Virtual Distances between Languages

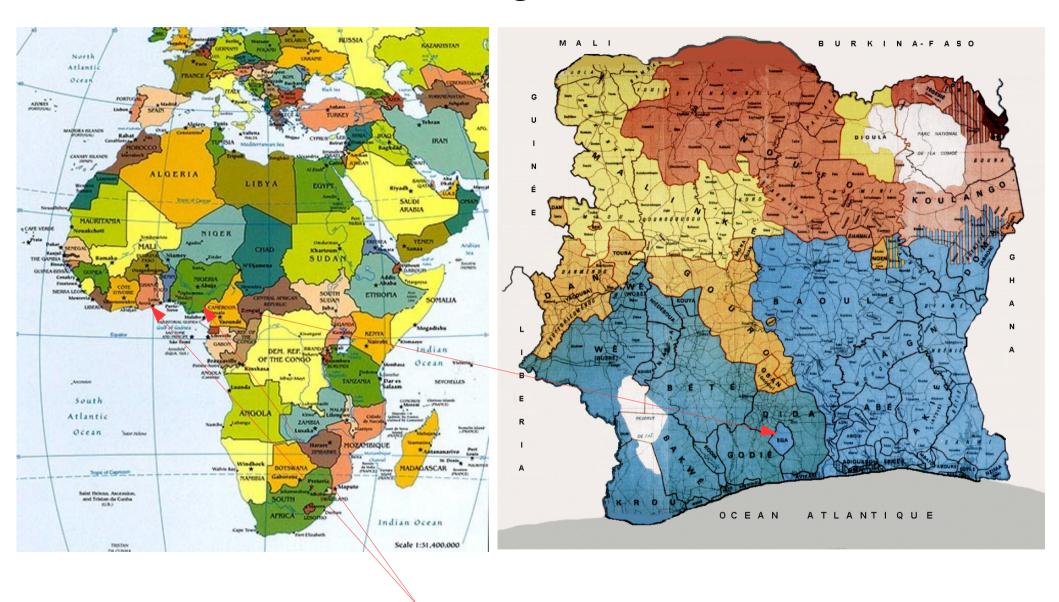
Methods from Dialectometry and Stylometry in Digital Humanities

Dafydd Gibbon

Universität Bielefeld Germany

Bielefeld-Abidjan Cooperation 2015-11-16

Background



My work in teaching and in the field in West Africa has been mainly mainly in Côte d'Ivoire and Nigeria.

