

Parts of Speech

Part 2

ICS 482 Natural Language Processing

Lecture 10:
Husni Al-Muhtaseb

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ

ICS 482 Natural Language Processing

Lecture 10: Parts of Speech
Part 2

Husni Al-Muhtaseb

NLP Credits and Acknowledgment

These slides were adapted from presentations of the Authors of the book

SPEECH and LANGUAGE PROCESSING:

An Introduction to Natural Language Processing, Computational Linguistics, and Speech Recognition

and some modifications from presentations found in the WEB by several scholars including the following

NLP Credits and Acknowledgment

If your name is missing please contact me
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NLP Credits and Acknowledgment

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Hinrich Schütze

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Gina-Anne Levow

Guitao Gao

Qing Ma

Zeynep Altan

Previous Lectures

- Pre-start questionnaire
- Introduction and Phases of an NLP system
- NLP Applications - Chatting with Alice
- Finite State Automata & Regular Expressions & languages
- Deterministic & Non-deterministic FSAs
- Morphology: Inflectional & Derivational
- Parsing and Finite State Transducers
- Stemming & Porter Stemmer
- 20 Minute Quiz
- Statistical NLP – Language Modeling
- N-Grams
- Smoothing and N-Gram: Add-one & Witten-Bell
- Return Quiz 1
- Parts of Speech

Today's Lecture

- Continue with Parts of Speech
- Arabic Parts of Speech

Parts of Speech

Start with eight basic categories

- Noun اسم
- Verb فعل
- preposition حرف جر
- Pronoun ضمير
- adjective صفة
- adverb ظرف
- Article أداة
- Conjunction حرف عطف

These categories are based on morphological and distributional properties (not semantics)

Some cases are easy, others are not

Parts of Speech

- Closed classes
 - Prepositions: on, under, over, near, by, at, from, to, with, etc.
 - Determiners: a, an, the, etc.
 - Pronouns: she, who, I, others, etc.
 - Conjunctions: and, but, or, as, if, when, etc.
 - Auxiliary verbs: can, may, should, are, etc.
 - Particles: up, down, on, off, in, out, at, by, etc.
- Open classes:
 - Nouns:
 - Verbs:
 - Adjectives:
 - Adverbs:

Sets of Parts of Speech:

Tagsets

- There are various standard tagsets to choose from; some have a lot more tags than others
- The choice of tagset is based on the application
- Accurate tagging can be done with even large tagsets

Some of the known Tagsets (English)

- Brown corpus: 87 tags
- Penn Treebank: 45 tags
- Lancaster UCREL C5: 61 tags
- Lancaster C7: 145 tags

Some of Penn Treebank tags

NN	noun	JJ	adjective
NNP	proper noun	CC	coord conj
DT	determiner	CD	cardinal number
IN	preposition	PRP	personal pronoun
VB	verb	RB	adverb
-R	comparative		
-S	superlative or plural		
-\$	possessive		

Verb inflection tags

VBP	base present	<i>take</i>
VB	infinitive	<i>take</i>
VBD	past	<i>took</i>
VBG	present participle	<i>taking</i>
VBN	past participle	<i>taken</i>
VBZ	present 3sg	<i>takes</i>
MD	modal	<i>can, would</i>

The entire Penn Treebank tagset

Tag	Description	Example	Tag	Description	Example
CC	Coordin. Conjunction	<i>and, but, or</i>	SYM	Symbol	<i>+, %, &</i>
CD	Cardinal number	<i>one, two, three</i>	TO	“to”	<i>to</i>
DT	Determiner	<i>a, the</i>	UH	Interjection	<i>ah, oops</i>
EX	Existential ‘there’	<i>there</i>	VB	Verb, base form	<i>eat</i>
FW	Foreign word	<i>mea culpa</i>	VBD	Verb, past tense	<i>ate</i>
IN	Preposition/sub-conj	<i>of, in, by</i>	VBG	Verb, gerund	<i>eating</i>
JJ	Adjective	<i>yellow</i>	VBN	Verb, past participle	<i>eaten</i>
JJR	Adj., comparative	<i>bigger</i>	VBP	Verb, non-3sg pres	<i>eat</i>
JJS	Adj., superlative	<i>wildest</i>	VBZ	Verb, 3sg pres	<i>eats</i>
LS	List item marker	<i>1, 2, One</i>	WDT	Wh-determiner	<i>which, that</i>
MD	Modal	<i>can, should</i>	WP	Wh-pronoun	<i>what, who</i>
NN	Noun, sing. or mass	<i>llama</i>	WP\$	Possessive wh-	<i>whose</i>
NNS	Noun, plural	<i>llamas</i>	WRB	Wh-adverb	<i>how, where</i>
NNP	Proper noun, singular	<i>IBM</i>	\$	Dollar sign	<i>\$</i>
NNPS	Proper noun, plural	<i>Carolinas</i>	#	Pound sign	<i>#</i>
PDT	Predeterminer	<i>all, both</i>	“	Left quote	<i>(‘ or “</i>
POS	Possessive ending	<i>'s</i>	”	Right quote	<i>(’ or ”</i>
PP	Personal pronoun	<i>I, you, he</i>	(Left parenthesis	<i>([, (, { , <</i>
PP\$	Possessive pronoun	<i>your, one's</i>)	Right parenthesis	<i>(] ,) , } , ></i>
RB	Adverb	<i>quickly, never</i>	,	Comma	<i>,</i>
RBR	Adverb, comparative	<i>faster</i>	.	Sentence-final punc	<i>(. ! ?)</i>
RBS	Adverb, superlative	<i>fastest</i>	:	Mid-sentence punc	<i>(: ; ... - -)</i>
RP	Particle	<i>up, off</i>			

UCREL C5

Tag	Description	Example
PNX	reflexive pronoun	<i>itself, ourselves</i>
POS	possessive 's or '	
PRF	the preposition <i>of</i>	
PRP	preposition (except <i>of</i>)	<i>for; above, to</i>
PUL	punctuation – left bracket	(or [
PUN	punctuation – general mark	: ! , ; - ? ...
PUQ	punctuation – quotation mark	' ' ”
PUR	punctuation – right bracket) or]
TOO	infinitive marker <i>to</i>	
UNC	unclassified items (not English)	
VBB	base forms of <i>be</i> (except infinitive)	<i>am, are</i>
VBD	past form of <i>be</i>	<i>was, were</i>
VBG	-ing form of <i>be</i>	<i>being</i>
VBI	infinitive of <i>be</i>	
VBN	past participle of <i>be</i>	<i>been</i>
VBZ	-s form of <i>be</i>	<i>is, 's</i>
VDB	base form of <i>do</i> (except infinitive)	<i>does</i>
VDD	past form of <i>do</i>	<i>did</i>
VDG	-ing form of <i>do</i>	<i>doing</i>
VDI	infinitive of <i>do</i>	<i>to do</i>
VDN	past participle of <i>do</i>	<i>done</i>
VDZ	-s form of <i>do</i>	<i>does</i>
VHB	base form of <i>have</i> (except infinitive)	<i>have</i>
VHD	past tense form of <i>have</i>	<i>had, 'd</i>
VHG	-ing form of <i>have</i>	<i>having</i>
VHI	infinitive of <i>have</i>	
VHN	past participle of <i>have</i>	<i>had</i>
VHZ	-s form of <i>have</i>	<i>has, 's</i>
VM0	modal auxiliary verb	<i>can, could, will, 'll</i>
VVB	base form of lexical verb (except infin.)	<i>take, live</i>
VVD	past tense form of lexical verb	<i>took, lived</i>
VVG	-ing form of lexical verb	<i>taking, living</i>
VVI	infinitive of lexical verb	<i>take, live</i>
VVN	past participle form of lex. verb	<i>taken, lived</i>
VVZ	-s form of lexical verb	<i>takes, lives</i>
XX0	the negative <i>not</i> or <i>n't</i>	
ZZ0	alphabetical symbol	<i>A, B, c, d</i>

Tagging

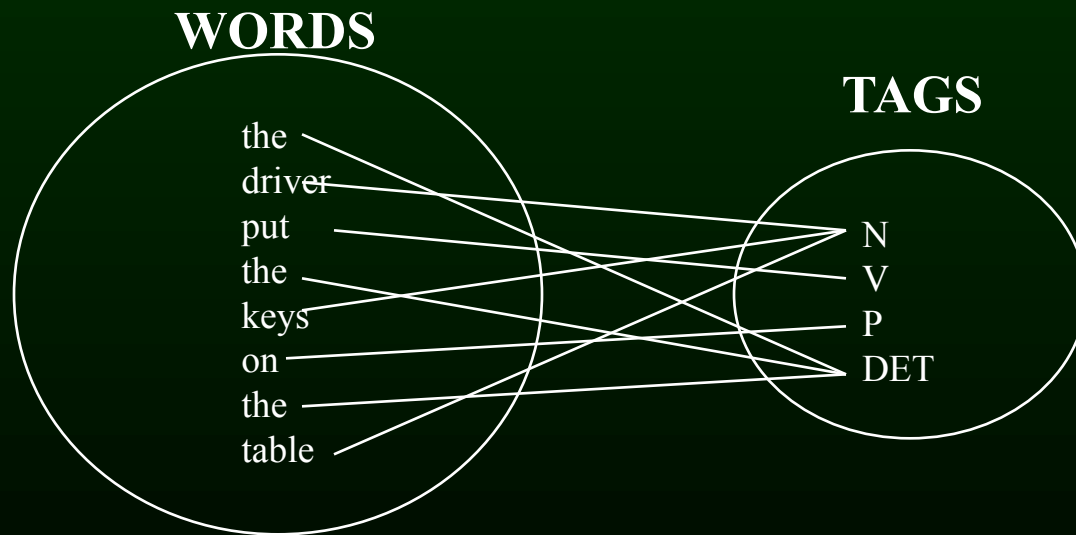
- Part of speech tagging is the process of assigning parts of speech to each word in a sentence...

Assume we have

- A tagset
- A dictionary that gives you the possible set of tags for each entry
- A text to be tagged
- A reason?

POS Tagging: Definition

- The process of assigning a part-of-speech or lexical class marker to each word in a corpus:



Tag Ambiguity (updated)

	87-tagset	45-tagset
Unambiguous (1 tag)	44,019	38,857
Ambiguous (2-7 tags)	5,490	8,844
2 tags	4,967	6,731
3 tags	411	1621
4 tags	91	357
5 tags	17	90
6 tags	2 (well, beat)	32
7 tags	2 (still, down)	6 (well, set, round, open, fit, down)
8 tags		4 ('s, half, back, a)
9 tags		3 (that, more, in)

- Most words are unambiguous
- Many of the most common English words are ambiguous

Tagging: Three Methods

- Rules
- Probabilities (Stochastic)
- Transformation-Based: Sort of both

Rule-based Tagging

- Use dictionary (lexicon) to assign each word a list of potential POS
- Use large lists of hand-written disambiguation rules to identify a single POS for each word.
- Example of rules: NP \rightarrow Det (Adj*) N
 - For example: *the clever student*

Probabilities: Tagging with lexical frequencies

- Sami is expected to race tomorrow.
- Sami/NNP is/VBZ expected/VBN to/TO **race**/VB tomorrow/NN
- People continue to inquire the reason for the race for outer space.
- People/NNS continue/VBP to/TO inquire/VB the/DT reason/NN for/IN the/DT **race**/NN for/IN outer/JJ space/NN
- Problem: assign a tag to **race** given its lexical frequency
- Solution: we choose the tag that has the greater
 - $P(\text{race}|\text{VB})$
 - $P(\text{race}|\text{NN})$
- Actual estimate from the Switchboard corpus:
 - $P(\text{race}|\text{NN}) = .00041$
 - $P(\text{race}|\text{VB}) = .00003$

Transformation-based: The Brill Tagger

- An example of Transformation-based Learning
- Very popular (freely available, works fairly well)
- A SUPERVISED method: requires a tagged corpus
- Basic idea: do a quick job first (using frequency), then revise it using contextual rules

An example

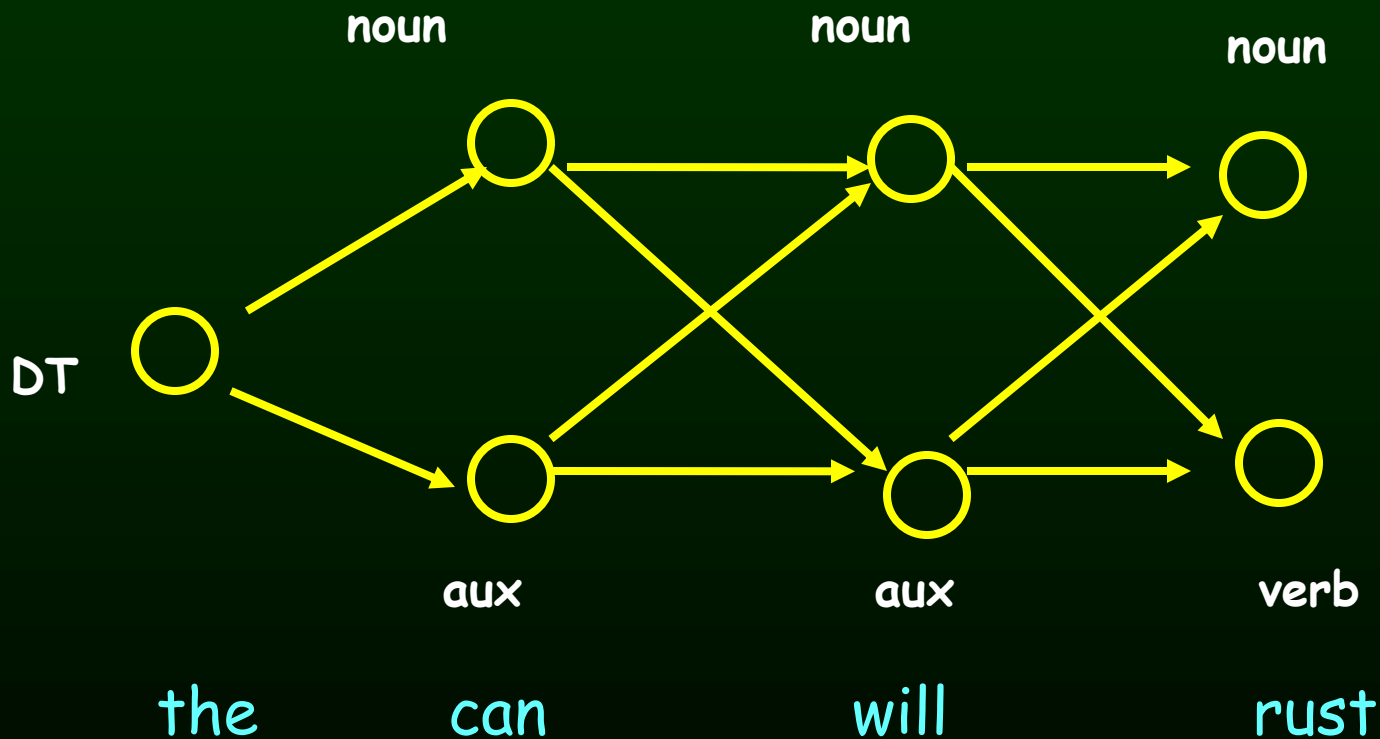
- Examples:
 - It is expected to **race** tomorrow.
 - The **race** for outer space.
- Tagging algorithm:
 1. Tag all uses of “race” as NN (most likely tag in the Brown corpus)
 - It is expected to **race/NN** tomorrow
 - the **race/NN** for outer space
 2. Use a transformation rule to replace the tag NN with VB for all uses of “race” preceded by the tag TO:
 - It is expected to **race/VB** tomorrow
 - the **race/NN** for outer space

Stochastic (Probabilities)

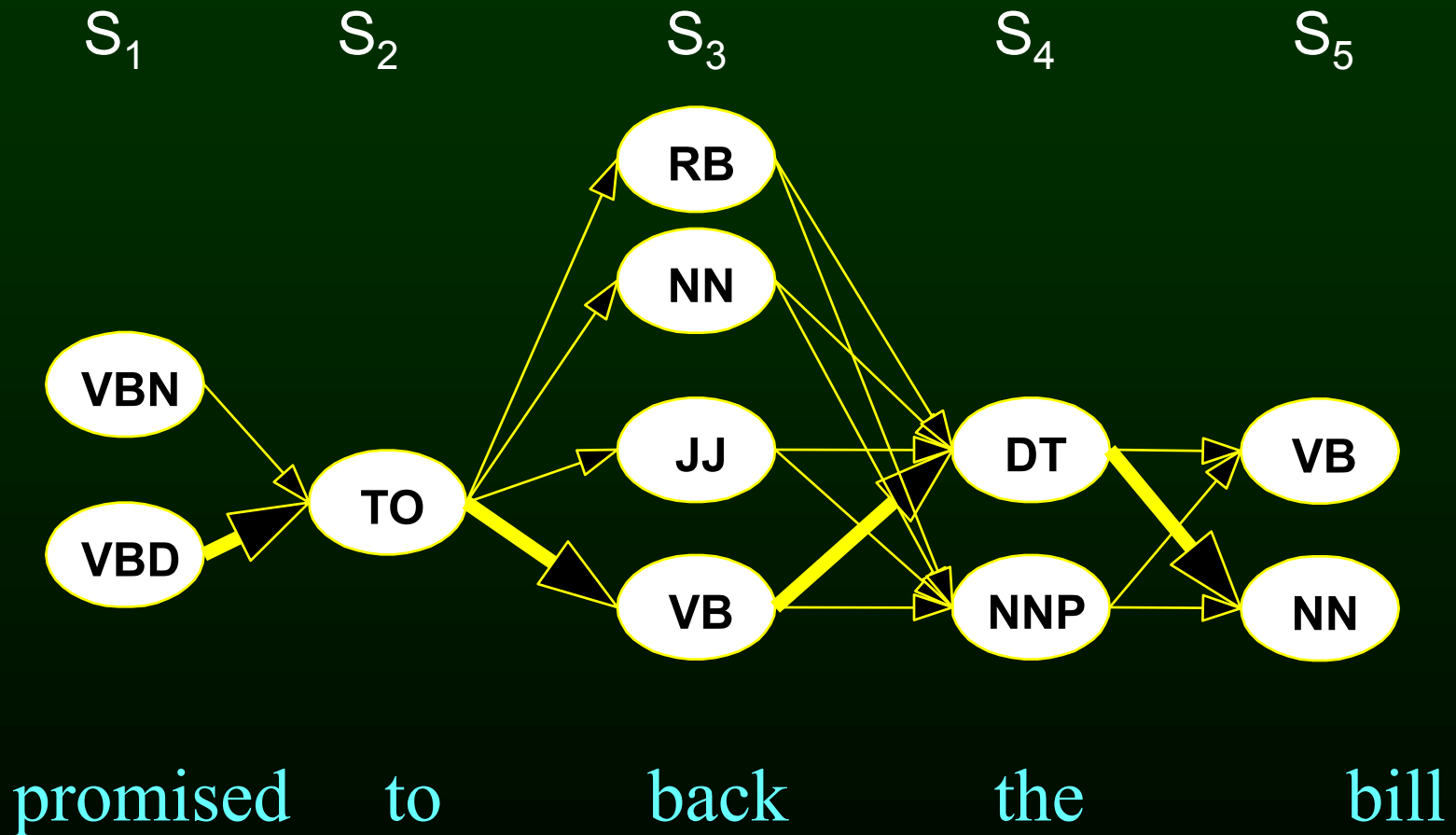
- Simple approach
 - Disambiguate words based on the probability that a word occurs with a particular tag
- N-gram approach
 - The best tag for given words is determined by the probability that it occurs with the n previous tags
- Viterbi Algorithm
 - Trim the search for the most probable tag using the best N Maximum Likelihood Estimates (N is the number of tags of the following word)
- Hidden Markov Model combines the above two approaches

Viterbi Maximum Likelihood Estimates

Want the most likely path through this graph.



Viterbi Maximum Likelihood Estimates



Viterbi Maximum Likelihood Estimates

- We want the best set of tags for a sequence of words (a sentence)
- W is a sequence of words
- $W = w_1 w_2 w_3 \dots w_n$
- T is a sequence of tags
- $T = t_1 t_2 t_3 \dots t_n$

$$\arg \max P(T | W) = \frac{P(W | T)P(T)}{P(W)}$$

- $P(w)$ is common

Viterbi Maximum Likelihood Estimates

- We want the best set of tags for a sequence of words (a sentence)
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$$\arg \max P(T | W) = P(W | T) P(T)$$

- $P(w)$ is common

Stochastic POS Tagging: Example

- 1) Sami is expected to **race** tomorrow.
- 2) People continue to inquire the reason for the **race** for outer space.

Stochastic POS Tagging: Example

Example: suppose $w_i = \text{race}$, a verb (VB) or a noun (NN)?

Assume that other mechanism has already done the best tagging to the surrounding words, leaving only **race** untagged

- 1) Sami/NNP is/VBZ expected/VBN to/TO **race/?** tomorrow/NN
- 2) People/NNS continue/VBP to/TO inquire/VB the/DT reason/NN for/IN the/DT **race/?** For/IN outer/JJ space/NN

Bigram

$$t_i = \arg \max_j P(t_j | t_{i-1}) P(w_i | t_j)$$

Simplify the problem:
to/To **race/???**
the/DT **race/???**

$P(\text{VB}|\text{TO}) P(\text{race} | \text{VB})$
 $P(\text{NN}|\text{TO}) P(\text{race} | \text{NN})$

Where is the data?

