)='% stem<(^ SUBJ)(^ OBJ)(^ OBL)>' (^ OBL OBJ PCASE)=c
           | (^ PRED)='% stem<(^ SUBJ)(^ OBJ)>'}.
           N XLE (^ GLOSS)=agency (^ PRED)='%stem' (^ PERS)=3
agency
            { (^ NUM) (^ NUM) ~= sg | (^ NUM) = sg } "the default number is singular".
  N XLE (^ GLOSS)=Palestinian (^ PRED)='% stem' (^ PERS)=3 فلسطيني
Palestinian { (^ NUM) (^ NUM) ~= sg | (^ NUM) = sg } "the default number is singular";
           ADJ XLE (* PRED)='%stem' (* GLOSS) = 'Palestinian'
            { (^ ATYPE)=c predicative | (^ ATYPE)= attributive}.
```

Grammar Rules: PS-rules and functional equations

```
MT ARABIC RULES (1.0)
S_Nonequational --> "There are three word orders permitted in Arabic: VSO, SVO and VOS"
        { VS0
         SVO
         I VOS}.
VSO --> V: ^=! @DefSTense (^ VTYPE)~= copular (^ COMP-TYPE)=verbal
          {(^ SUBJ PRED)=c 'pro' (^ SUBJ NUM) = (^ AGR NUM)
           | (^ SUBJ PRED)~= 'pro' (^ AGR NUM)=sg)}
          (^ AGR GEND)=(^ SUBJ GEND) (^ AGR PERS)=(^ SUBJ PERS);
        {NP: (^SUBJ)=! (! FIRST-CONJ)=+
                  (! CASE)=nom (! PRON-TYPE) ~=pers
        e: (^ SUBJ PRED)='pro' "ProDrop"
                  (^ AGR PERS) = (! PERS) (^ AGR NUM) = (! NUM) (^ AGR GEND) = (! GEND) }
```

(NP: (^OBJ)=! (! CASE)=acc).

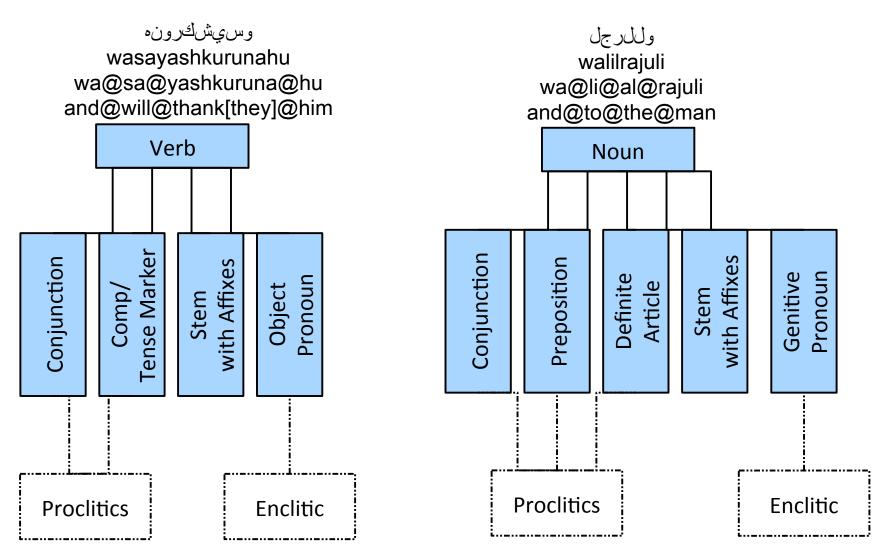
Output: c-structures and f-structures



```
PRED
         '<[فاسطيني:2] ,[فيئة:1]>ساعد'
TNS-ASP 4 TENSE past, MOOD indicative
         PERS 3, NUM sg, GEND fem
AGR
           'فاسطینی' PRED
                     DET 11 DET-TYPE def
                      a1 dual
           NUM
OBJ
           NTYPE a NSYN common
           PERS 3, HUMAN +,
           GLOSS Palestinian, GEND masc,
           DEF +, CASE acc
           PRED
           SPEC
SUBJ
           NTYPE 5 NSYN common
           PERS 3, NUM sg, HUMAN -,
           GLOSS agency, GEND fem,
           FIRST-CONJ +, DEF +, CASE nom
STMT-TYPE decl, PASSIVE -, GLOSS help,
COMP-TYPE verbal
```



Tokenization in XLE



Tokenization in XLE

Deterministic Tokenizer

```
وللرجل (walirrağul: and to the man)
(walirrağul: and to the man) هال ال المرجل (walirrağul: and @to@the@man@
```

Non-Deterministic Tokenizer

```
وللرجل (walirrağul: and to the man)
@المرجل (wali@al@rağul@and@to@the@man@
وهال@الرجل
وهالرجل
والرجل
والرجل
```

- English parser
 - Input sentence:

The President led his country in reform.