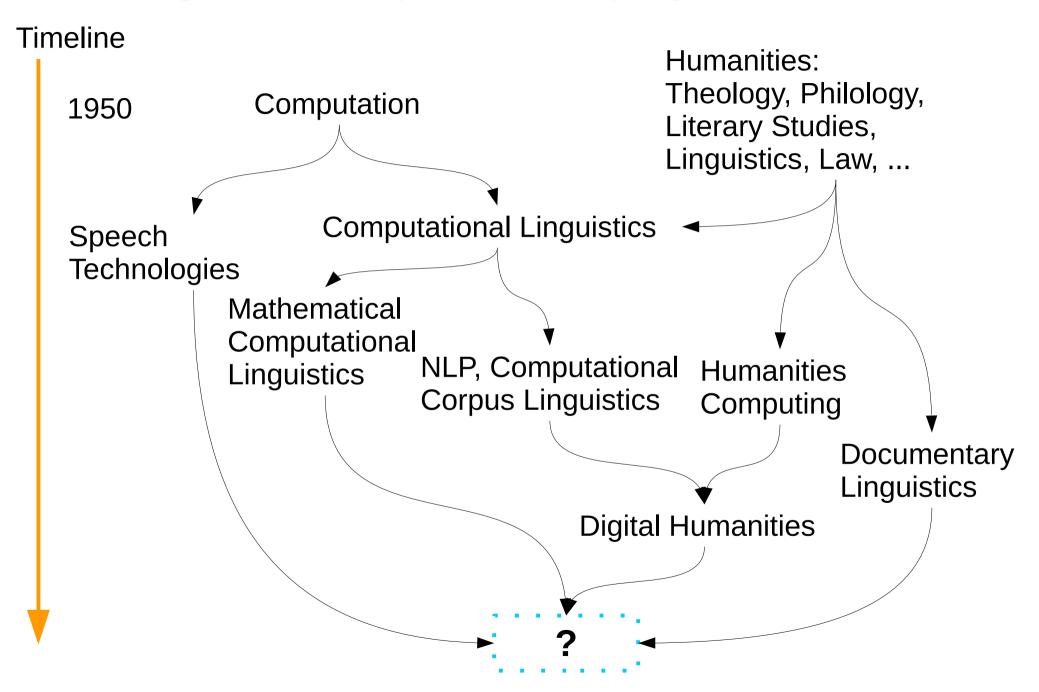
From fieldwork to Digital Humanities



Research goals:

Analysis of oral literature in a Digital Humanities framework. Phonetic analysis; sound-gesture synchronisation in speech and music. Visualisation of differences between languages as 'virtual distances'.

A personal view of the ancestry of Digital Humanities



Background in Linguistics

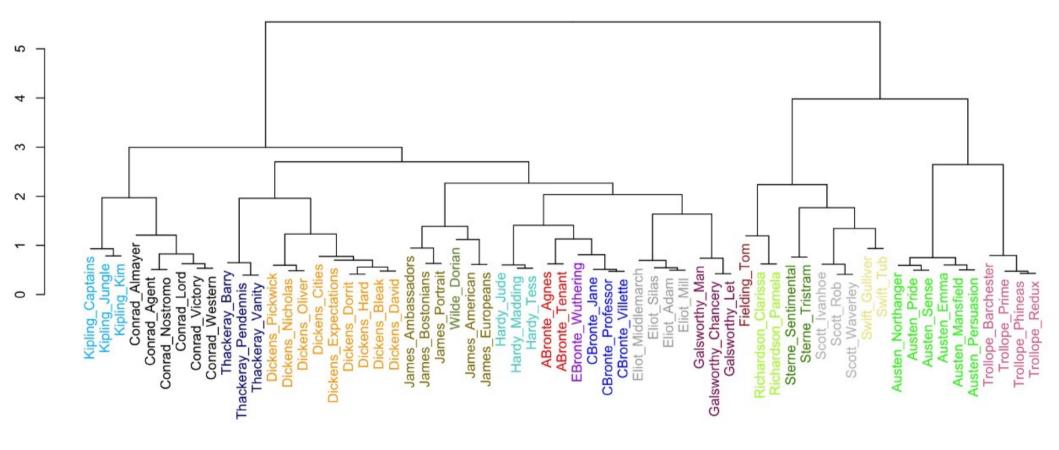
Lexicostatistics

- Morris Swadesh (Swadesh wordlist)
 - create wordlists for each language (interviews, corpus)
 - determine cognate words (related in form and meaning)
 - determine pairwise similarities → triangular distance table
 - create family tree using the table
 - Problems: accidental similarity, borrowing, taboo, ...