Look at the Brown and Switchboard corpora

$$P(\text{NN} \mid \text{TO}) = 0.021$$

$$P(\text{VB} \mid \text{TO}) = 0.34$$

If we are expecting a verb, how likely it would be “race”

$$P(\text{race} \mid \text{NN}) = 0.00041$$

$$P(\text{race} \mid \text{VB}) = 0.00003$$

Finally:

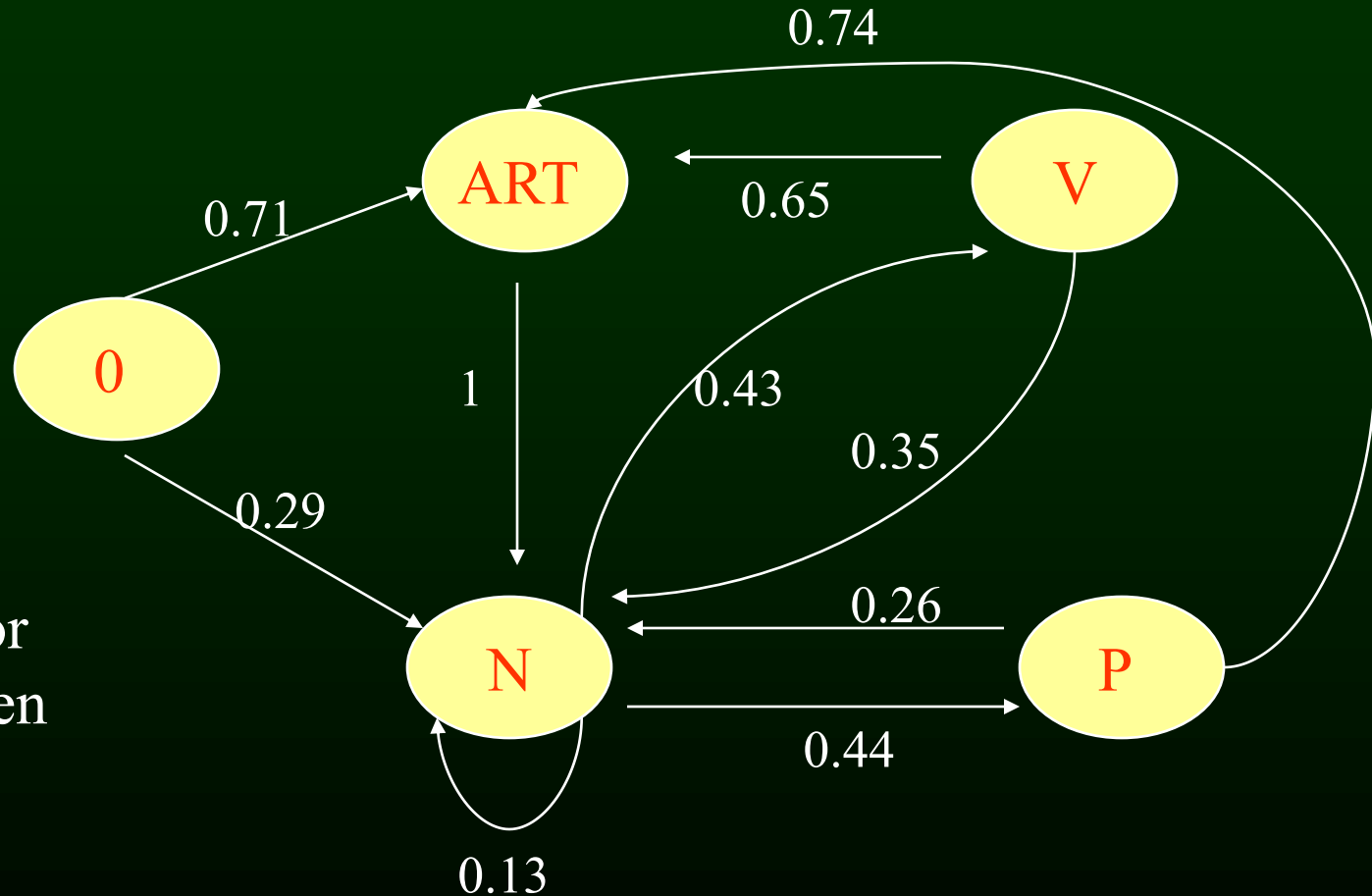
$$P(\text{NN} \mid \text{TO}) P(\text{race} \mid \text{NN}) = 0.000007$$

$$P(\text{VB} \mid \text{TO}) P(\text{race} \mid \text{VB}) = 0.00001$$

Example: Bigram of Tags from a Corpus

Cat	# at i	Pair	# at i, i+1	Bigram	Estimate
0	300	0, ART	213	Prob(ART 0)	0.71
0	300	0, N	87	Prob(N 0)	0.29
ART	558	ART, N	558	Prob(N ART)	1
N	833	N, V	358	Prob(V N)	0.43
N	833	N, N	108	Prob(N N)	0.13
N	833	N, P	366	Prob(P N)	0.44
V	300	V, N	75	Porb(N V)	0.35
V	300	V, ART	194	Prob(ART V)	0.65
P	307	P, ART	226	Prob (ART P)	0.74
P	307	P, N	81	Prob (N P)	0.26

A Markov Chain



assume
0.0001 for
any unseen
bigram

Word Counts

	N	V	ART	P	Total
flies	21	23	0	0	44
fruit	49	5	1	0	55
like	10	30	0	21	61
a	1	0	201	0	202
the	1	0	300	2	303
flower	53	15	0	0	68
flowers	42	16	0	0	58
birds	64	1	0	0	65
others	592	210	56	284	1142
Total	833	300	558	307	1998

Computing Probabilities using previous Tables

$$P(\textit{the} / \textit{ART}) = 300/558 = 0.54$$

$$P(\textit{flies} / \textit{N}) = 0.025$$

$$P(\textit{flies} / \textit{V}) = 0.076$$

$$P(\textit{like} / \textit{V}) = 0.1$$

$$P(\textit{like} / \textit{P}) = 0.068$$

$$P(\textit{like} / \textit{N}) = 0.012$$

$$P(\textit{a} / \textit{ART}) = 0.360$$

$$P(\textit{a} / \textit{N}) = 0.001$$

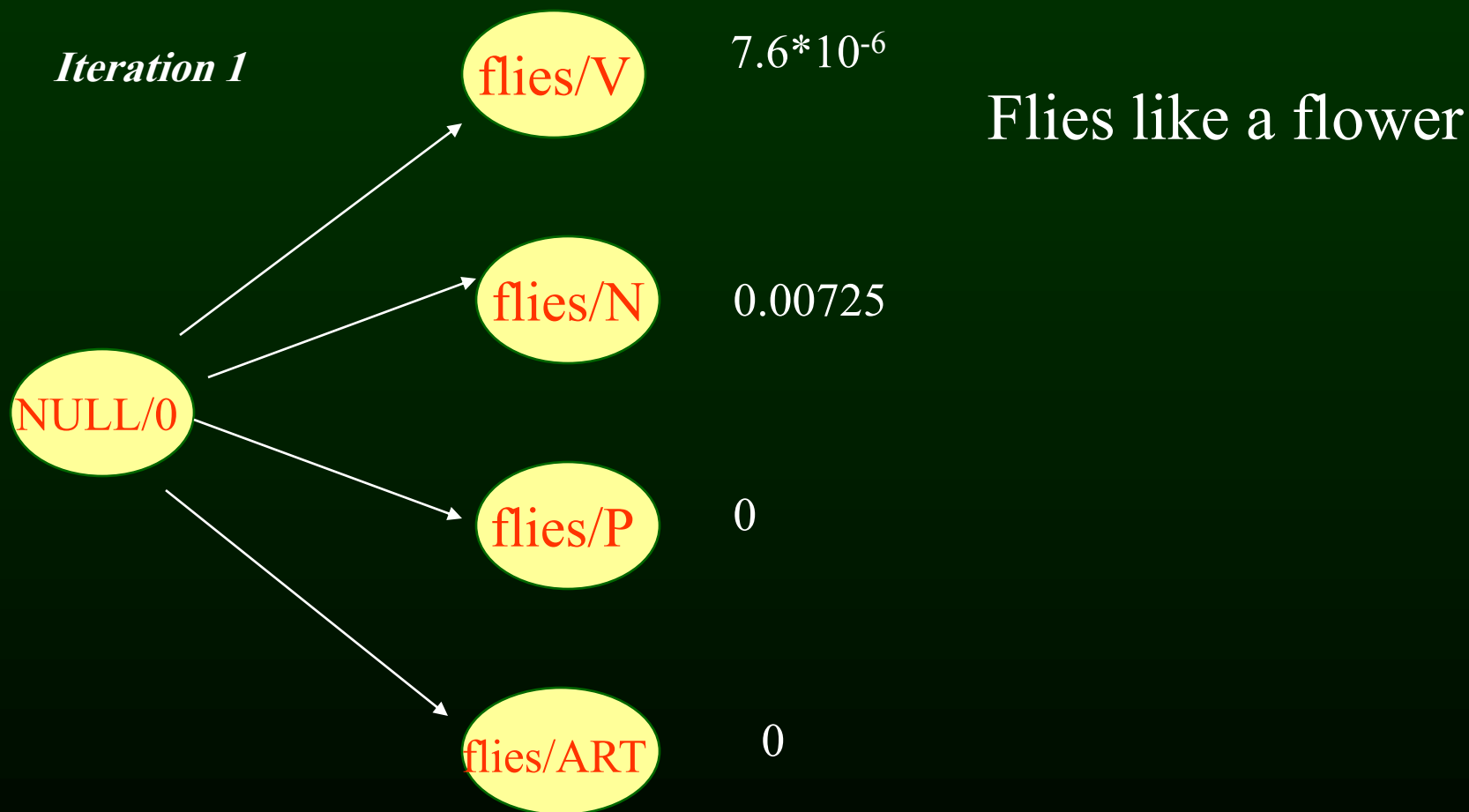
$$P(\textit{flower} / \textit{N}) = 0.063$$