- Formatted sentence:

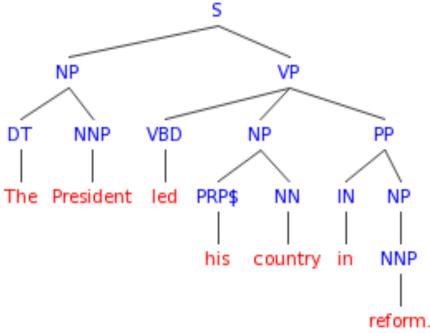
(The President led his country in reform.)

(VBZ has) (RB n't)
(NNP Chicago) (POS 's)

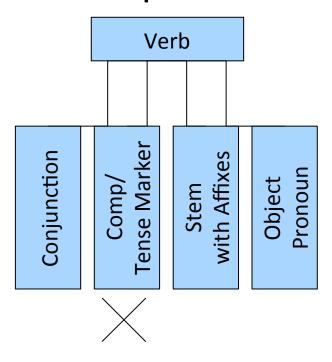
- English parser
 - Output:

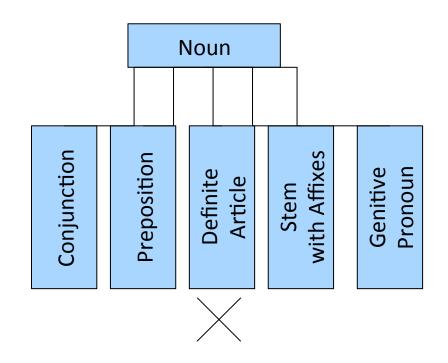
(S (NP (DT The) (NNP President)) (VP (VBD led) (NP (PRP\$ his) (NN country)) (PP (IN in) (NP (NNP reform.)))))

– Tree



Arabic parser





- Arabic parser
 - Input sentence:

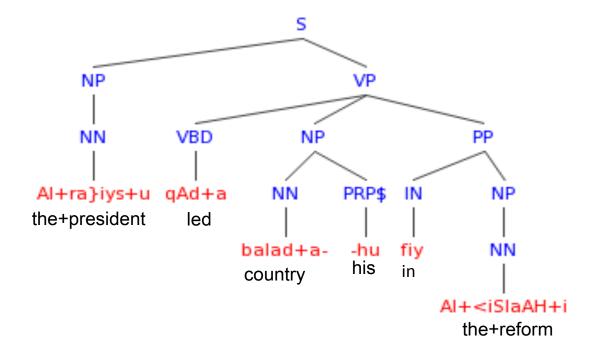
The President let his country in reform.

- Formatted sentence:
 - Alra}iysu qAda baladahu fiy Al<iSlaAHi
 - Alra}iysu qAda balada- -hu fiy Al<iSlaAHi
 - Al+ra}iys+u qAd+a balad+a- -hu fiy Al+<iSlaAH+i
 - (Al+ra}iys+u qAd+a balad+a- -hu fiy Al+<iSlaAH+i)

- Arabic parser
 - Output:

(S (NP (NN Al+ra}iys+u)) (VP (VBD qAd+a) (NP (NN balad+a-) (PRP\$ -hu)) (PP (IN fiy) (NP (NN Al+<iSlaAH+i)))))

– Tree



Multiword Expressions

Multiword Expressions in XLE

Three types of MWEs

 Fixed Expressions: Lexically, morphologically and syntactically rigid. A word with spaces.