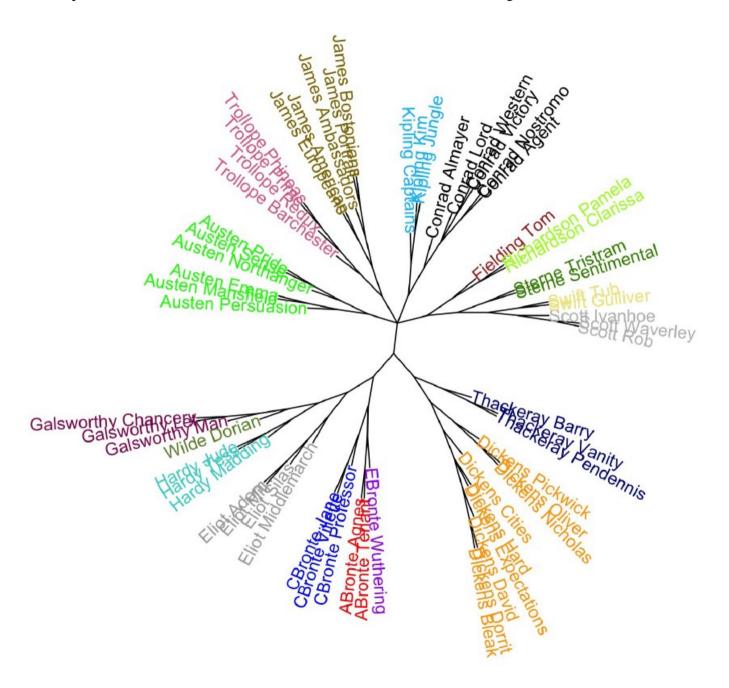
Dialectometry

- Two major centres in Europe:
 - Rijksuniversiteit Groningen
 - Universität Salzburg
- Similar methodology in Digital Humanities:
 - Stylometry: authorship attribution, style typology
 - MFW: most frequent lexical words, most frequent grammatical words

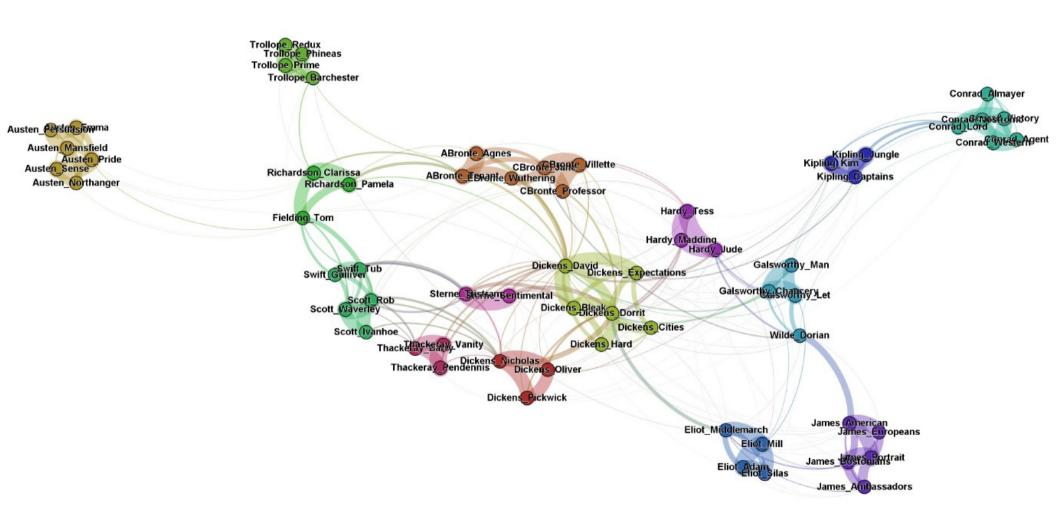
Stylometric visualisation – Maciej Eder, Kraków