$$P(\textit{flower} / \textit{V}) = 0.05$$

$$P(\textit{birds} / \textit{N}) = 0.076$$

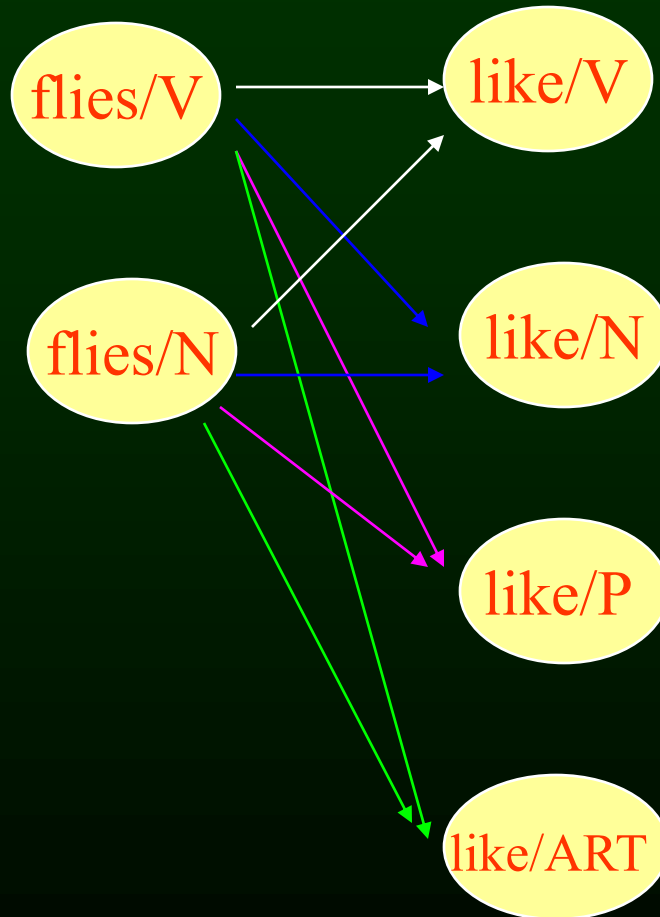
Viterbi Algorithm - Example

assume 0.0001 for any unseen bigram



Viterbi Algorithm - Example

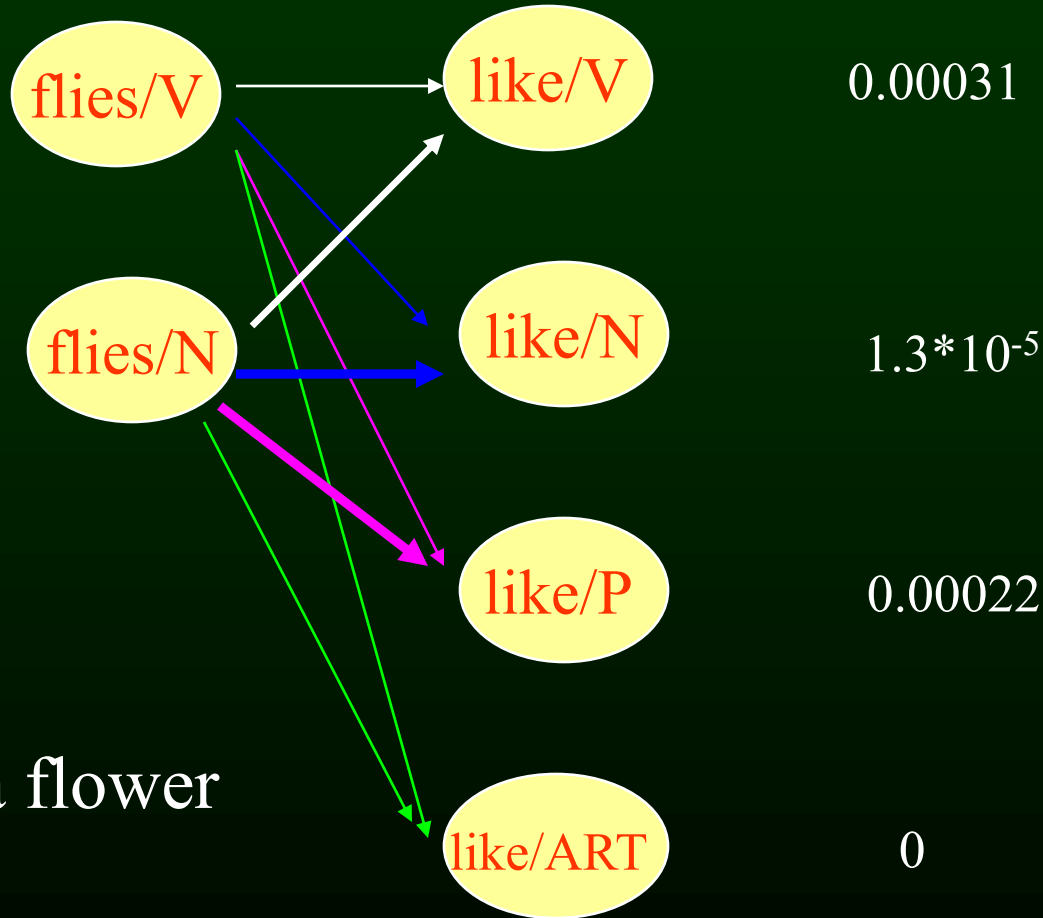
Iteration 2



Flies like a flower

Viterbi Algorithm - Example

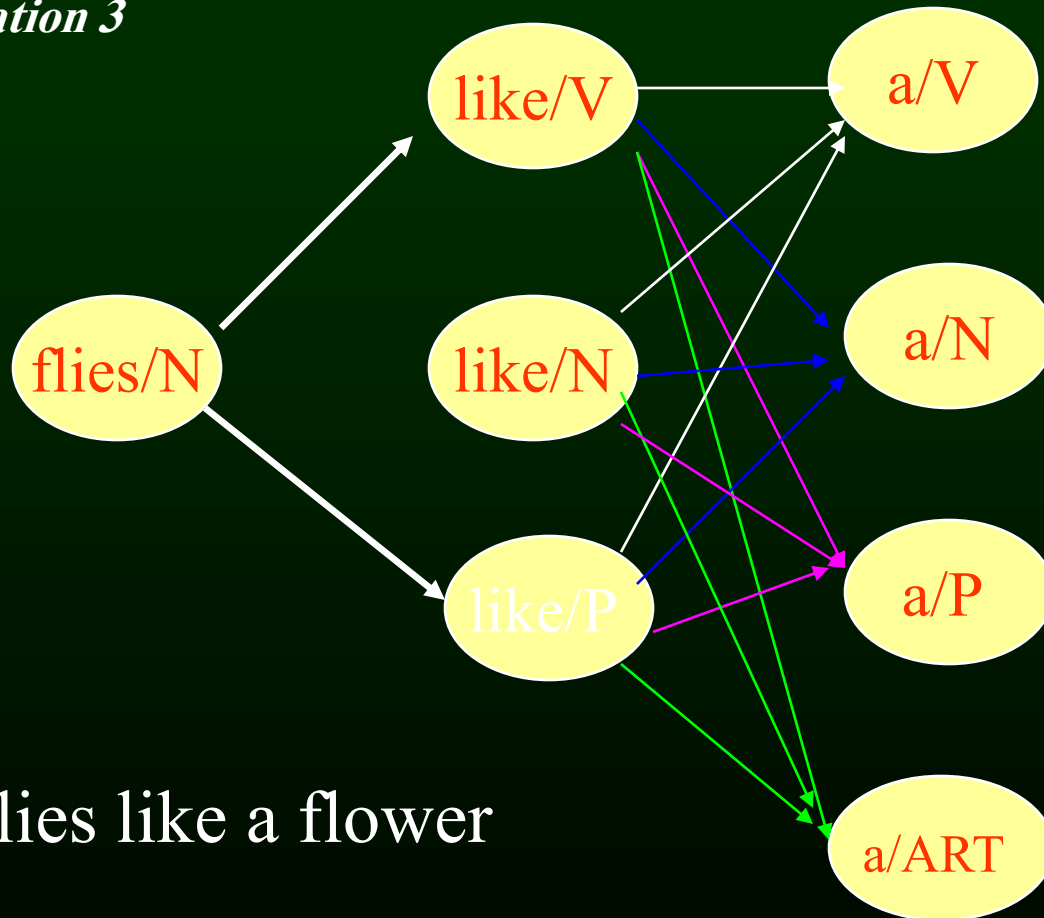
Iteration 2



Flies like a flower

Viterbi Algorithm - Example

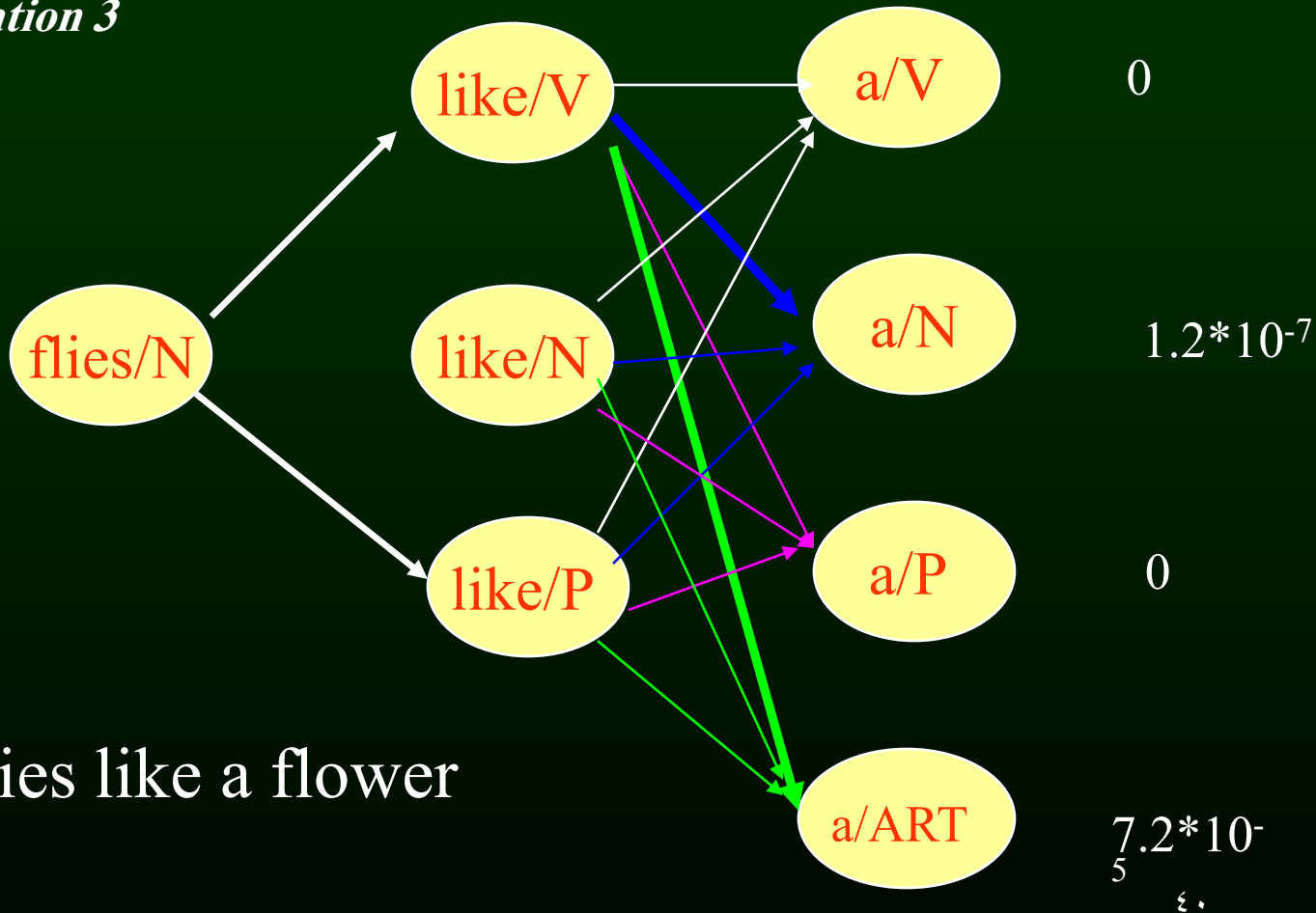
Iteration 3



Flies like a flower

Viterbi Algorithm - Example

Iteration 3

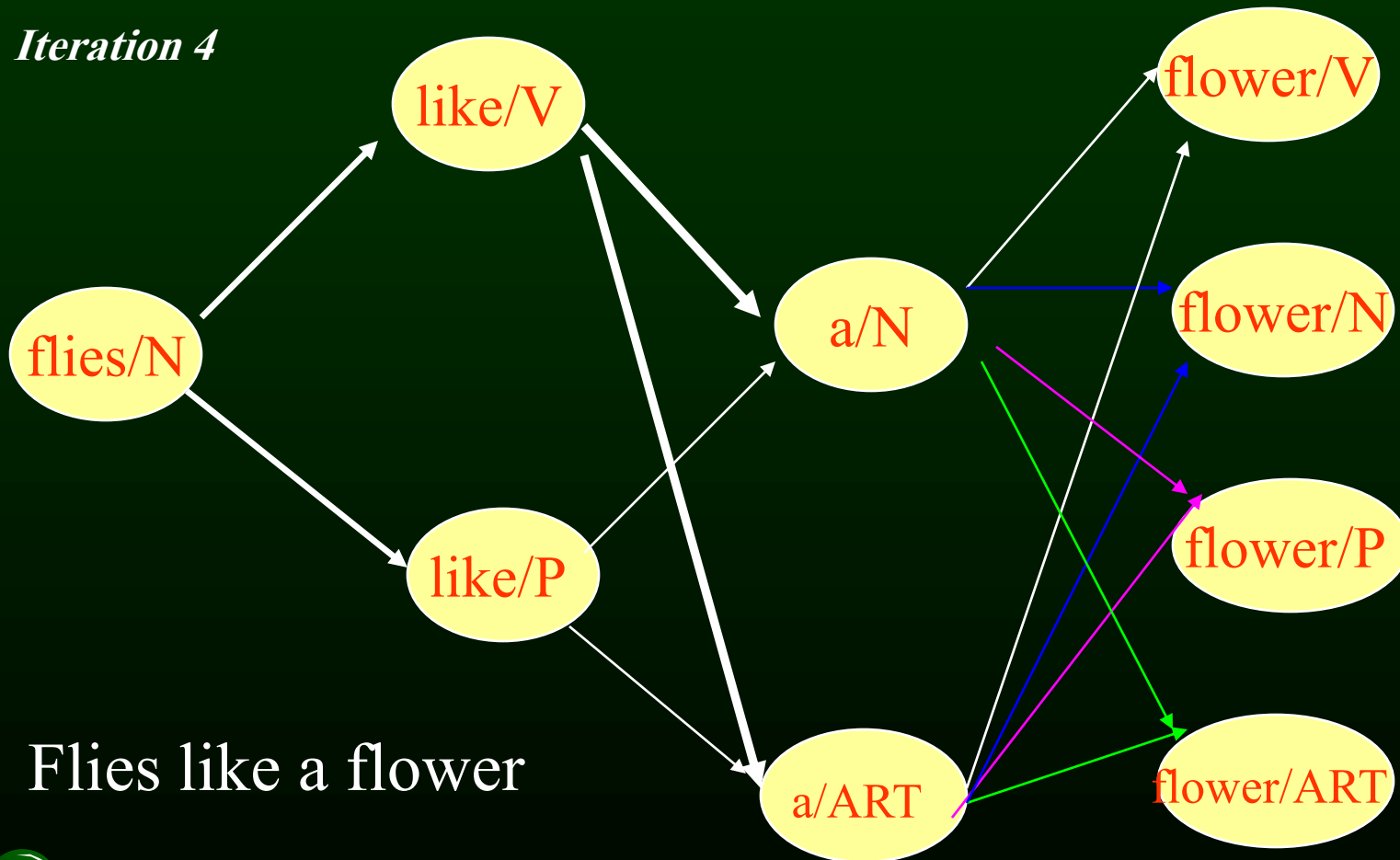


Flies like a flower

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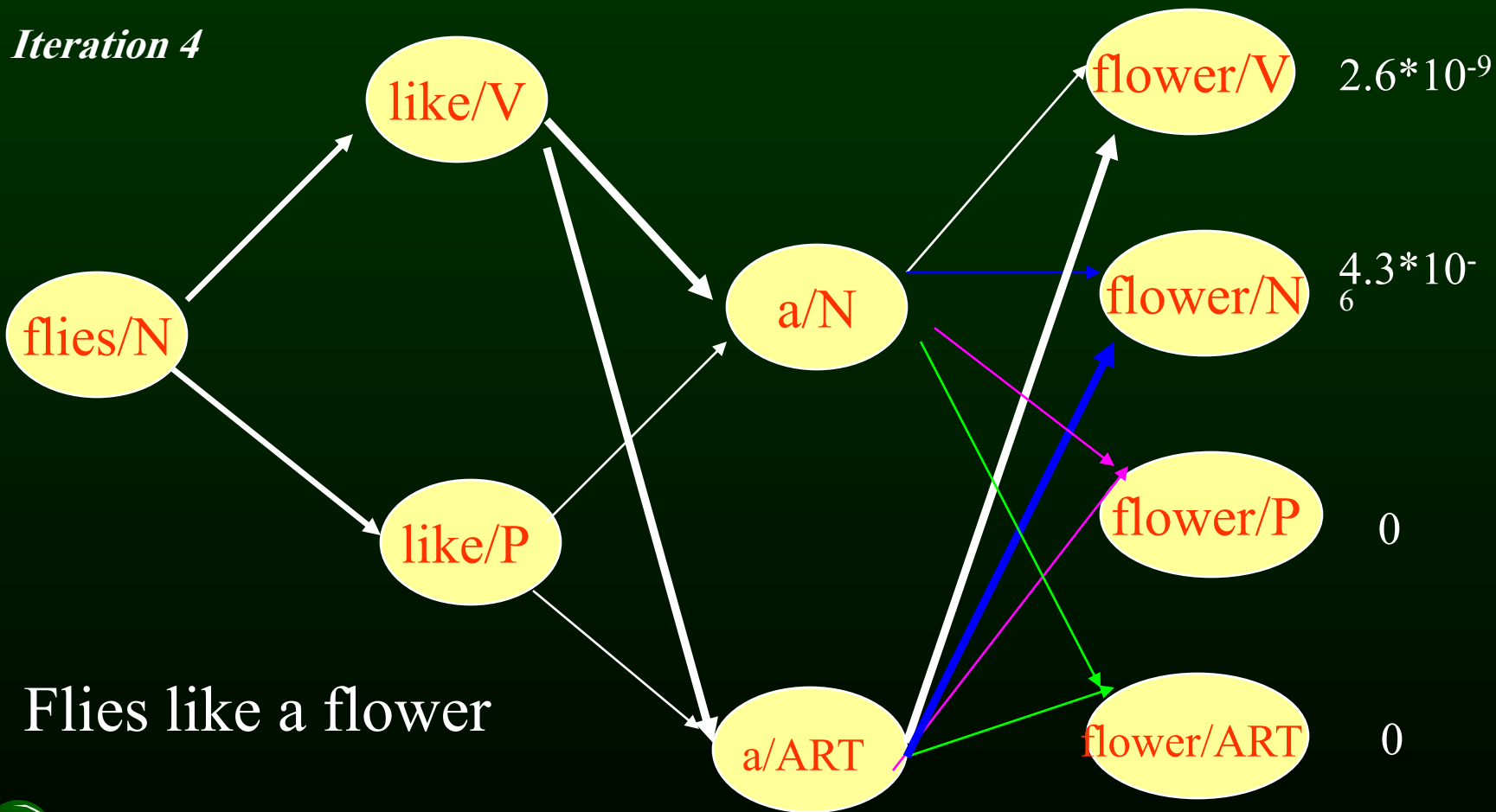
Viterbi Algorithm - Example

Iteration 4



Flies like a flower

Viterbi Algorithm - Example



Performance

- This method has achieved 95-96% correct with reasonably complex English tagsets and reasonable amounts of hand-tagged training data.
- Forward pointer... its also possible to train a system without hand-labeled training data

How accurate are they?

- POS Taggers boast accuracy rates of 95-99%
 - Vary according to text/type/genre
 - Of pre-tagged corpus
 - Of text to be tagged
- Worst case scenario: assume success rate of 95%
 - Prob(one-word sentence) = .95
 - Prob(two-word sentence) = $.95 * .95 = 90.25\%$
 - Prob(ten-word sentence) = 59% approx

- End of Part 1

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ
معالجة اللغات الطبيعية

Natural Language Processing

Lecture 10: Parts of Speech 2-2 **Morphosyntactic Tagset Of Arabic**

أوسام صرفية - نحوية للعربية

Husni Al-Muhtaseb

أوسام صرفية - نحوية للعربية

- شيرين خوجة Shereen Khoja
- ١٧٧ وسمًا 177 tags
- ١٠٣ للأسماء 103 Nouns
- ٥٧ للأفعال 57 Verbs
- ٩ للأدوات 9 Particles
- ٧ للفضلة 7 Residual
- ١ لعلامات الترقيم 1 Punctuation

اللغة العربية

- اللغة العربية من اللغات السامية
- تبنى الكلمات بإضافة زوائد إلى الجذر واستخدام أوزان محددة
- درس مدرس
- ثلاثة أجناس: three genders
 - مذكر Masculine
 - مؤنث Feminine
 - حيادي Neuter

اللغة العربية

- الخطاب Three persons
 - المتكلم The speaker
 - المخاطب The person being addressed
 - الغائب The person that is not present
- العدد Three numbers
 - مفرد Singular
 - مثنى Dual
 - جمع Plural

اللغة العربية

• حالات الفعل three moods of the verb

• الرفع Indicative

• النصب Subjunctive

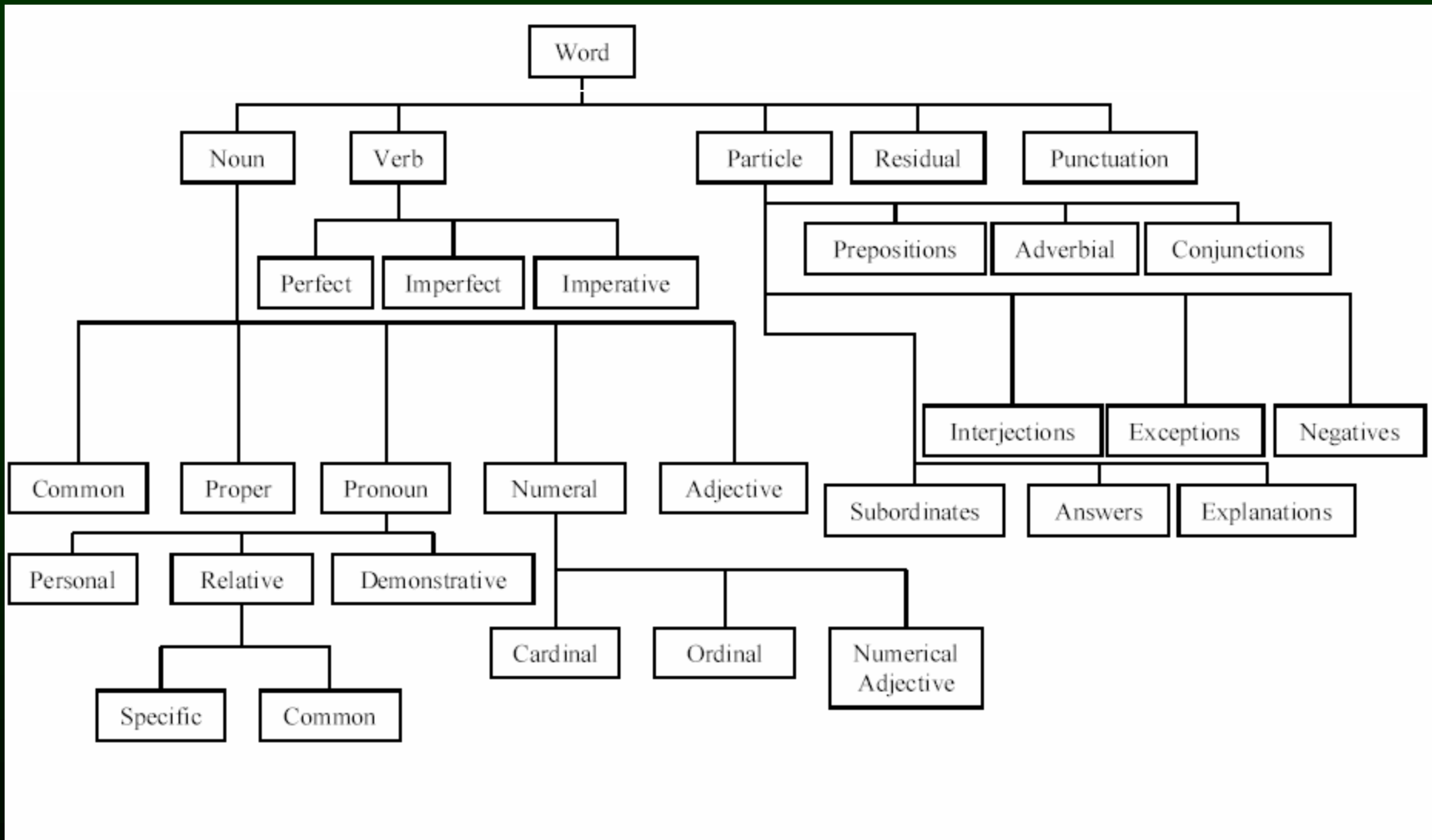
• الجزم Jussive

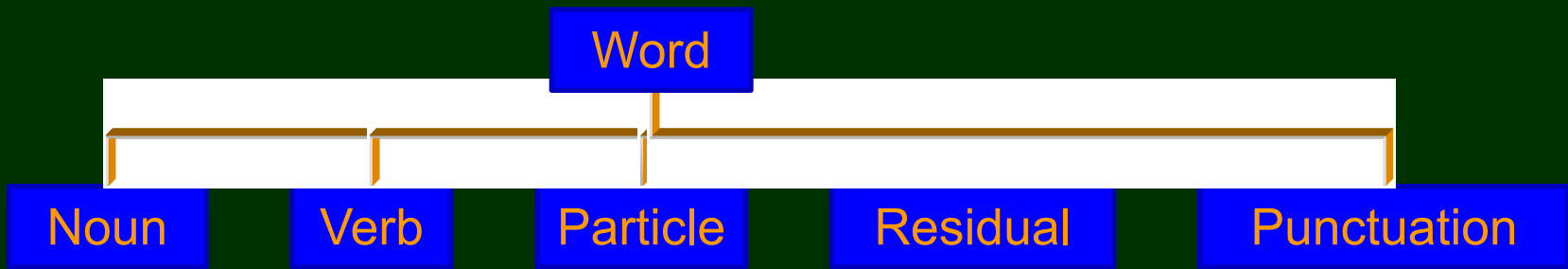
• حالات الاسم three case forms of the noun