New York

United Nations

- Semi-Fixed Expressions: Lexically, or morphologically flexible
 - Sweep somebody under the rug/carpet
 - Transitional period(s)

مدينة ملاهي \ مدينة ترفيهية
 قاعدة عسكرية

- Syntactically-flexible Expressions
 - •to let the cat out of the bag
 - The cat was let out of the bag.

Multiword Expressions

- MWEs are important
 - High frequency in natural language (30-40%)
 - Important for MT, literal translation is usually wrong
 - When taken as a block, they relieve the parser from the burden of processing component words
 - We collected 34,658 MWEs in addition to 45,202
 Named Entities

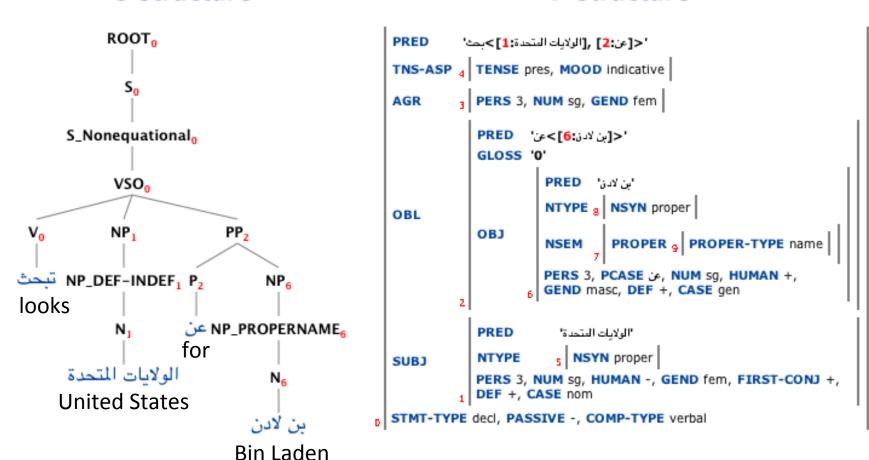
Multiword Expressions in XLE

تبحث الولايات المتحدة عن بن لادن

The United States looks for Bin Laden.

C-structure

F-structure



Multiword Expressions in Bikel

Compositional, yet detectable in the English treebank

```
(NP (DT the) (NNP United) (NNP Kingdom)
(NP (NNP New) (NNP York) )
(NP (DT the) (NNP Middle) (NNP East) )
(NP (NNP Saudi) (NNP Arabia)
(NP (NNP Las) (NNP Vegas)
(NP (NNP Los) (NNP Angeles)
(CONJP (IN in) (NN addition) (TO to)
```

Multiword Expressions in Bikel

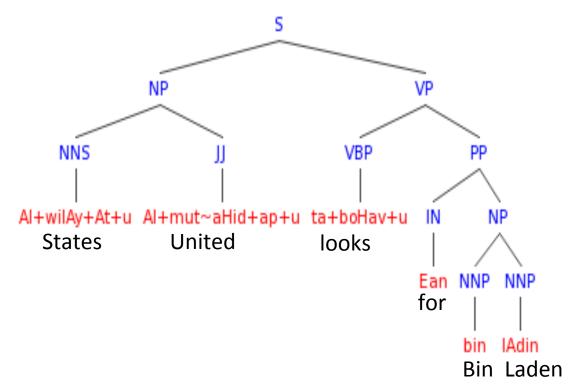
 Compositional, undetectable, sometimes inconsistent, in Arabic treebank

```
ان جلي س Los Angeles
(NP (NOUN PROP luws)
      (NOUN PROP > anojiliys)
الولايات المتحدة United States
(NP (DET+NOUN+NSUFF FEM PL+CASE DEF NOM AI+wilAy+At+u)
       (DET+ADJ+NSUFF FEM SG+CASE DEF NOM AI+mut~aHid+ap+u))
الشرق الأوسط The Middle East
(NP (DET+NOUN+CASE DEF GEN AI+$aroq+i)
      (DET+ADJ+CASE_DEF_GEN Al+>awosaT+i))
in addition to إضافة إلى
(CONJP (NOUN+NSUFF FEM SG+CASE INDEF ACC <iDAf+ap+F) (PREP <ilaY))
(NP-ADV (NP (NOUN+NSUFF_FEM_SG+CASE_INDEF_ACC -<iDAf+ap+F)) (PP (PREP <ilaY)
(NP (NP (NOUN PROP EarafAt))
```

Multiword Expressions in Bikel

Example

The United States looks for Bin Laden. الول اي ات المتحدة تبحث عن بن لادن (S (NP (NNS Al+wilAy+At+u) (JJ Al+mut~aHid+ap+u)) (VP (VBP ta+boHav+u) (PP (IN Ean) (NP (NNP bin) (NNP IAdin)))))



Statistical Parsing

Coverage of the statistical parser on sentence
 40 words:

Arabic: 75.4%

– Chinese: 81%

- English: 87.4%

(Bikel, 2004)

Arabic is "far below" the required standard.