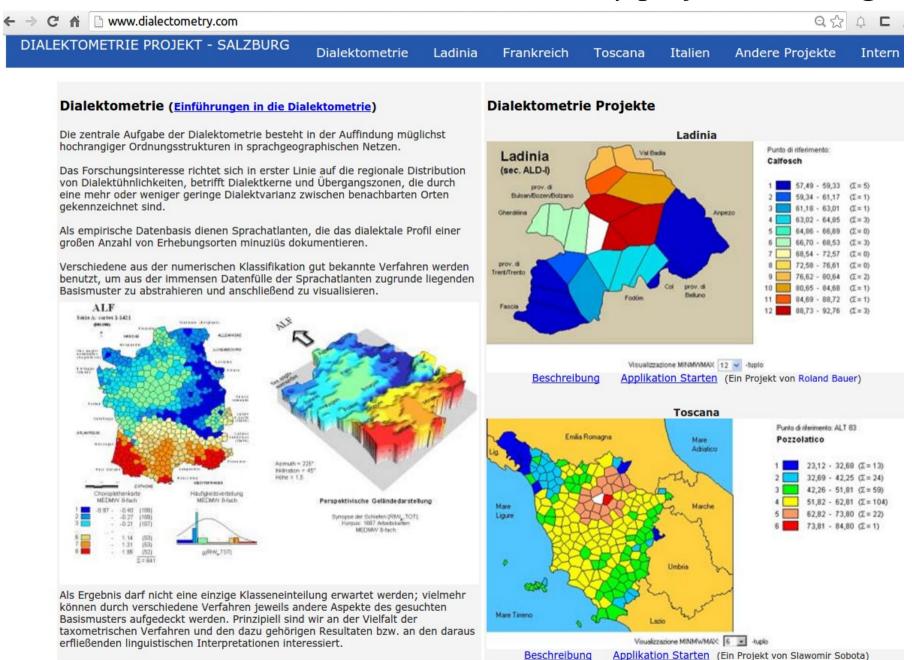
Stylometric visualisation – Maciej Eder, Kraków



Stylometric visualisation – Maciej Eder, Kraków



Dialectometric visualisation – Dialectometry project, Salzburg



[Universität Salzburg, Fachbereich Romanistik, Erzabt-Klotz-Str. 1, A-5020 Salzburg, E-Mail: webmaster(at)dialectometry(dot)com

Language similarity visualisation – Gerhard Jäger, Tübingen

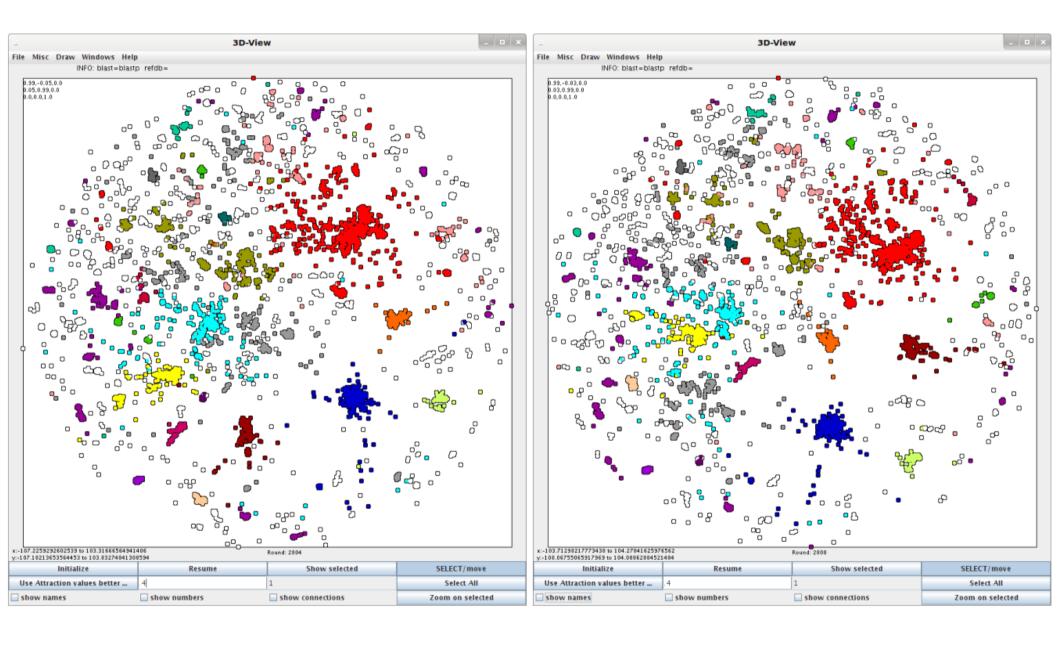


Figure 3: Languages of the world

Language typology - visualisation of 'virtual distances'

Project: comparison of Kru languages

Hypothesis:

The geographical distances between the Kru languages are reflected in the differences of their consonant systems.