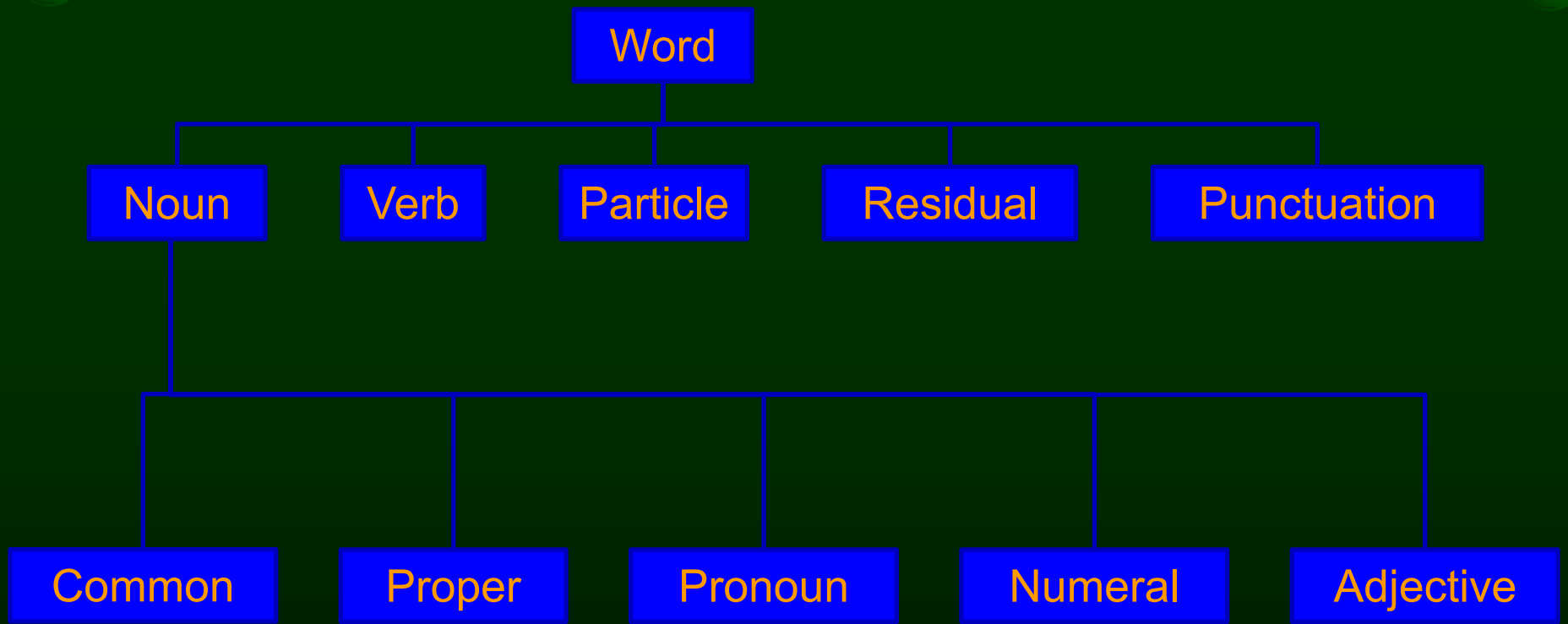
• الرفع Nominative

• النصب Accusative

• الجر Genitive







Word

Noun

Verb

Particle

Residual

Punctuation

Common

Proper

Pronoun

Numeral

Adjective

Personal

Relative

Demonstrative

Word

Noun

Verb

Particle

Residual

Punctuation

Common

Proper

Pronoun

Numeral

Adjective

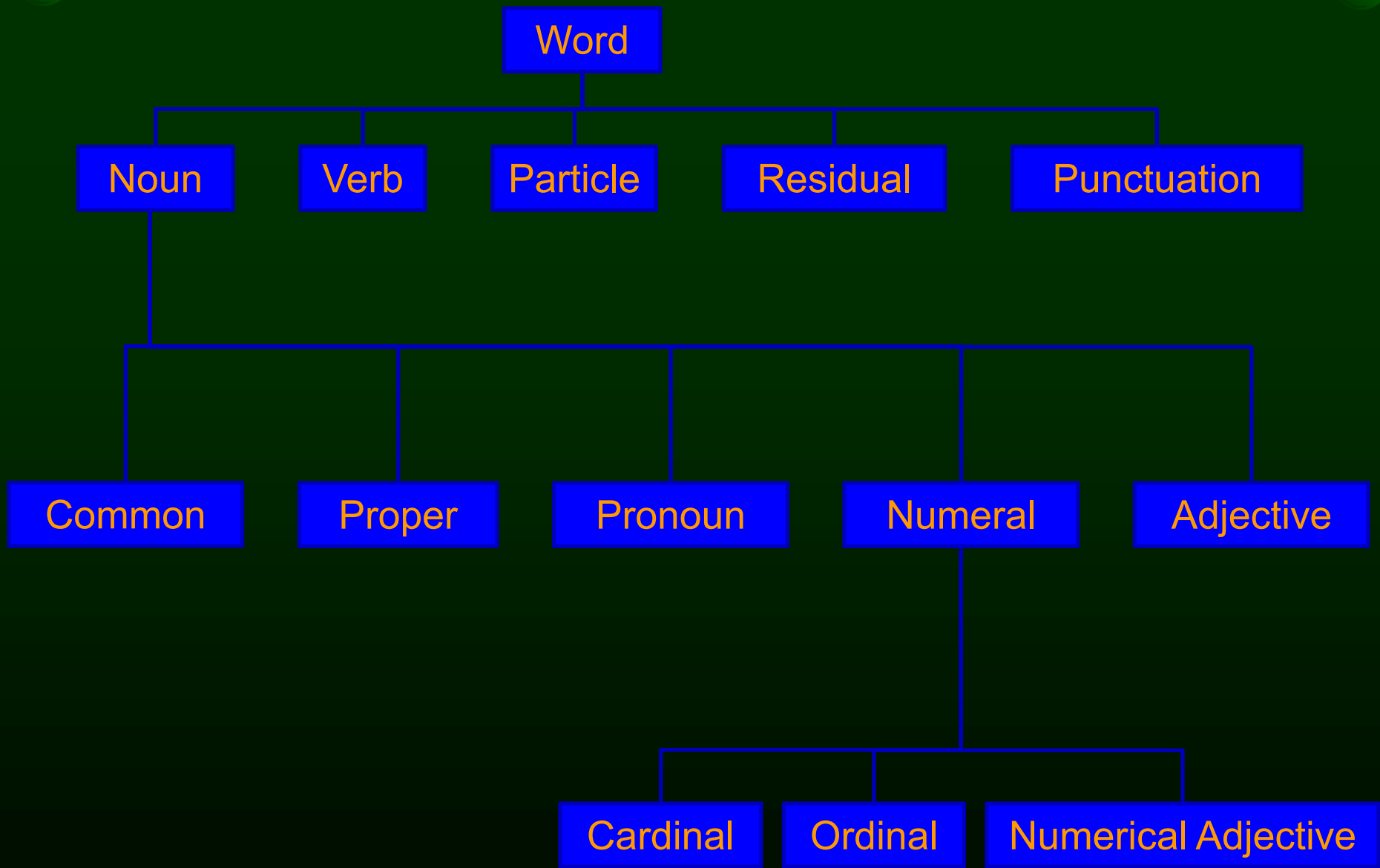
Personal

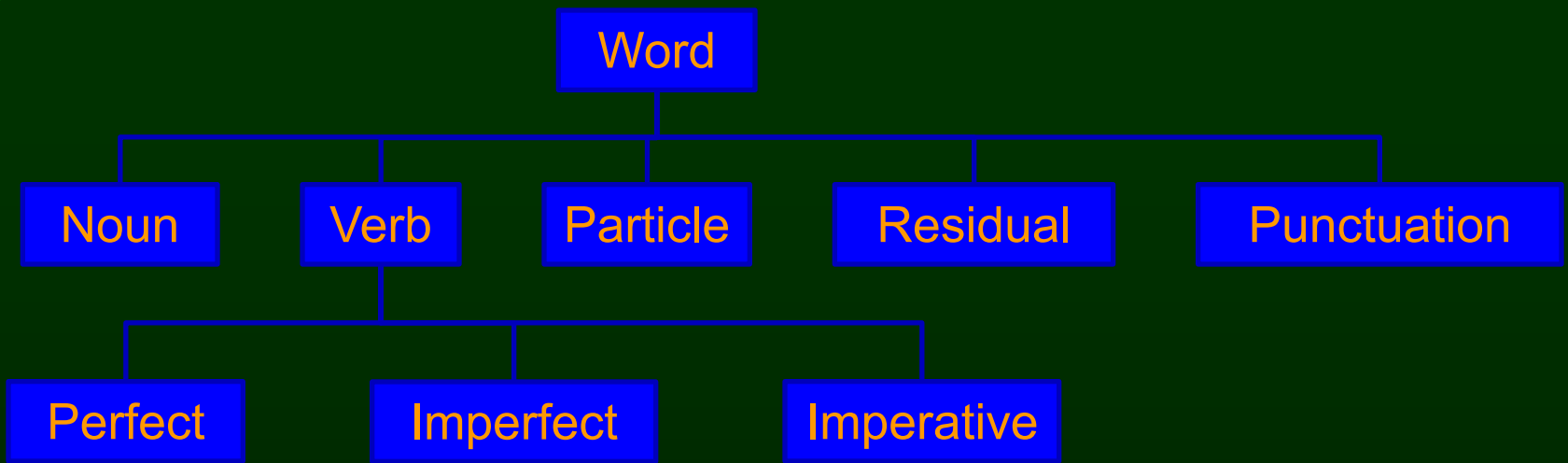
Relative

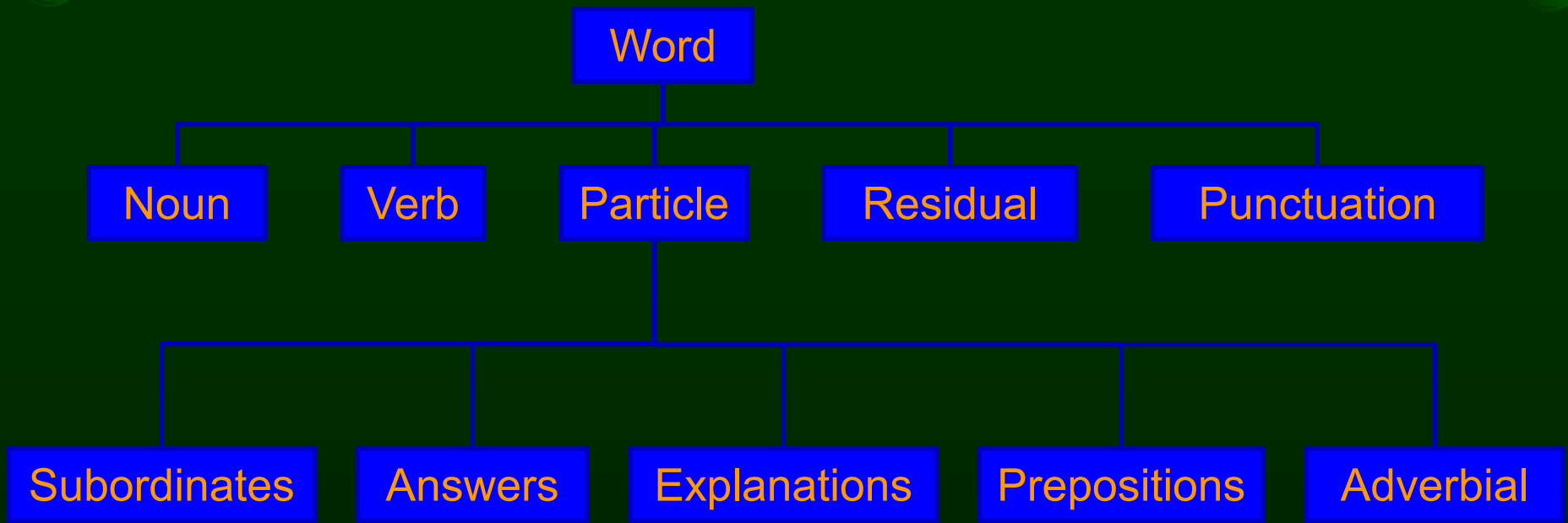
Demonstrative

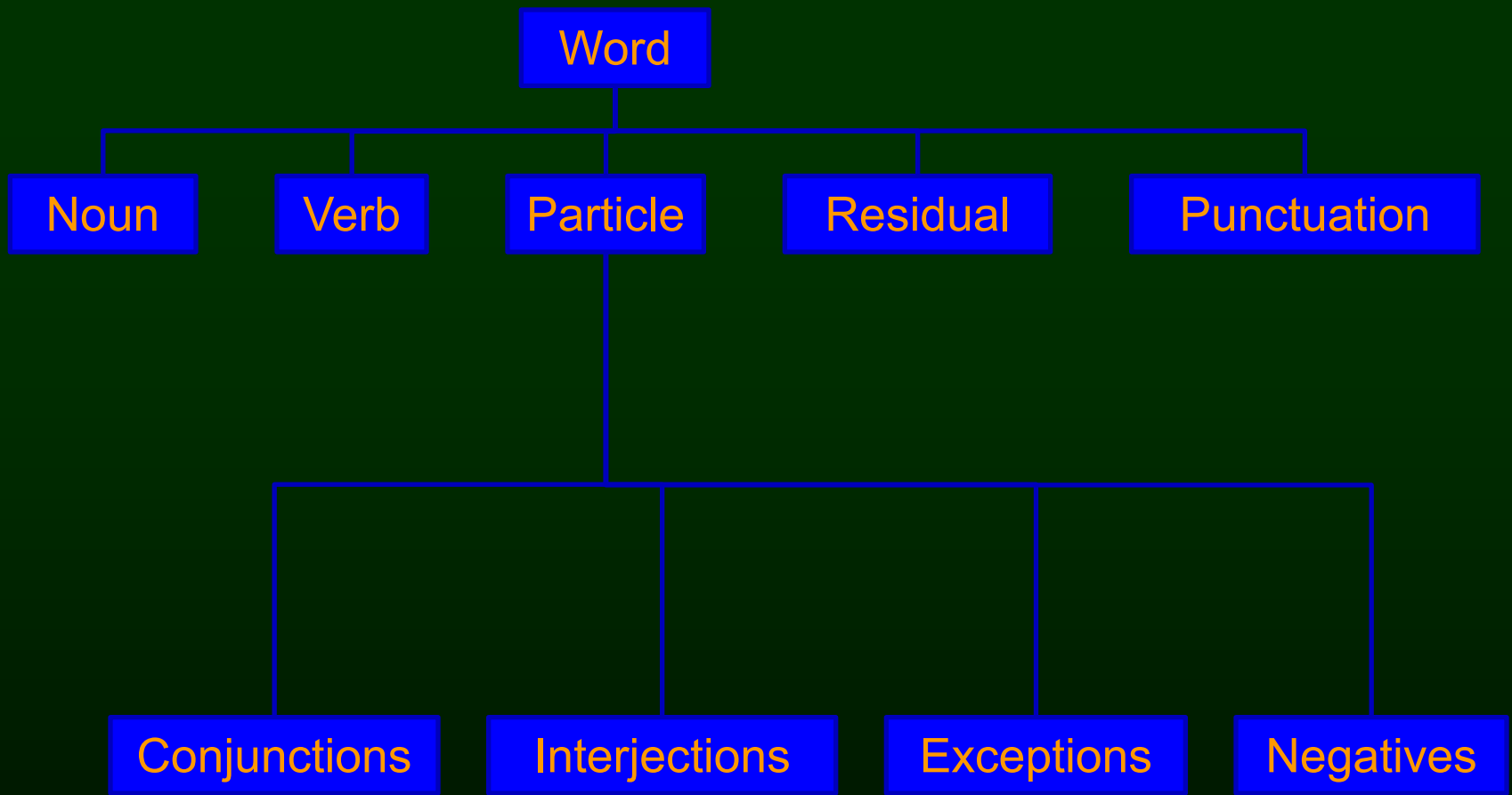
Specific

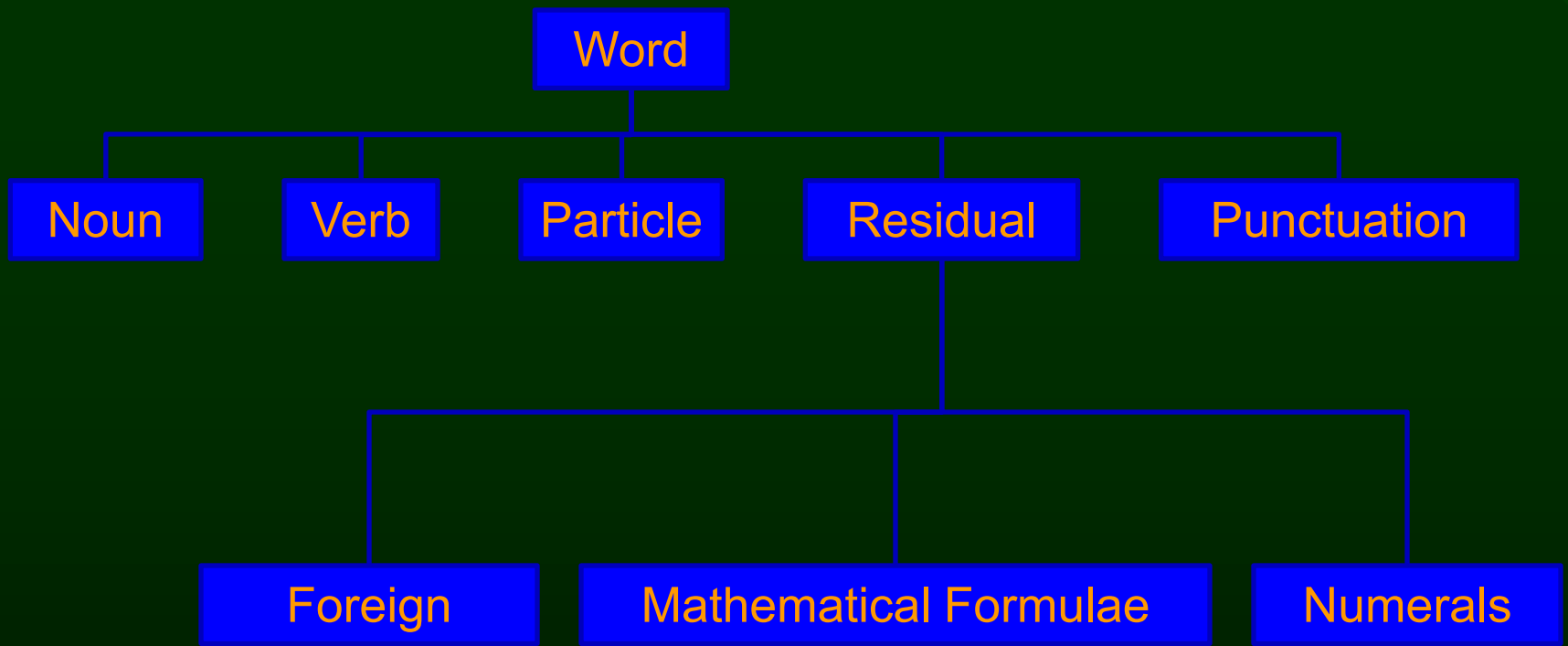
Common

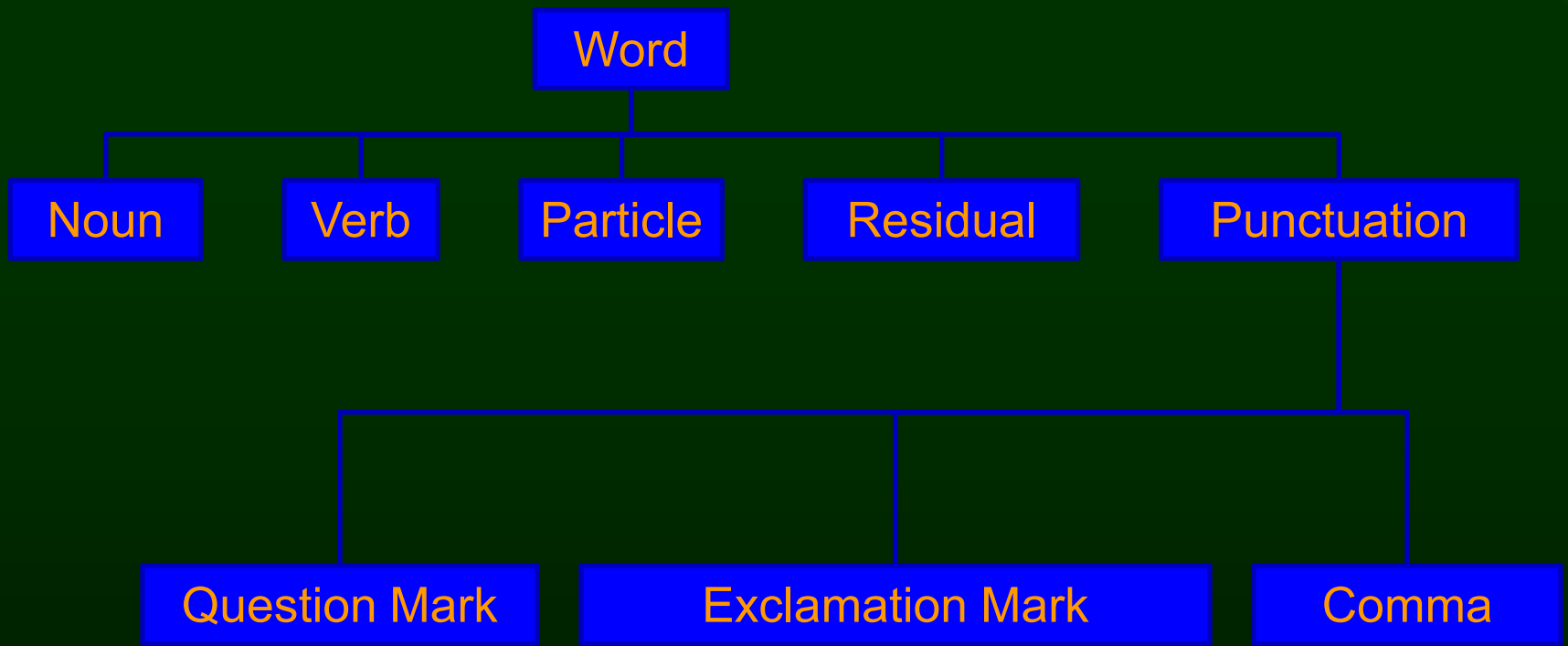




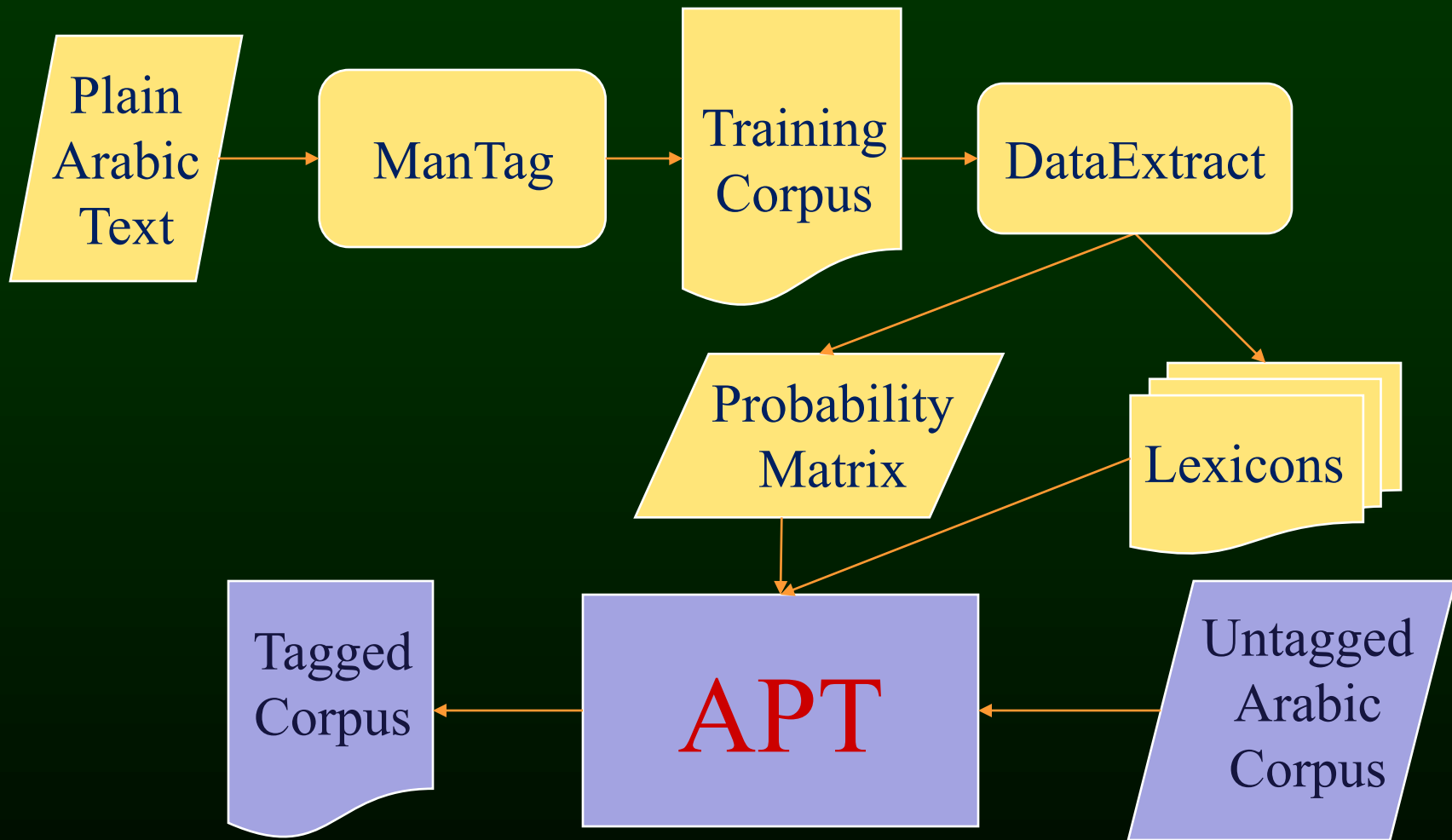








Arabic POS Tagger



DataExtract Process

- Takes in a tagged corpus and extracts various lexicons and the probability matrix
 - Lexicon that includes all clitics.
 - (Sprout, 1992) defines a clitic as “a *syntactically* separate word that functions *phonologically* as an affix”
 - Lexicon that removes all clitics before adding the word

DataExtract Process

- Produces a probability matrix for various levels of the tagset
 - Lexical probability: probability of a word having a certain tag
 - Contextual probability: probability of a tag following another tag

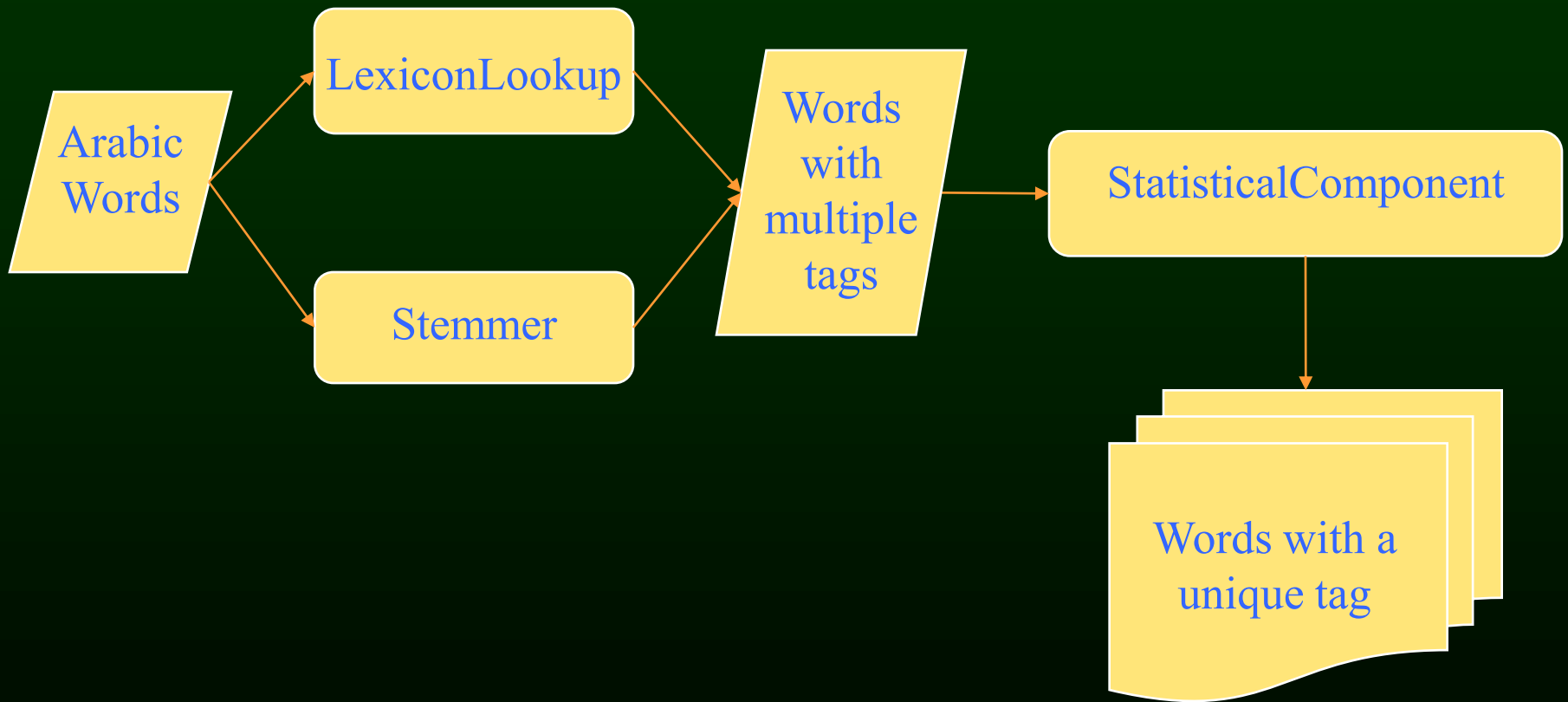
DataExtract Process

	N	V	P	No.	Pu.
N	0.711	0.065	0.143	0.010	0.071
V	0.926	0.037	0.0	0.008	0.029
P	0.689	0.199	0.085	0.016	0.011
No.	0.509	0.06	0.098	0.009	0.324
Pu.	0.492	0.159	0.152	0.046	0.151

Arabic Corpora

- 59,040 words of the Saudi “al-Jazirah” newspaper, dated 03/03/1999
- 3,104 words of the Egyptian “al-Ahram” newspaper, date 25/01/2000
- 5,811 words of the Qatari “al-Bayan” newspaper, date 25/01/2000
- 17,204 words of al-Mishkat, an Egyptian published paper in social science, April 1999

APT: Arabic Part-of-speech Tagger



كلمة

علامة ترقيم

اسم

فضلة

أداة

فعل

عدد

علم

صفة

ضمير

جنس

تفسير

ماض

صفة عددية

إشارة

شخصي

موصول

تمني

مضارع

عددي

عام

نفي

أمر

ترتيبي

محدد

استثناء

ظرف

إجابة

جر

عطف

نداء

٥ فئات أساسية

- 1. N [noun] اسم
- 2. V [verb] فعل
- 3. P [particle] أداة
- فضلة: الكلمات الغربية والصيغ الرياضية والأرقام
- 4. R [residual]
• علامة ترقيم (كلها)
- 5. PU [punctuation]: all

الاسم

- 1.1. C [common] • جنس
- 1.2. P [proper] • علم
- 1.3. Pr [pronoun] • ضمير
- 1.4. Nu [numeral] • عدد
- 1.5. A [adjective] • صفة