 (Kulick et al., 2006)

- Why Arabic performs poorly? (Kulick et al. 2006)
 - The ATB tag set is very large and dynamic, this is why they are mapped into 20 PTB tags. The tagset reduction is extreme and important information is lost.

Verb

- » IV3FS+IV+IVSUFF MOOD:I
- » IV3MS+IV+IVSUFF_MOOD:J
- » PV+PVSUFF_SUBJ:3MS
- » IVSUFF_DO:3MP

- Noun

- » NOUN+CASE DEF ACC
- » DET+NOUN+NSUFF_FEM_PL+CASE_DEF_GEN
- » NOUN+NSUFF_FEM_SG+CASE_DEF_GEN

- Why Arabic performs poorly? (Kulick et al. 2006)
 - Average sentence length in Arabic is 32 compared to
 23 in English
 - Significant number of POS tag inconsistencies, for example lys is tagged as NEG_PART and PV
 - 5% of VP in Arabic have non-verbal heads
 - Base Noun Phrases (NPB) are 30% in English compared to 12% in Arabic.
 - Construct states in Arabic roughly correspond to possession constructions in English

- Why Arabic performs poorly? (Kulick et al. 2006)
 - Arabic has a much greater variance in sentence structure than English.

Sentence Type	Arabic %	English %
VSO	62	0
SVO	17	90
No VP	19	11
Subjectless VP	2	0

Major revision of Arabic treebank guidelines 08



- Common wisdom: statistical parsers are:
 - Shallow: They do not mark syntactic and semantic dependencies needed for meaning-sensitive applications

(Kaplan et al., 2004)

XLE: "We parse the web."

```
\Theta \Theta \Theta
                  X 1 valid F-structure for ROOT:1016
 kill | prev | next | Commands Views 🔲 a 🔟 c 📕 n 🔟 s
     lock | F-structure | 0::*
    "We parse the web."
                   'parse([22:we], [70:web]>'
          PRED
          SUBJ 342 NTYPE NSYN pronoun
                346 case nom, Human +, Num pl, PERS 1, PRON-TYPE pers
                        'web'
                   PRED
                   CHECK [LEX-SOURCE countnoun-lex]
                746 NTYPE NSEM [COMMON count]
                858
                         NSYN common
                930
       159
                70|SPEC
                941 CASE obl, NUM sq, PERS 3
      1149 CHECK [SUBCAT-FRAME V-SUBJ-0BJ]
      1158 TNS-ASP MOOD indicative, PERF -_, PROG -_, TENSE pres
      1016 CLAUSE-TYPE decl, PASSIVE -, VTYPE main
```

- Common wisdom is not entirely true.
- DCU: "We can also parse the web."