Method:

Data mining with legacy language atlases

Results:

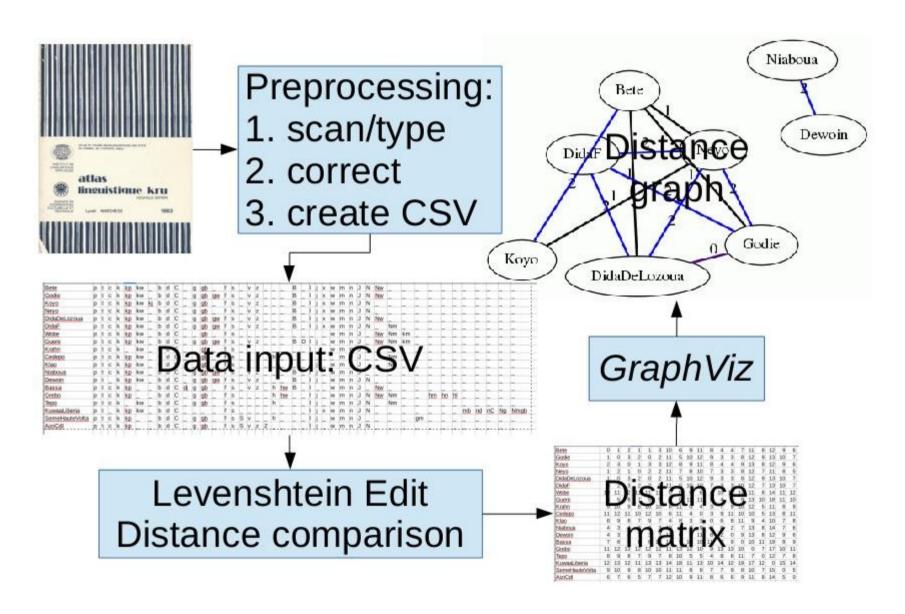
Coming up

Here:

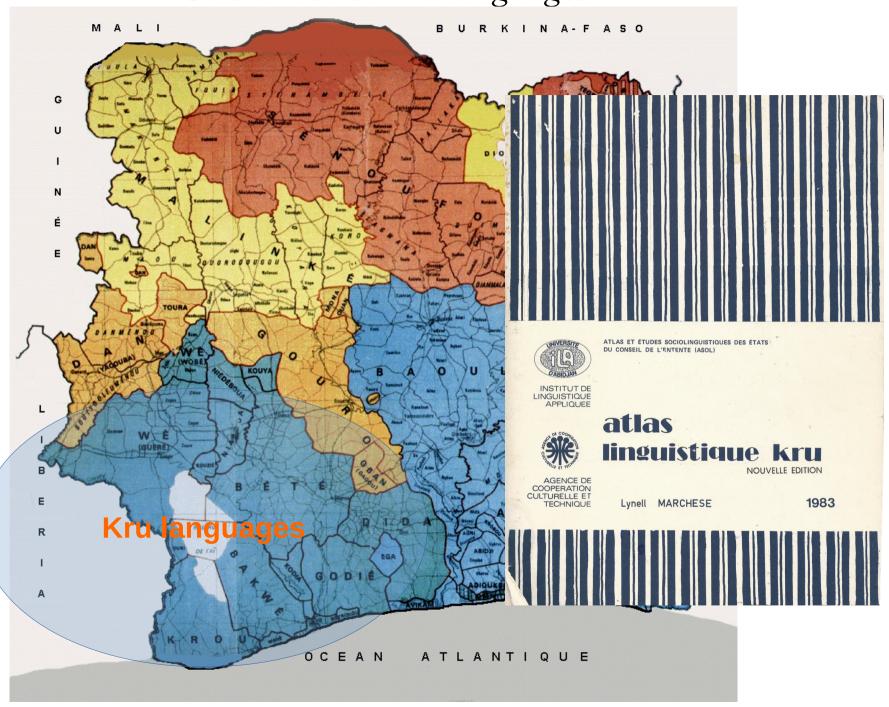
Visualisation of relations induced from the legacy data for 19 Kru languages