أمثلة على الأسماء

• Singular, masculine, accusative, common noun

أخذ الولد كتاباً

مفرد مذكر منصوب اسم جنس

• Singular, masculine, genitive, common noun

درست من الكتاب

مفرد مذكر مجرور اسم جنس

• Singular, feminine, nominative, common noun

هذه مدرسة

مفرد مؤنث مرفوع اسم جنس

1.3.1. P [personal] الشخصية

detached words such as **هو** مثل منفصلة

attached to a word **متصلة** متصلة

to nouns to indicate **كتابها** مع الأسماء للملكية مثل
possession

to verbs as direct **ضربه** متصلة مع الأفعال كمفعول به مثل
object

prepositions **فيه** متصلة مع حروف الجر مثل

1.3.2. R [relative] الموصولة

1.3.3. D [demonstrative] الإشارة

الضمائر

Third person, singular, masculine, personal •
pronoun

هو

ضمير الغائب مفرد مذكر شخصي

Singular, feminine, demonstrative pronoun •

هذه

اسم إشارة مؤنث مفرد

الموصول الضمائر *Relative Pronoun*

1.3.2.1. S [specific]

• محددة

1.3.2.2. C [common]

• عامة

• أمثلة

• اللتان

ضمير موصول محدد مؤنث مثنى

• Dual, feminine, specific, relative pronoun

• الذين

ضمير موصول محدد مذكر جمع

• Plural, masculine, specific, relative pronoun

• ضمير موصول عام

• Common, relative pronoun

العدد

1.4.1. Ca [cardinal] عددي •

1.4.2.O [ordinal] ترتيب •

1.4.3. Na [numerical adjective]: صفة عددية •

• تصف عدد جهات شكل ثماني

• وصف شخصين مثلا ثنائي

• أمثلة

Singular, masculine, nominative, indefinite cardinal number •

أربعة

مفرد مذكر مرفوع نكرة عدد عددي

Singular, masculine, nominative, indefinite ordinal number •

رابع

مفرد مذكر مرفوع نكرة عدد ترتيبي

Singular, masculine, numerical adjective •

رباعي

مفرد مذكر صفة عددية

الخواص اللغوية للاسم

الجنس Gender

M [masculine] مذكر

F [feminine] مؤنث

N [neuter] محايد

العدد Number

Sg [singular] مفرد

Du [dual] مثنى

Pl [plural] جمع

الخطاب Person

1 [first] متكلم

2 [second] مخاطب

3 [third] غائب

الحالة Case

N [nominative] رفع

A [accusative] نصب

G [genitive] جر

التعريف Definiteness

D [definite] معرفة

I [indefinite] نكرة

الأفعال Verbs

- 1. P [perfect] تام (ماض) •
- 2. I [imperfect] مستمر (مضارع) •
- 3. Iv [imperative] أمر •
- أمثلة •

كَسَرْتُ •

فعل ماض حيادي مفرد متكلم

First person, singular, neuter, perfect verb •

أَكْسِرُ •

فعل مضارع مرفوع مفرد متكلم

First person, singular, neuter, indicative, imperfect verb •

اِكْسِرْ •

فعل أمر مذكر مفرد مخاطب

Second person, singular, masculine, imperative verb •

الخواص اللغوية للفعل *Verbal Attributes* *Used*

- الجنس Gender
 - مذکر [masculine] M
 - مؤنث [feminine] F
 - حيادي [neuter] N
- العدد Number
 - مفرد [singular] Sg
 - مثنى [plural] Pl
 - جمع [dual] Du
- الخطاب Person
 - متكلم [first] 1
 - مخاطب [second] 2
 - غائب [third] 3
- الحالة Mood
 - رفع [indicative] I
 - نصب [subjunctive] S
 - جزم [jussive] J

الأدوات

- 1.1. Pr [prepositions] جر •
- 1.2. A [adverbial] ظرف •
- 1.3. C [conjunctions] عطف •
- 1.4. I [interjections] نداء •
- 1.5. E [exceptions] استثناء •
- 1.6 N [negatives] نفي •
- 1.7. A [answers] جواب •
- 1.8. X [explanations] تفسير •
- 1.9. S [subordinates] تمني •

أمثلة على الأدوات

- جر: في Prepositions “in”
- ظرف: سوف Adverbial particles “shall”
- عطف: و Conjunctions “and”
- نداء: يا Interjections “you”
- استثناء: سوى Exceptions “Except”
- نفي: لم Negatives “Not”
- جواب: أجل Answers “yes”
- تفسير: أي Explanations “that is”
- تمني: لو Subordinates “if”

اوسمة شيرين

Tag	Description of word category	Example (Arabic)
NCSgMNI	Singular, masculine, nominative, indefinite common noun	كتاب
NCSgMAI	Singular, masculine, accusative, indefinite common noun	كتابا
NCSgMGI	Singular, masculine, genitive, indefinite common noun	كتاب
NCSgMND	Singular, masculine, nominative, definite common noun	الكتاب
NCSgMAD	Singular, masculine, accusative, definite common noun	الكتاب
NCSgMGD	Singular, masculine, genitive, definite common noun	الكتاب
NCSgFNI	Singular, feminine, nominative, indefinite common noun	مدرسة
NCSgFAI	Singular, feminine, accusative, indefinite common noun	مدرستاً
NCSgFGI	Singular, feminine, genitive, indefinite common noun	مدرسة
NCSgFND	Singular, feminine, nominative, definite common noun	المدرسة

اوسمة شيرين

	Common noun	
NCDuFAD	Dual, feminine, accusative, definite common noun	المدرستين
NCDuFGD	Dual, feminine, genitive, definite common noun	المدرستين

NCPIMNI	Plural, masculine, nominative, indefinite common noun	كتبٌ - مسلمون
NCPIMAI	Plural, masculine, accusative, indefinite common noun	كتبًا - مسلمين
NCPIMGI	Plural, masculine, genitive, indefinite common noun	كتبٍ - مسلمين
NCPIMND	Plural, masculine, nominative, definite common noun	الكتبُ - المسلمون
NCPIMAD	Plural, masculine, accusative, definite common noun	الكتبَ - المسلمين
	Plural, masculine, nominative, definite	

اوسمة شيرين

NCPIFGD	Plural, feminine, genitive, definite common noun	المسلمات - المدارس
NP	Proper noun	شيرين - جده
NPrPSg1	First person, singular, neuter, personal pronoun	كتابي - ضرتني - أنا
NPrPSg2M	Second person, singular, masculine, personal pronoun	كتابك - أنت
NPrPSg2F	Second person, singular, feminine, personal pronoun	كتابك - أنت
NPrPSg3M	Third person, singular, masculine, personal pronoun	هو - كتابه
NPrPSg3F	Third person, singular, feminine, personal	هي - كتابها

اوسمة شيرين

NACPIFAD	Plural, feminine, accusative, definite adjective	السعيدات
NACPIFGD	Plural, feminine, genitive, definite adjective	السعيدات
VPSg1	First person, singular, neuter, perfect verb	كسرت
VPSg2M	Second person, singular, masculine, perfect verb	كسرت
VPSg2F	Second person, singular, feminine, perfect verb	كسرت
VPSg3M	Third person, singular, masculine, perfect verb	كسر
VPSg3F	Third person, singular, feminine, perfect verb	كسرت
VPDu2	Second person, dual, neuter, perfect verb	كسرتما

اوسمة شيرين

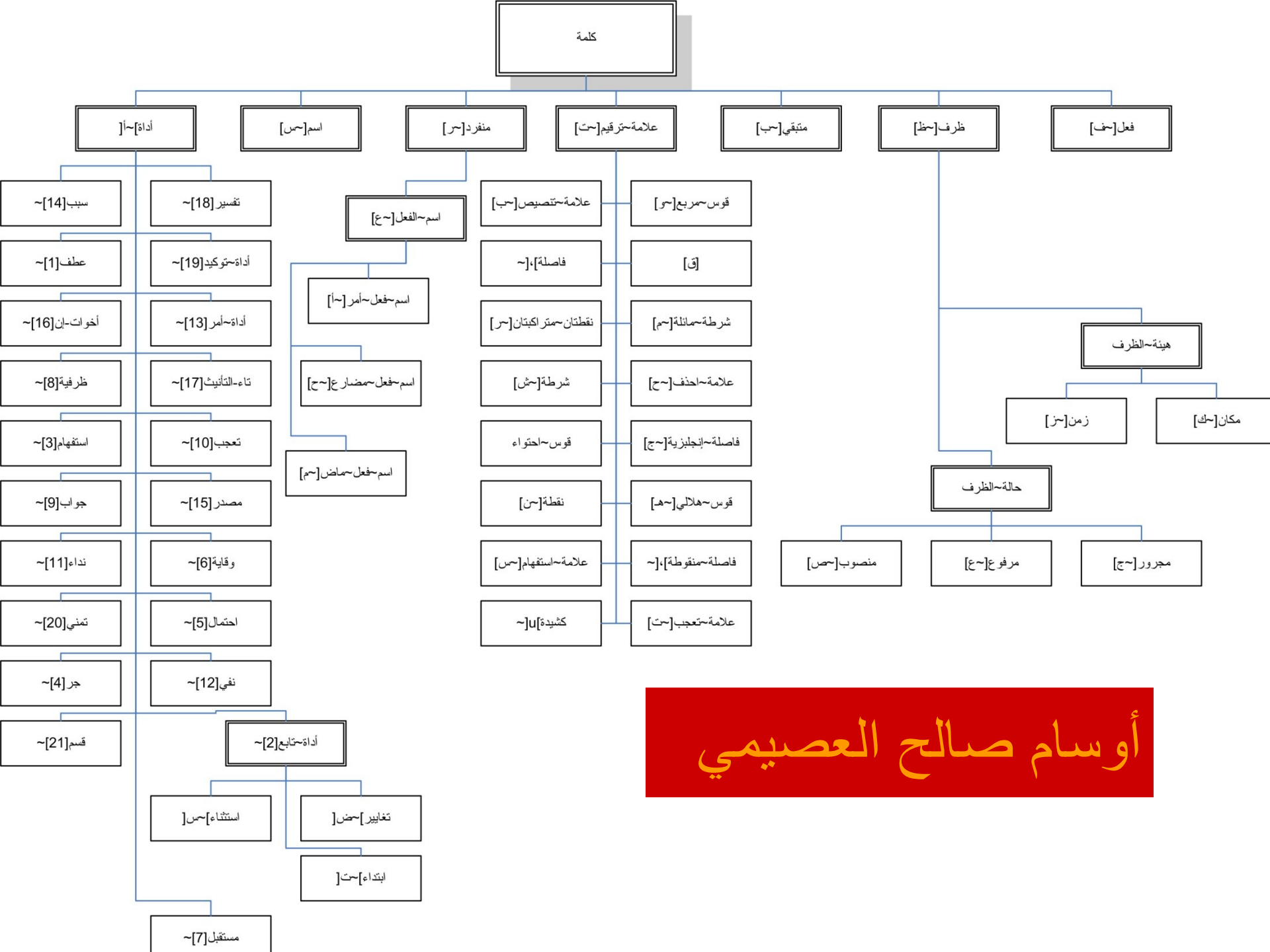
VPP12F	Second person, plural, feminine, perfect verb	كسرتن
VPP13M	Third person, plural, masculine, perfect verb	كسروا
VPP13F	Third person, plural, feminine, perfect verb	كسرن
VISg1I	First person, singular, neuter, indicative, imperfect verb	أكسر
VISg1S	First person, singular, neuter, subjunctive, imperfect verb	أكسر

اوسمة شيرين

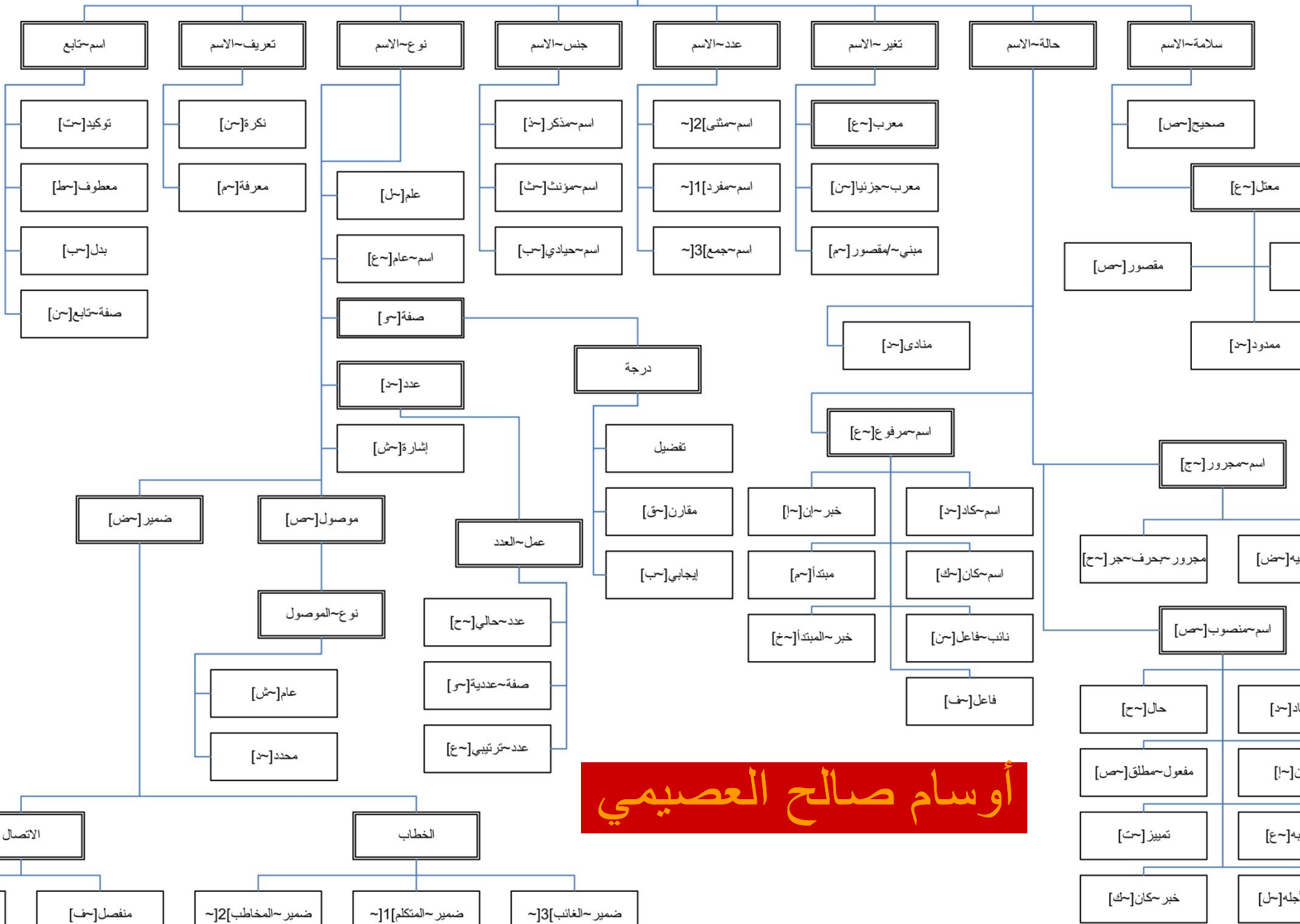
PN	Negative	لا - لعل
PW	Answers	اي
PX	Explanations	لو - ما
PS	Subordinates	رو - جور
RF	Residual, foreign	+
RM	Residual, mathematical	3
RN	Residual, number	الاثنين
RD	Residual, day of the week	محرم
Rmy	Residual, month of the year	ونس
RA	Residual, abbreviation	أل
RO	Residual, other	؟
PU	Punctuation	

بحث_VPSg3M خادم_NCSgMNI الحرمين_NCDuMAD الشريفين_NCDuMGD الملك_NCSgMND فهد_NP بن_
 NCSgMNI عبد_NCSgMAI العزيز_NCSgMAD آل_R سعود_NP برفية_NCSgFNI نهضة_NCSgFGI الى_PPr
 فخامة_NCSgFGI الرئيس_NCSgMGD الكسندر_RF كوستيفسكي_RF رئيس_NCSgMNI جمهورية_NCSgFGI
 بولندا_RF بمناسبة_NCSgFGI اليوم_NCSgMAD الوطني_NCSgMND لبلاد_NPrPSg3M NCPLFGI_PPr
 وأعرب_VPSg3M الملك_NCSgMND المفدى_NCSgMAD باسمه_NPrPSg3M NCSgMGI_PPr وياسم_
 NCSgMGI_PPr شعب_NCSgMGI وحكومة_NCSgFGI PC المملكة_NCSgFGD العربية_NCSgFGD السعودية_
 NCSgFGD عن_PPr اخلص_NCSgFNI النهائي_NCPLMND متعبا_NCSgMAI لفخامته_
 NCSgFGI_NPrPSg3M دوام_NCSgMNI الصحة_NCSgFGD والسعادة_NCSgFGD PC ولشعب_
 NCSgMGI_PPr بولندا_RF الصديق_NCSgMND الأدهار_NCSgMND الداعم_NCSgMN . PU

أوسام صالح العصيمي

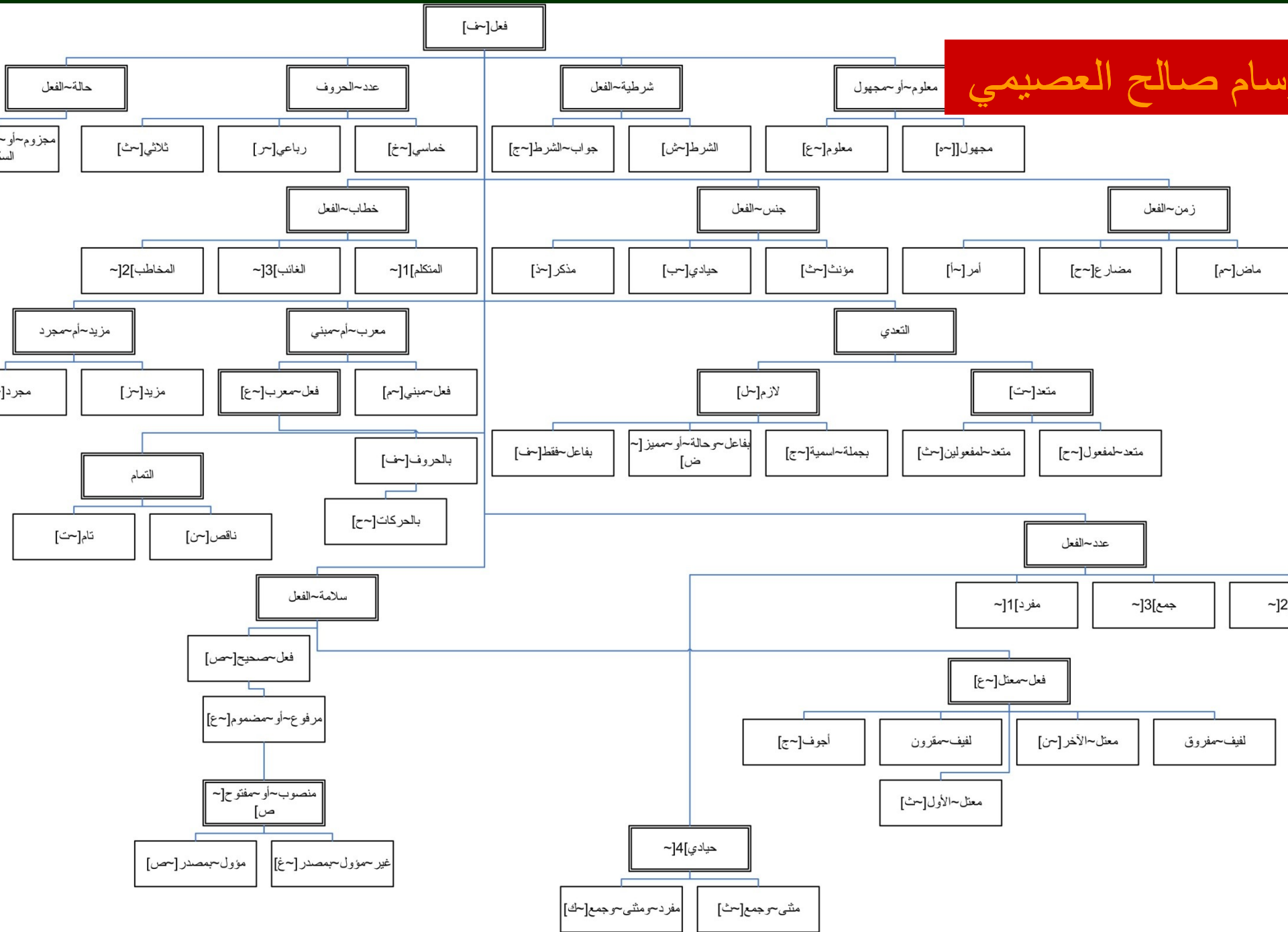


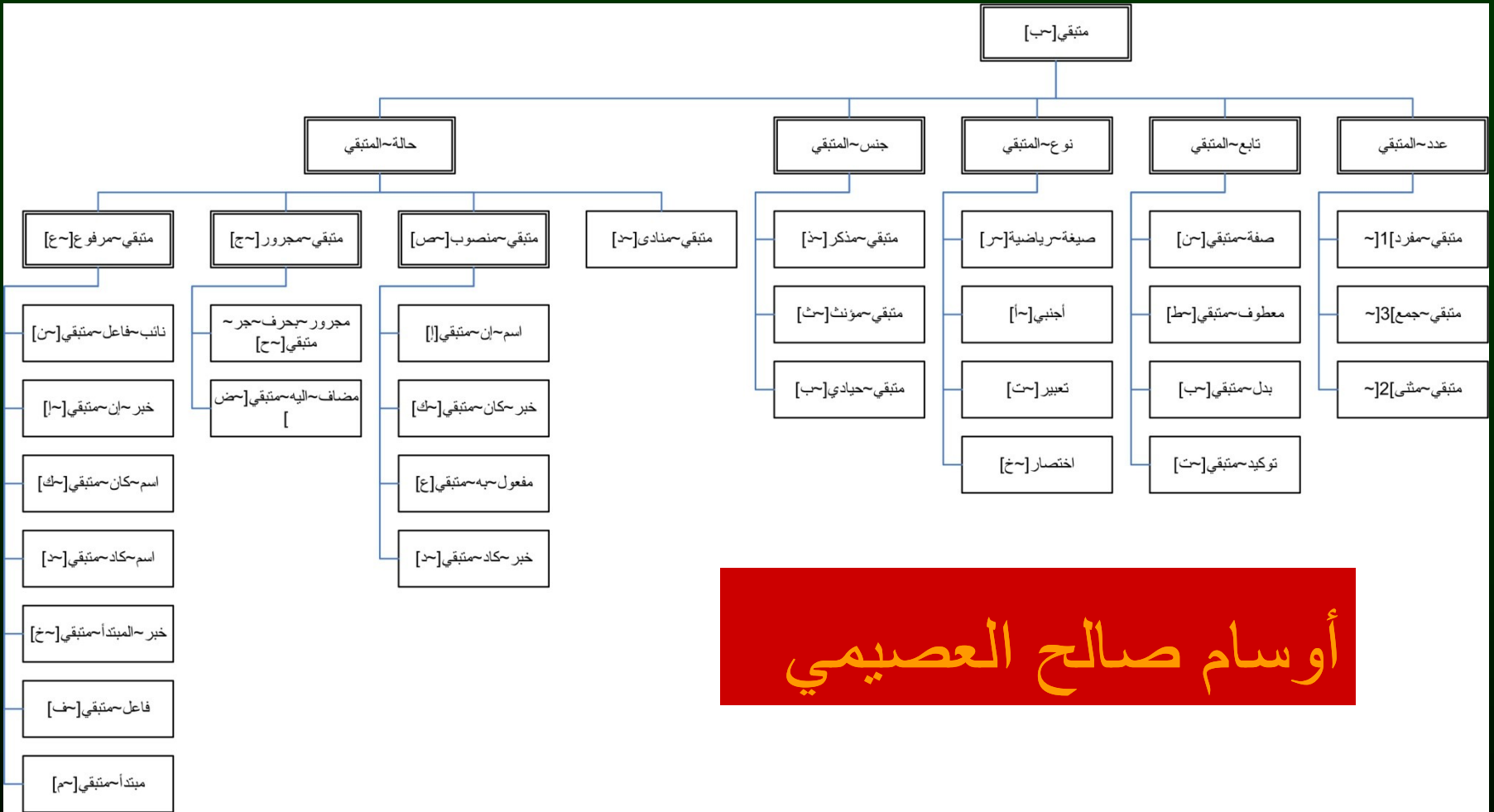
اسم [حس]