```
S
                                      subj : pred : pro
                                              pron_form : we
NP-SBJ
                  VP
                                      pred : can
                                      modal : +
                       VΡ
                                      adjunct : 1 : pred : also
 PRP
           ADVP
                                      xcomp : subj : pred : pro
                                                      pron_form : we
 We
                   VΒ
             RB
       can
                                               pred : parse
                                               obj : spec : det : pred : the
            also
                 parse
                                                     pred : web
                                                     num : sq
                        the web
                                                     pers : 3
```

Summary

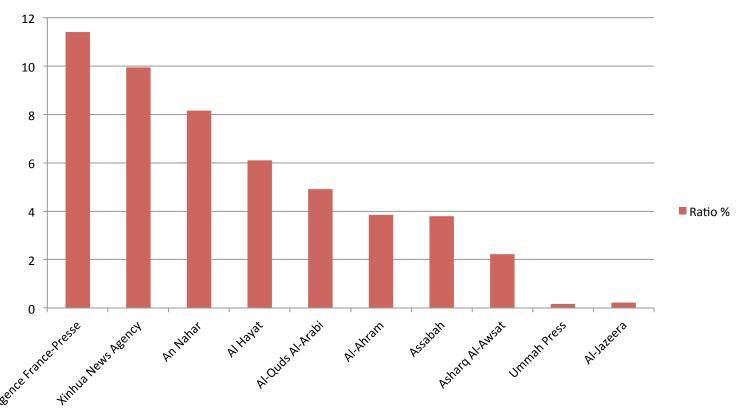
- Handcrafted grammars are built on assumptions and intuitions. They depend on how good these assumptions are.
- Handcrafted grammar can be improved by:
 - Effectively managing the development project
 - Making use of statistical facts (treebanks, and TIGERSearch)

- Statistical grammars are built on facts. They depend on how true these facts are.
- Statistical grammar can be improved by:
 - Improving the quality and size of treebanks.

- Statistical grammars are more efficient because:
 - there is a clear separation between the algorithm and the data structure
 - there is a clear division of labour, the linguists fight their battle, and the engineers fight their own battle

How frequent is spelling errors in news web sites?





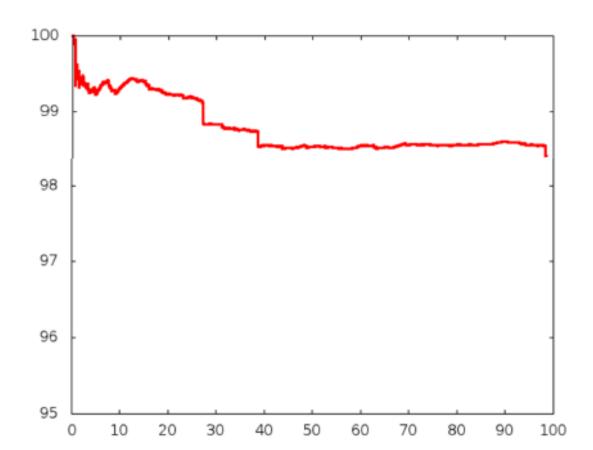
Creating a wordlist

English: 708,125, French: 338,989, Polish: 3,024,852

	No. of	MS	MS
	Words	Accepted	Rejected
		_	
AraComLex ⁴	12,951,042	8,783,858	4,167,186
Arabic-Spell for	938,977	673,875	265,103
Aspell (using			
Buckwalter)			
1 billion-word	2,662,780	1,202,481	1460,447
corpus			
(Gigaword ⁵ and			
Al-Jazeera)			
Ayaspell for	292,464	230,506	61,958
Hunspell			
Total*	15,147,199	9,306,138	5,841,061

Spelling error detection

- 1. Matching against a word list 2. Character-based LM



Automatic correction of spelling errors

Spell Checker	First order ranking	
MS Spell Checker	80.54%	
Hunspell using Ayaspell	45.64%	
Approach 1: Edit distance &	68.20%	
Noisy Channel		
Approach 2: Adding heuristics	71.3%	
to Edit distance		
Approach 2 with	75%	
post-processing		

Integration with Applications

Applications of Arabic Language Technologies