أوسام صالح العصيمي

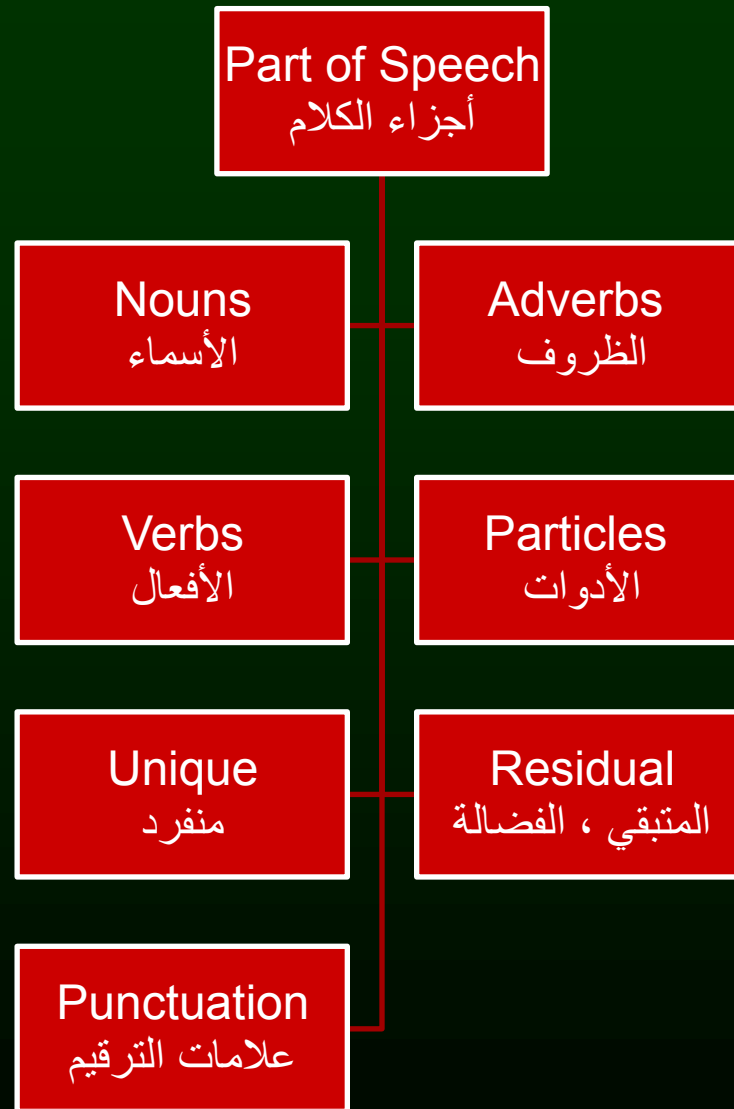
منفصل [حب] ضمير ~ المخاطب [2] ~ ضمير ~ المتكلم [1] ~ ضمير ~ الغائب [3] ~



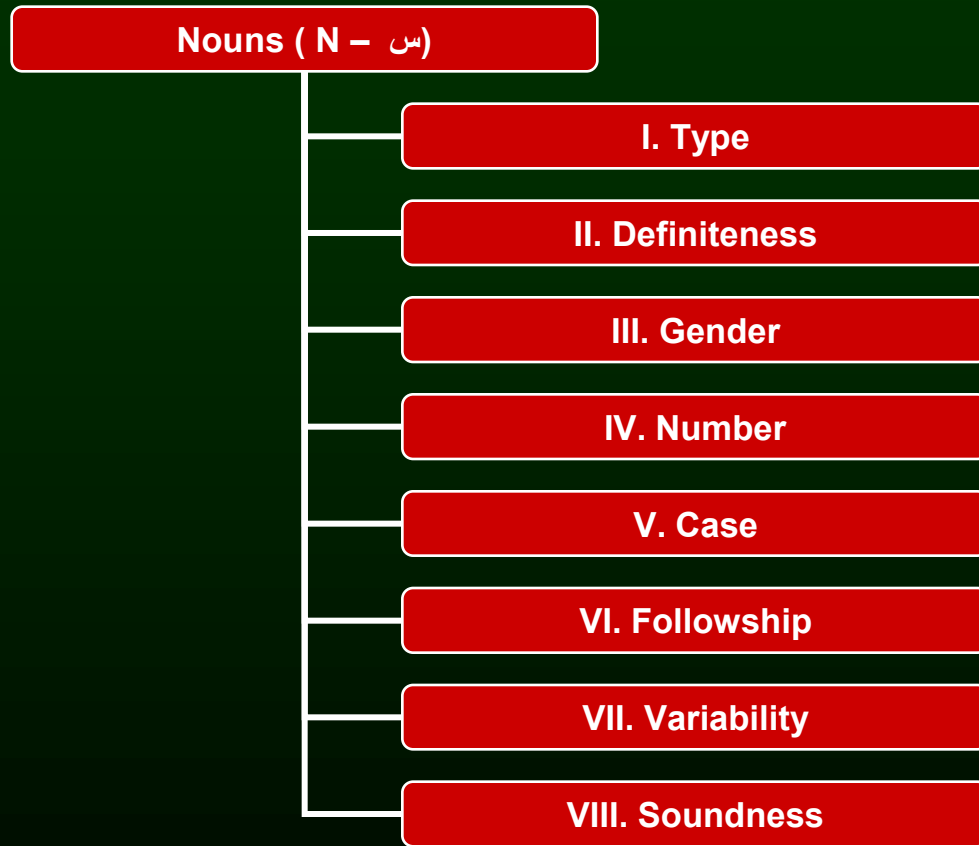


أوسام صالح العصيمي

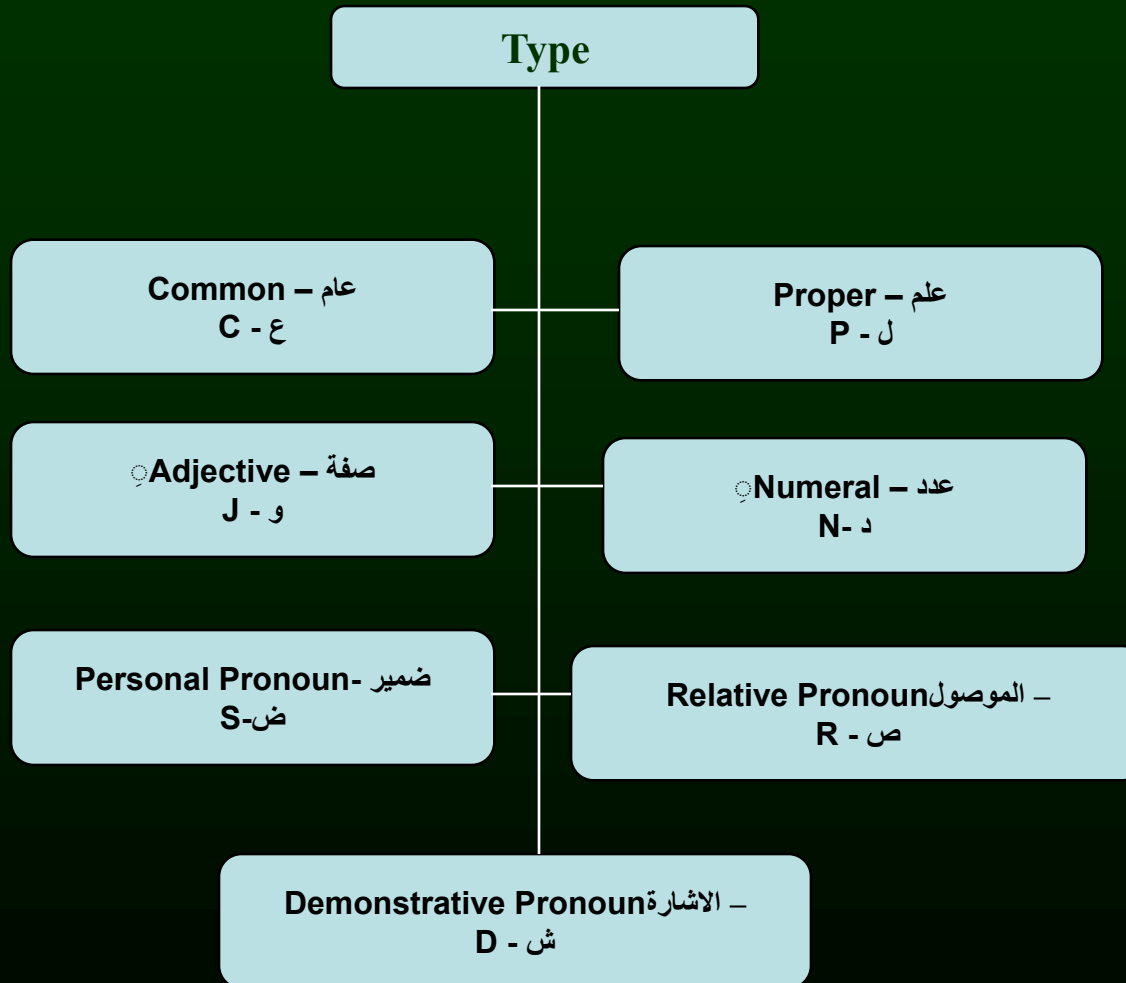
Parts of Speech



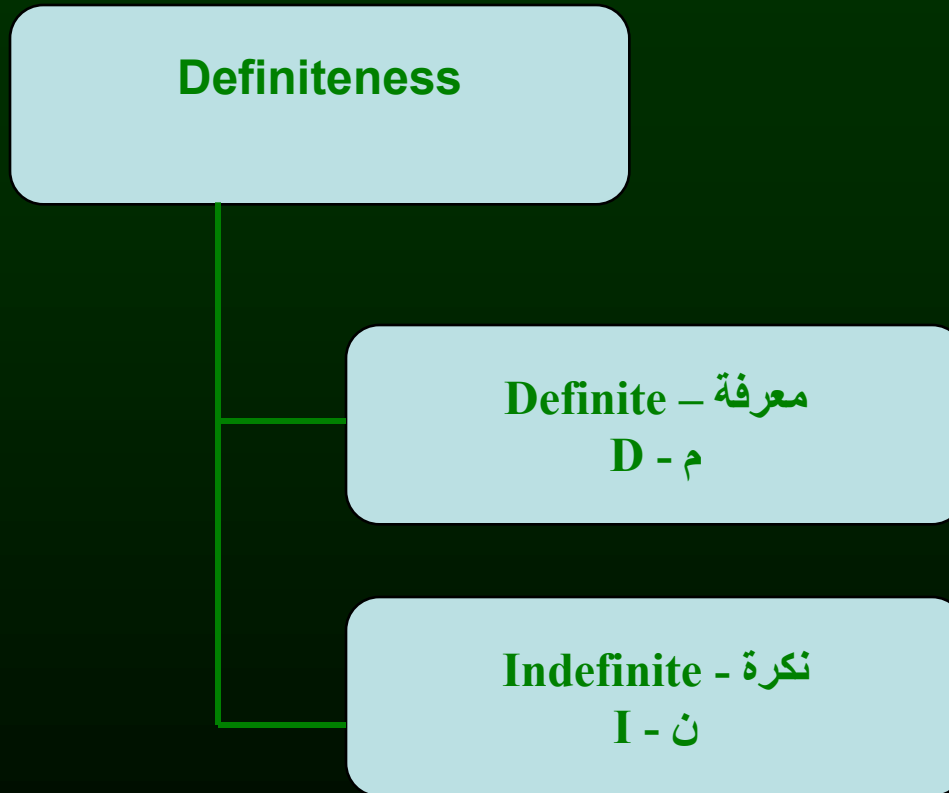
1. Noun



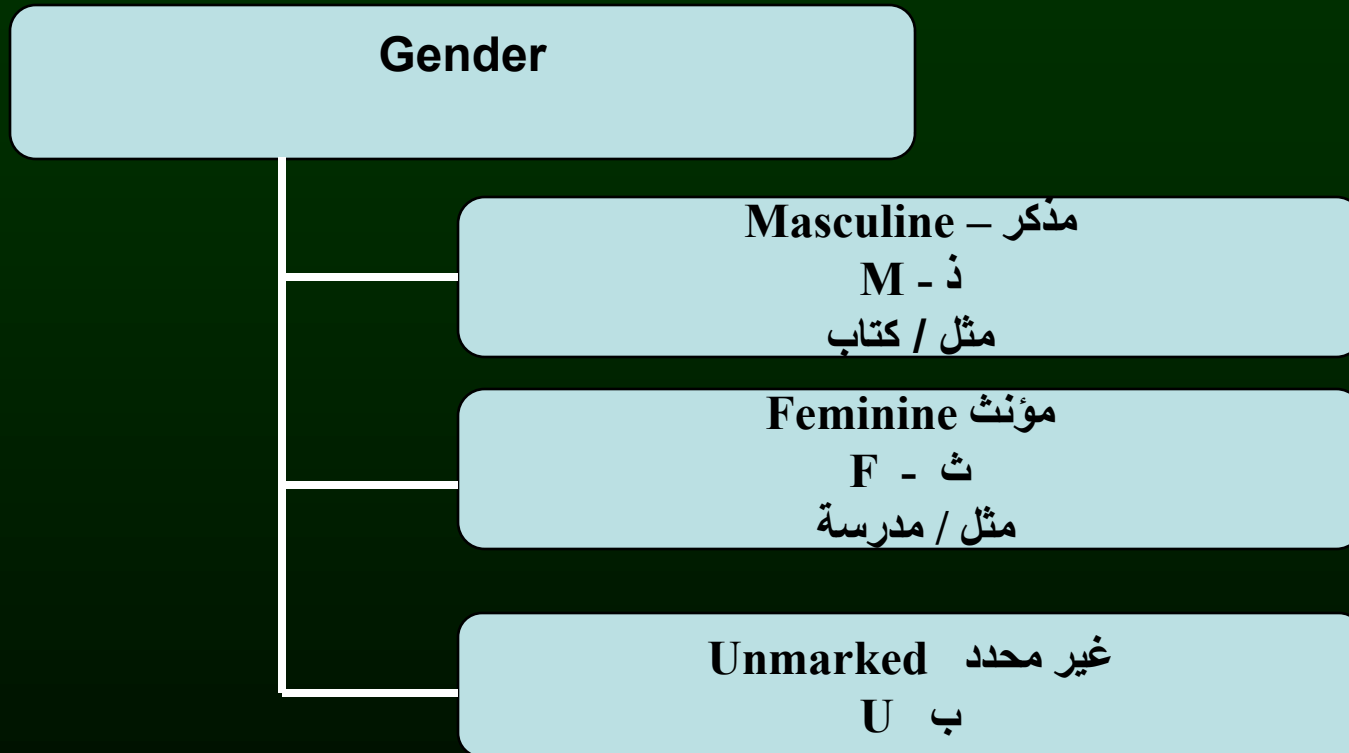
I. Type



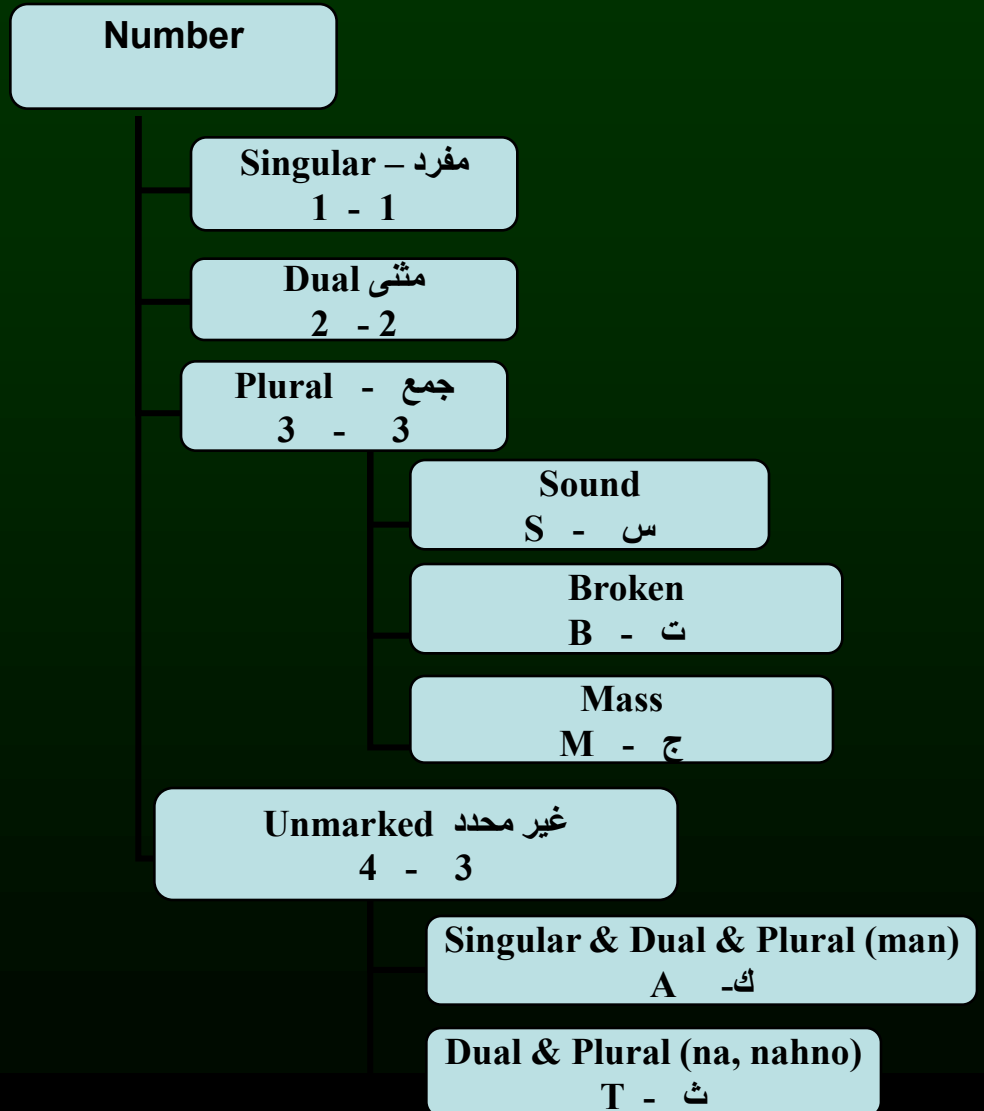
II. Definiteness



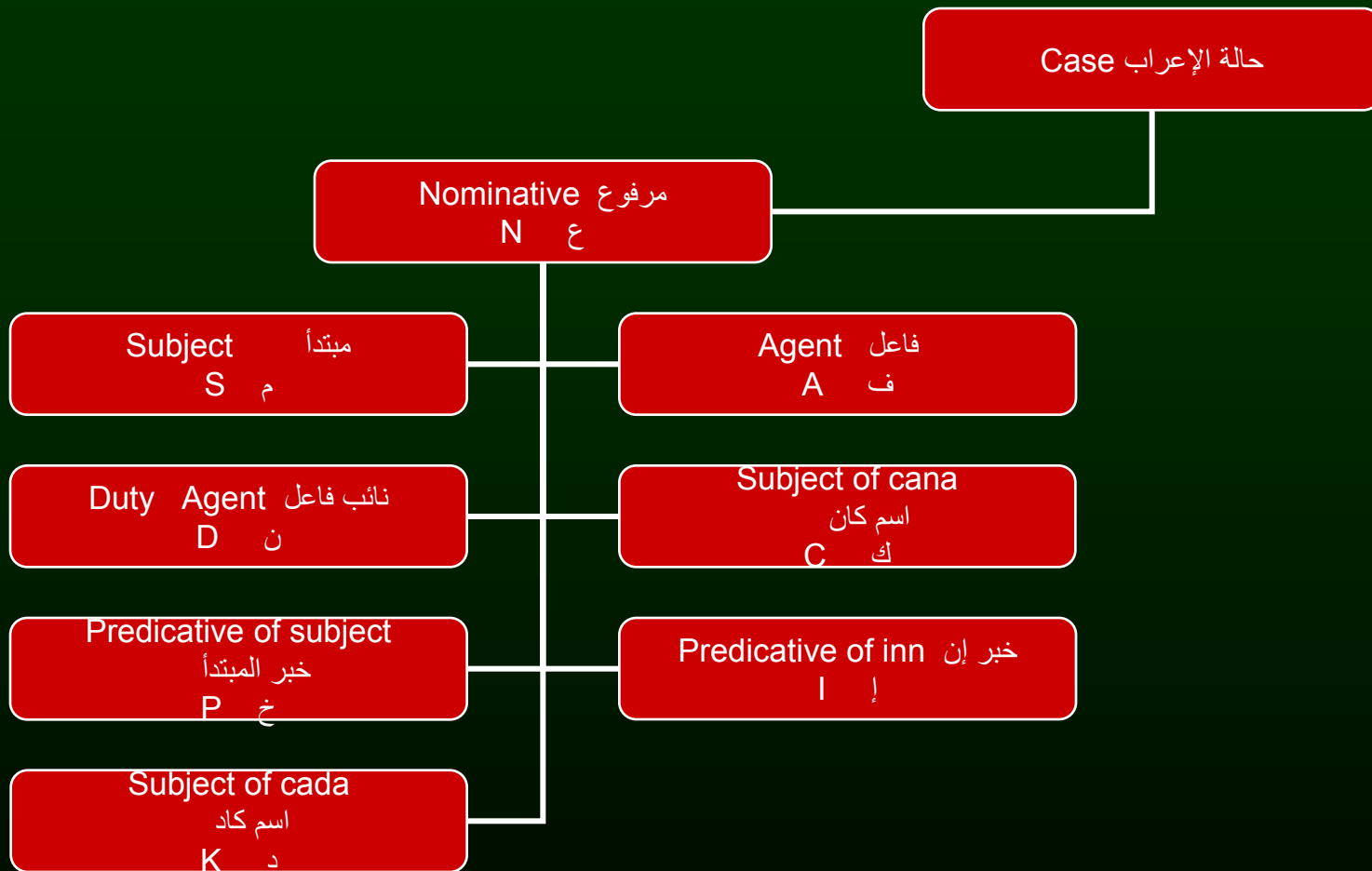
III. Gender



IV. Number



V. Case



حالة الإعراب Case

منصوب منصوب A ص

المفعول به Patient
P ع

خبر كان Predicative of cana
C ك

خبر كاد Predicative of cada
K د

اسم إن Subject of inn
I ا

حال State (manner)
S ح

تمييز Distinguative
D - ت

مفعول المطلق Infinitave
F ص

مفعول لأجله Cause
U ل

حالة الإعراب Case

مجرور Genitive
ج ج
G

مجرور بحروف الجر
Post – preposition
ح ح
P

مضاف إليه
Adjunct (post noun)
ض ض
A

حالة الإعراب Case

منادى
Vocative
د د
V

VI. Followship

Followship

Assertion

توكيد

A ت

Coordinated

معطوف

C ط

Attributive

صفة

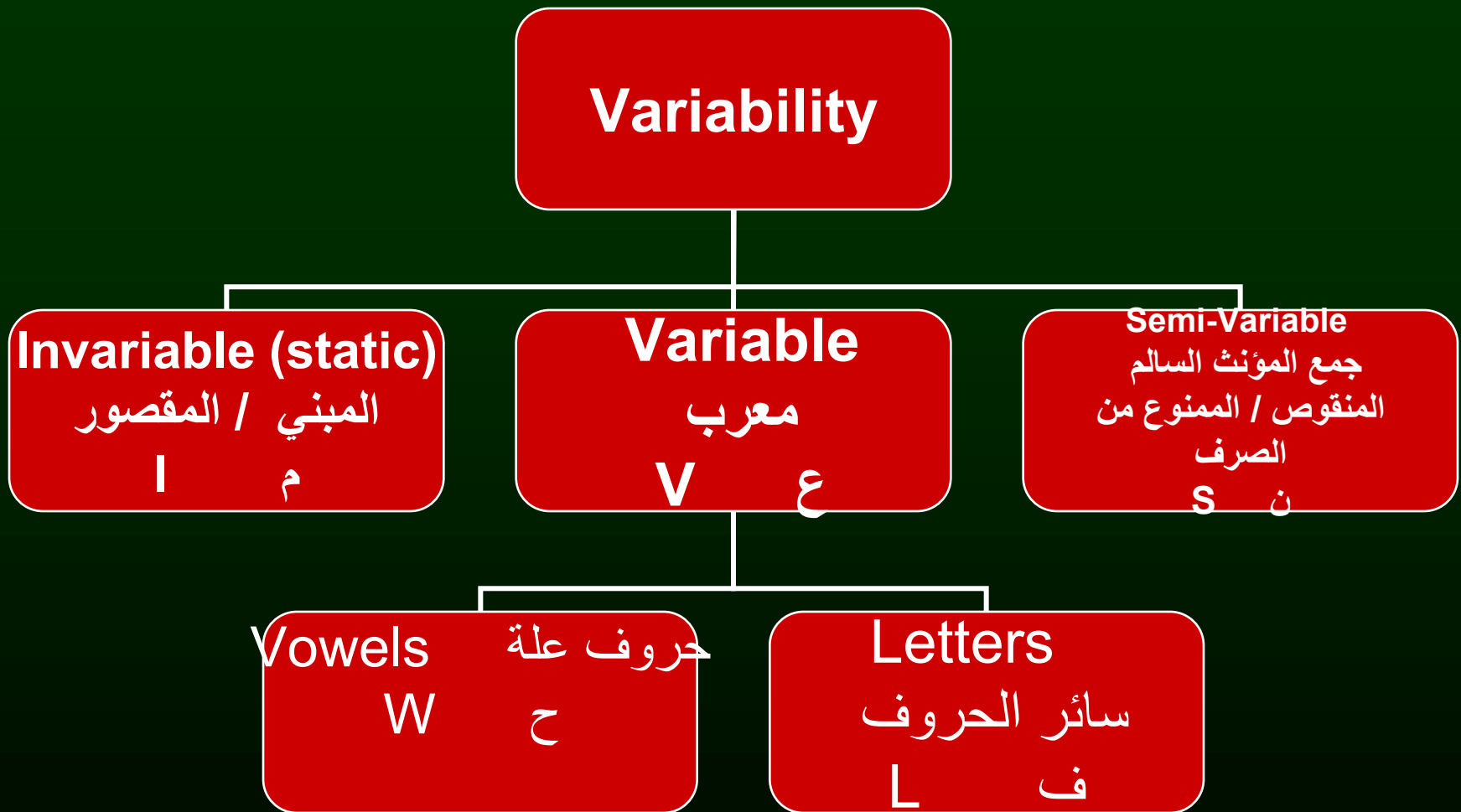
T ن

Substitute

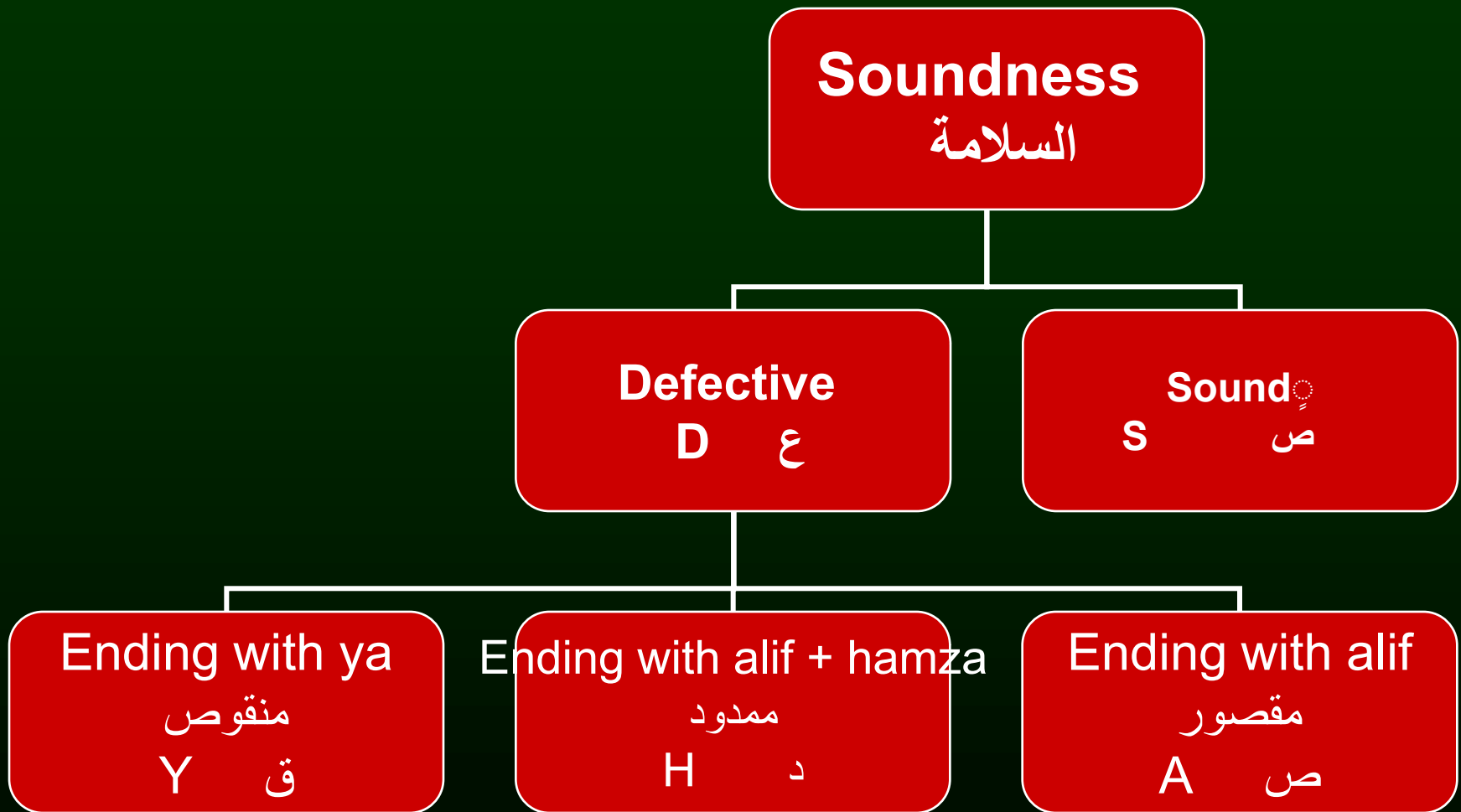
بدل

S ب

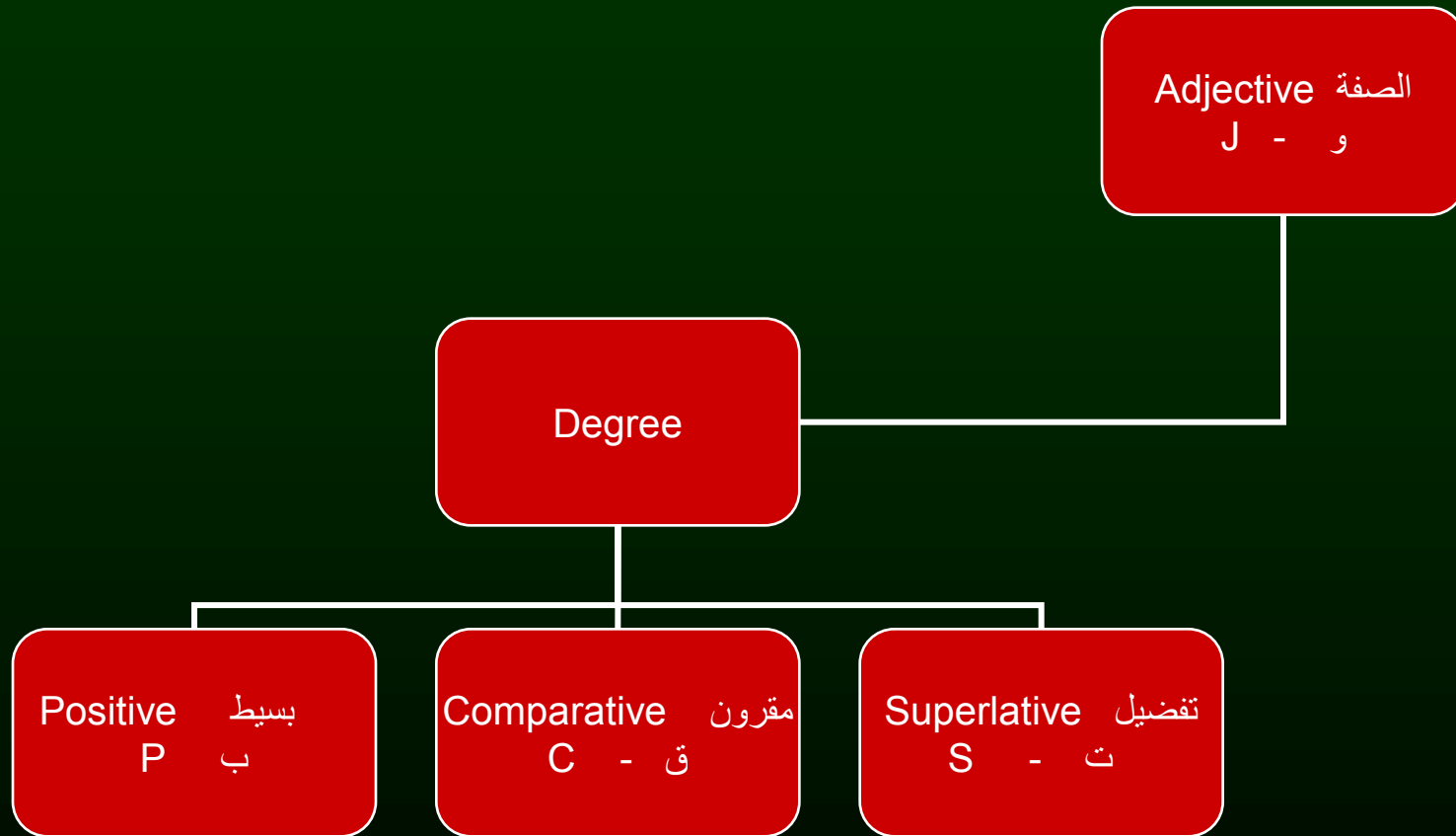
VII. Variability



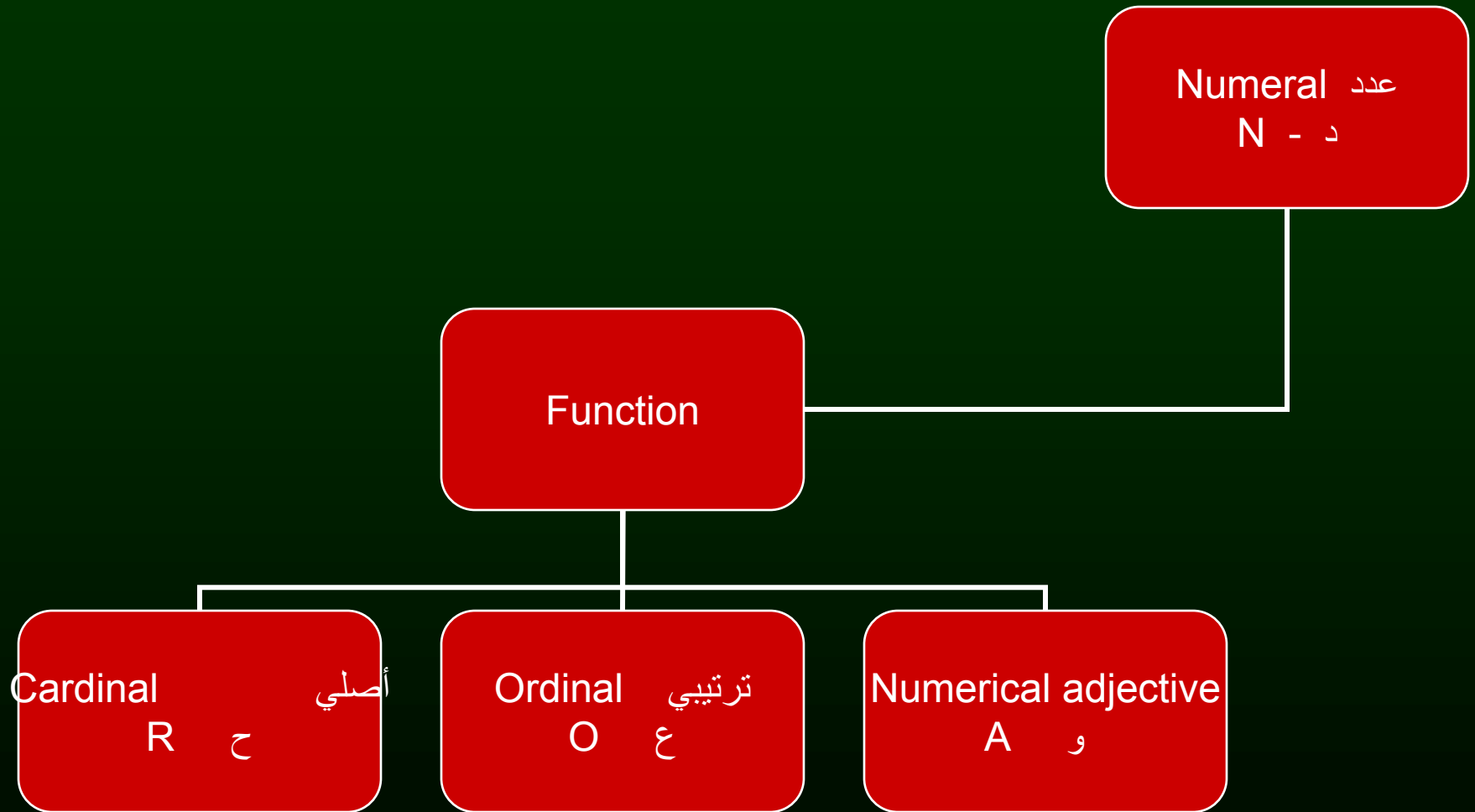
VII. Soundness



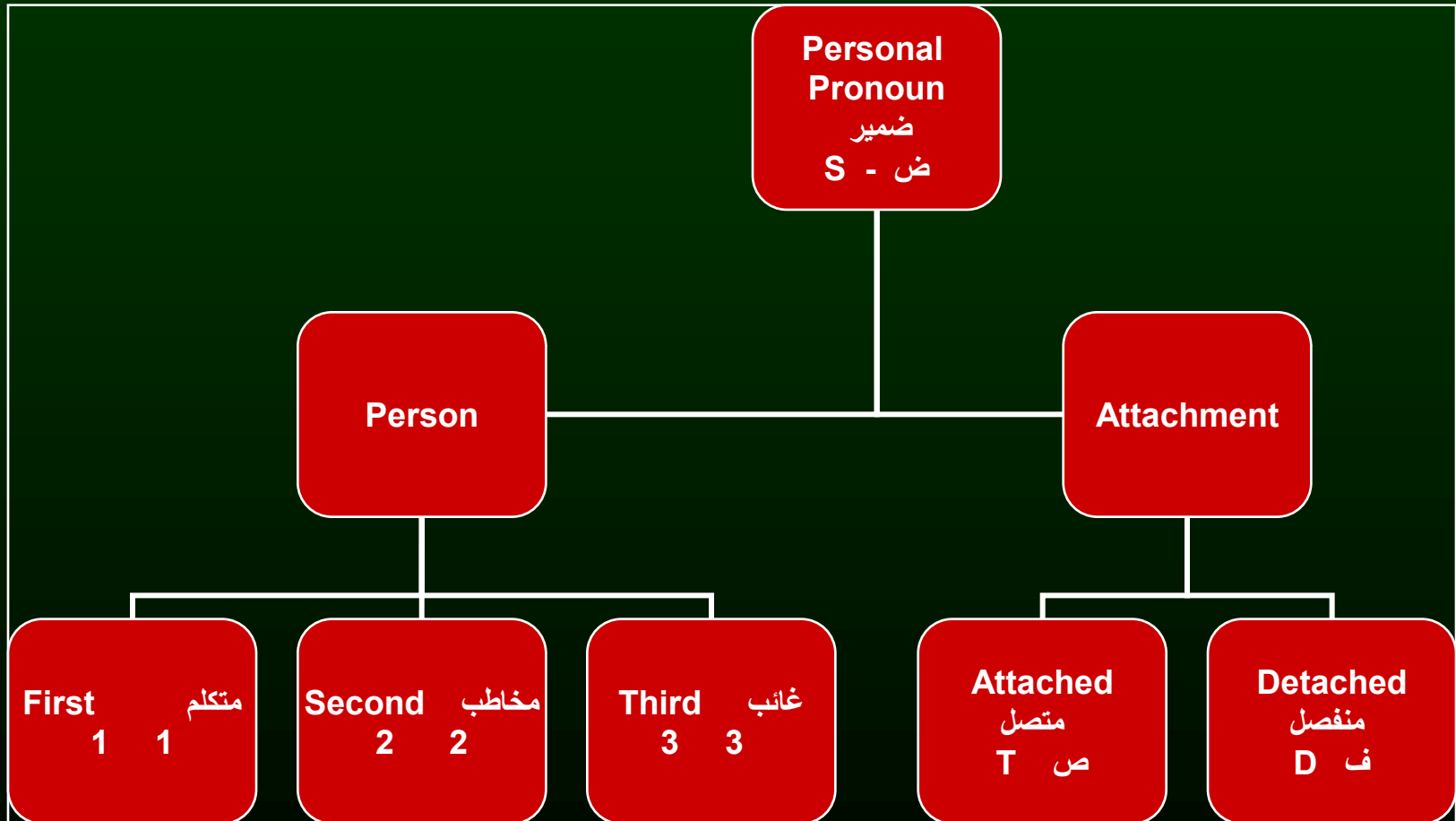
.. Type .. Adjective



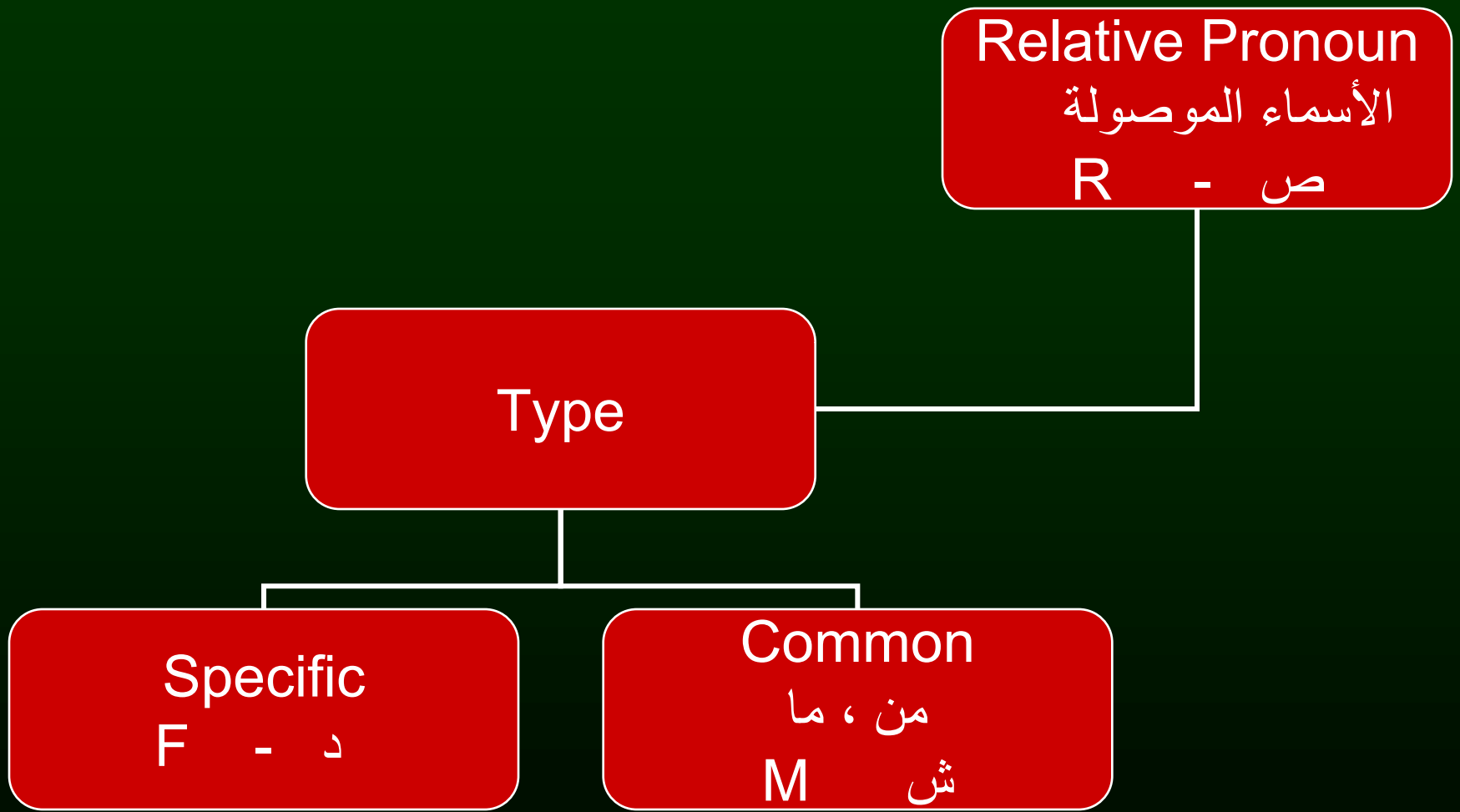
العدد Numeral Type



الضمائر ..Type ... Personal Pronoun



Type ... Relative Pronoun الأسماء الموصولة



Example

- فتحت المدرسة أبوابها
- المدرسة < Noun , Common, Definite ,
Feminine, Singular , Nominative (Agent)
, Φ , Variable- Vowels, Sound >

<N-C-D-F-1-NA-Φ-VW-S>

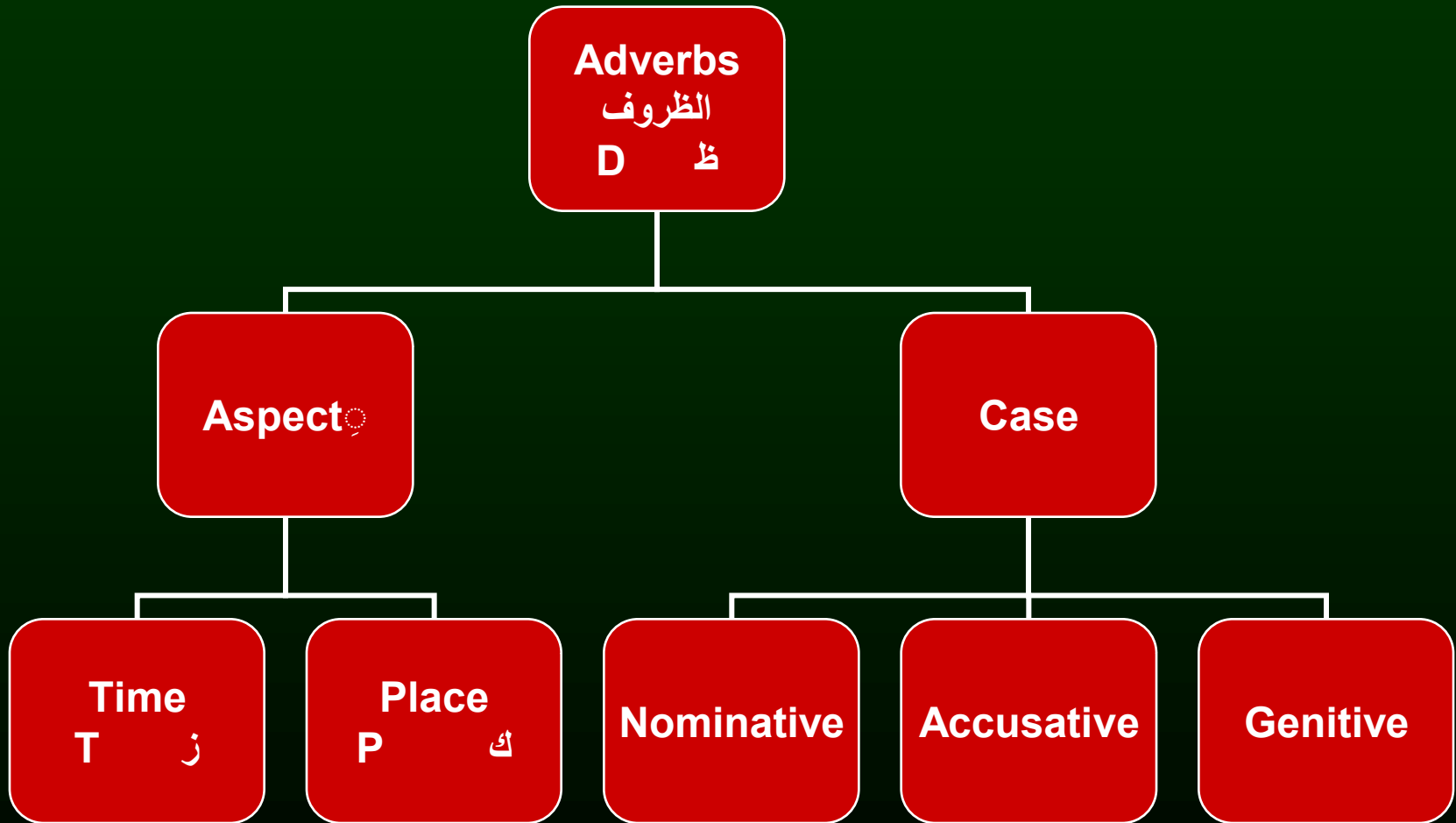
- أبواب

<N -C- I - F- 3B - AP- Φ - V W- S>

...

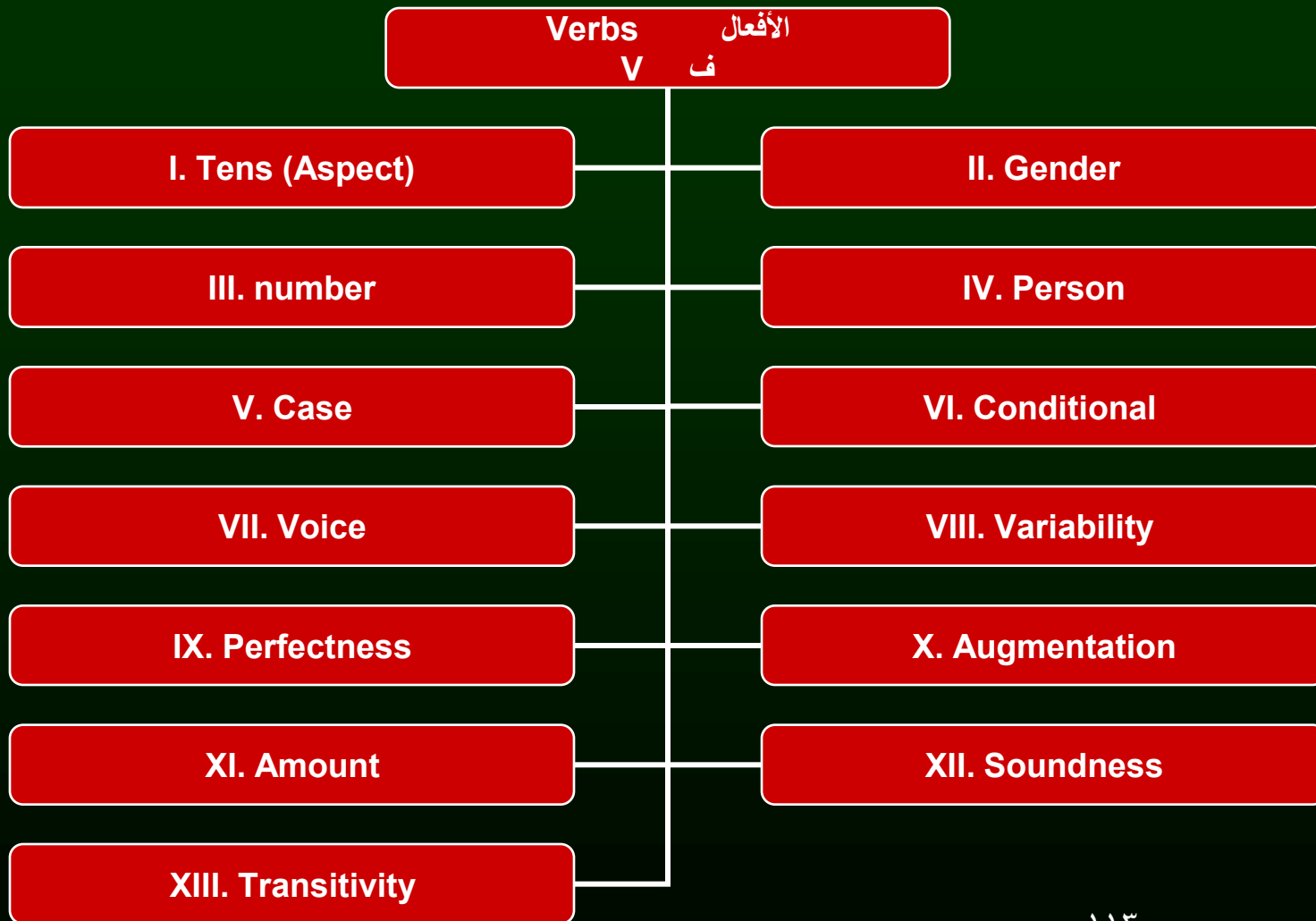
- ها < Noun , Personal Pronoun ,Definite , Feminine , Singular , Genitive post –noun (Adjunct) , Φ , Invariable (static) , Φ , Third , attached
- < N – S – D – F – 1 – GA – Φ – I – Φ – 3 – T >

2. Adverbs



- مثال
 - اتجه إلى **يمين** الشارع
 - **يمين** < Adverb , Place ,Genitive >
<D-P-G>

3. Verbs



I. Tense

1. Past (P- م)
2. Present (Durative \ Future).(R - ح)
3. Imperative (I - أ)

II. Gender

1. Masculine (M - ذ)
2. Feminine (F - ث)
3. Unmarked (U - ب)

III. Number

- Singular (1-1)
- Dual (2 -2)
- Plural (3-3)
- Unmarked (4-4)
 - Singular & Dual & Plural : verb of (man)
(A ك)
 - Dual & Plural : verb of (ma , nahno)
(T - ث)

IV. Person

1. First (1-1).
2. Second (2-2).
3. Third (3-3).

V. Case

1. Indicative (مرفوع أو مضموم) (N - ع)
2. Subjunctive (منصوب أو مفتوح) (A - ص)
 - Infinitive (مؤول بمصدر) (F - ص)
 - Non – Infinitive (N - غ)
3. Jussive (مجزوم أو مبني) (G - ج)

VI. Conditional

1. The condition (C - (ش))
2. The answer (A - (ج))

V. Voice

1. Active مبني للمعلوم (A - (ع))
2. Passive مبني للمجهول (P - (ج))

VIII. Variability

1. Invariable (Static) (I - م)
2. Variable (V - ع)
 - Vowels (W - ح).
 - Letters (L - ف)

IX. Perfectness

1. Perfect تام (P - ت)
2. Imperfect ناقص (Can and cada) (I – ن)

X. Augmentation

- Augmentation (A - زائد)
- Non – Augmentation (N - ج)

XI. Amount

- Trilateral (T - ثلاثي)
- Quadric-Literal (Q- رباعي).
- Penta – Literal (P-خماسي).

XII. Soundness

- Defective (D - ع):
 - Initial (I - في البداية) (I - ث)
 - مثل / يبس
 - Hollow (Meddle) (H-خ)
 - مثل / ضاع
 - Last (L - ن)
 - مثل / نسي
 - Initial + last (T - ف)
 - مثل / وعى
 - Hollow + Last (O - ق)
 - مثل / حوى
- Sound (S - ص)

XIII. Transitivity

- Transitive متعدي (T - ت)
 - One Patient مفعول واحد (O - ح)
 - مثل / أكل
 - Two Patient مفعولين (T - ث)
 - مثل / أعطى
- Intransitive لازم (I - ل)
 - Agent only فاعل فقط (A - ف)
 - Agent + State or Distinguitor (S - ض)
 - Nominal Sentence (N - ج)

- مثال /

- نامت الطفلة

نامت <Verb , Past , Feminine , Singular , Third ,
Subjunctive non infinitive , Φ , Active ,
Invariable (static) , Perfect , Augmented ,
Trilateral ,Sound, intransitive Agent only >

<V – P – F – 1- 3 – A N – Φ – A – I – P- A – T – S
– I A >

4. Particles أدوات

P أ

- Coordinating (1 - 1) حروف العطف
و / ف / ثم / أو / حتى لو / حتى / حتى ولو
- Subordinating (2 - 2)
 - Contrast (ض - c) (لكن / لكنّ / بل)
 - Exception (س - E) (إلا / غير / سوى / ما عدا / عدا)
 - Initial (ت - I) (و / ف الإبتدائية)
- Interrogative (3 - 3) (هل / أ)
- Preposition (4 - 4) (حروف الجر)

- Possibility (قد قبل المضارع) (5).
- Protection (نون الوقاية) (6).
- Future (سـ / سوف) (7).
- Conditional (ما / مهما / إذا / أيّ) (8).
- Answer (نعم / أجل / لا / إذن) (9).
- Exclamation (ما أو إنفعال) (10).
- 11. Interjection/Intrognative (يا / ايها / أيتها) (11).



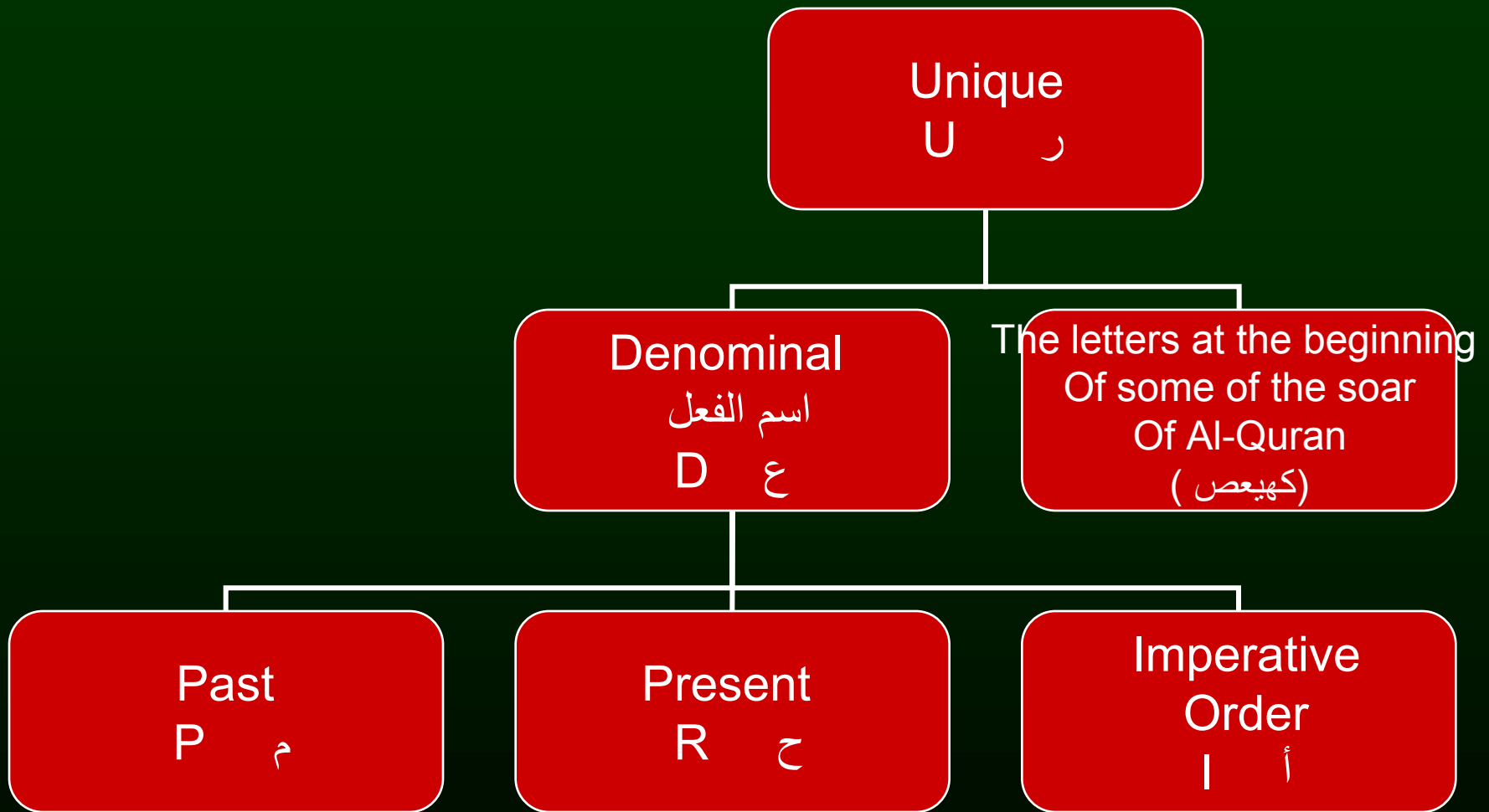
- Negative (ليس / لا / لم / لن) (12).
- Imperative (Order) (ـ) (13).
- Cause (حتى / لأجل / من أجل / لكي) (14).
- Gerund (أن / أنّ) (15).
- Deporticle (اخوات إن) (16).
- تاء التانيث (17).
- Explanation (أي) (18).

- Assertion (إنَّ / أنَّ / نونا التوكيد / قد و لقد على الماضي) (19).
- Wishing (لعل / ليت) (20).
- Swearness (و القسم) (21).

- مثال /

- نامت الطفلة
- < Particles , ta of Femininity >
<P - 17>.
- قال تعالى {وَالْمُرْسَلَاتِ عُرْفًا}
- < Particles , Swearness > و
<P - 21>

5. Unique (U - ر)



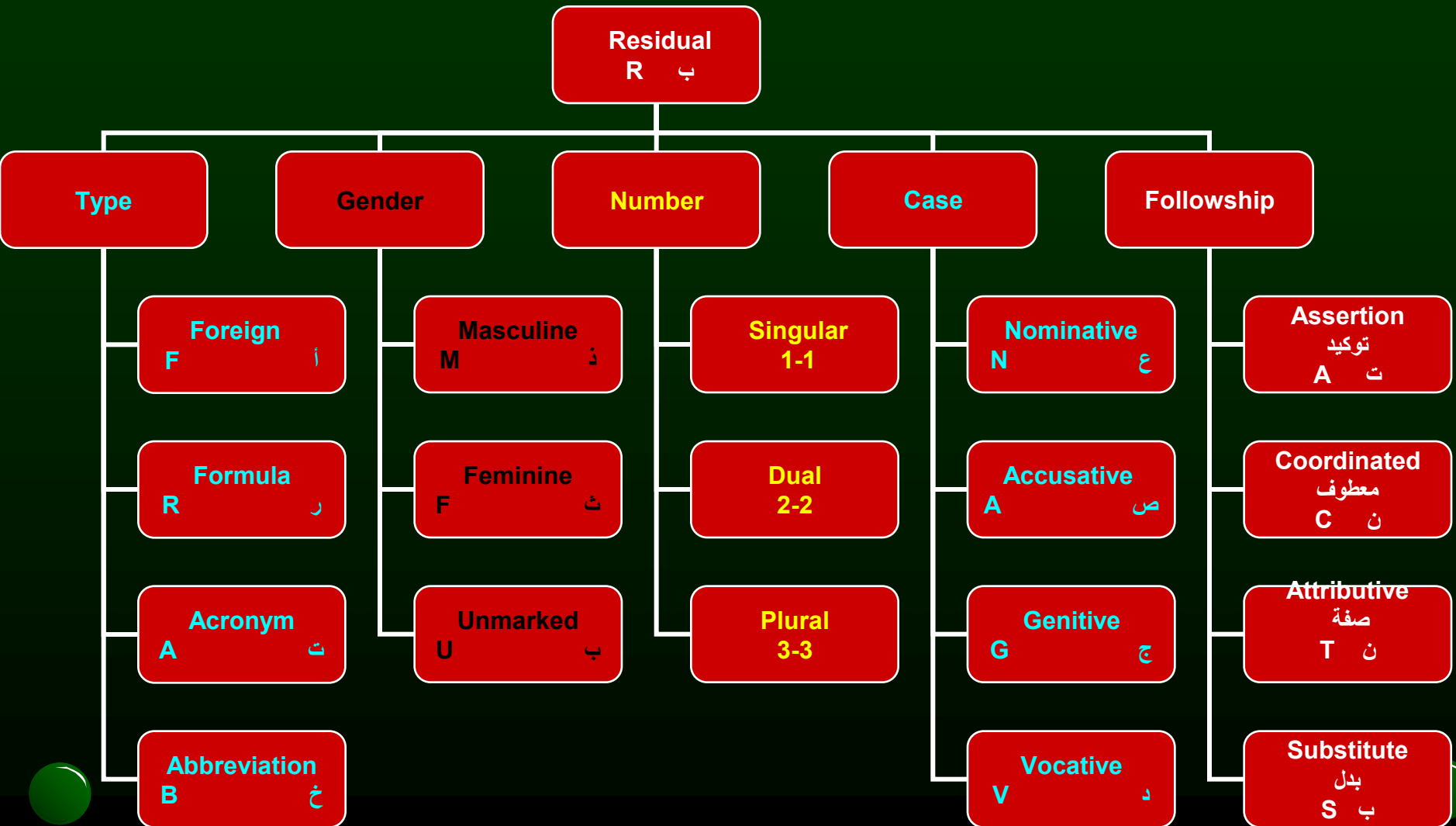
- مثال

قال تعالى { حم (١) تَنْزِيلُ الْكِتَابِ مِنْ اللَّهِ الْعَزِيزِ الْعَلِيمِ (٢) }

- <حم Unique , The litters at the beginning Of some of the soar Of Al-Quran >

- <U - L>

6. Residual



• مثال /

تلعب شركة أرامكو دورا كبيرا في اقتصاد الدولة

• أرامكو <Residual , Foreign , Feminine ,
Singular , Genitive Adjunct (post noun) >
<R – F – F – 1 – G A>

7. Punctuation (ت - ح)

- ? Question Mark (س- Q).
- ! Exclamation Mark (ت - X).
- ... Ellipsis (ح - E).
- . Full Stop (ن- F).
- ‘ Comma (ف- C).
- ; Dotted Comma (ق- D).
- - Hyphen (ش- H).

- - Interspersion Marks (I- ع).
- , The English Comma (G- ج).
- , , Interspersion Marks (R- ق).
- () Brackets (B- ه).
- " Quotation Marks (U- ب).
- : Colon (O- ر).
- [] Square Brackets (S- و).
- { }
- / slash (L- م).

- مثال /

- فتحت المدرسة أبوابها .

- <Punctuation , Full Stop>

<C - F>

To Part 2: Arabic POS

• السلام عليكم ورحمة الله