(of the 39 in the Ethnologue database of language metadata)

Visualisation workflow:

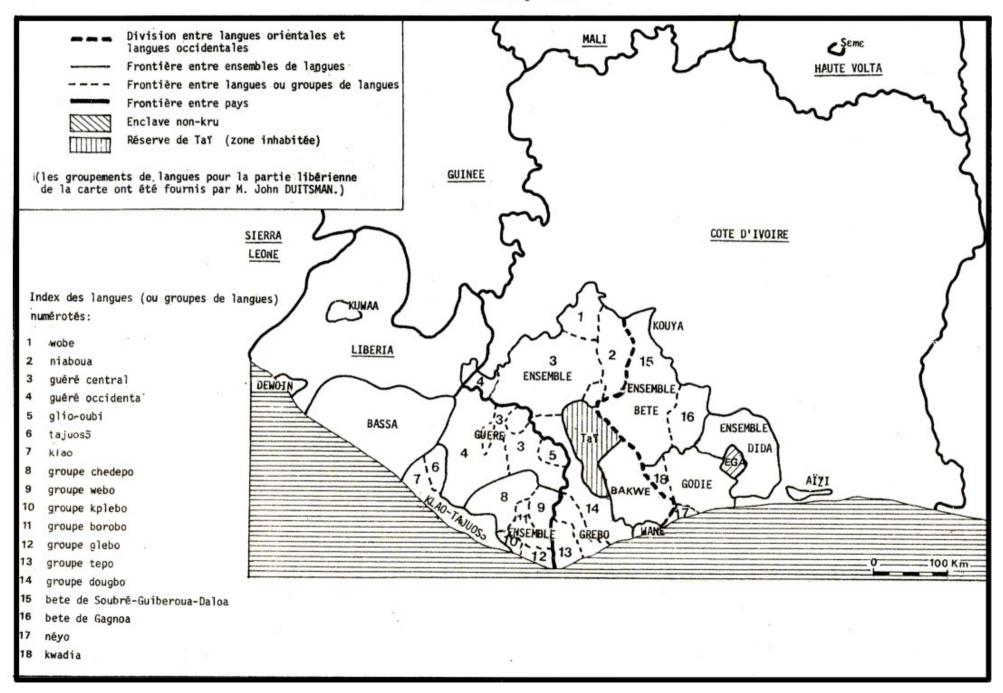


Côte d'Ivoire: Kru languages



Côte d'Ivoire: Kru languages

Carte I : Les langues kru



- SYSTEMES CONSONANTIQUES DE QUELQUES LANGUES ORIENTALES -

														-							
<u>bété</u> d	le Gu	ibér	oua	(Werle	, 197	6)			-	_		et I	Lagako		Коз	70.	(Koko	ra,	1976	5, p. 23)	
								(Marc	chese	, 19	375)				A						
Р	t	С	k	kp	C'W(1)		р	t	С	k	kp	kw			Р	t	С	k	kp C	Cj
p	d	j	g	gb				b	d	j	g	gb	gw			p	d	j.	g	gb	
f	S							f	s							f	s				
\ \ \ \	Z						1 8 1	. V	Z							V .	Z				
В	1	j	¥	W				В	1	·j	¥		W			å	1	j	Y (2	2) . w	
m	n	'n	Û	ŊW				m	n	'n	Û		ŋw			m	n	ת	ŋ		
Névo	(Gra	h)						Dida	de I	LOZOL	ıa (G	ratri	ix)		dic	la-f		(Si	méor	n, Dugas,	Kaye,
																(v	ata)	Ko	opma	an, 1981)	
Р	t	С	k	kp	CW			р	t	C	k	kp	kw			р	t	С	k	kp kw	
b	ď	j	g	gb				Ь	d	j	g	gb	gw			b	đ	j.	g	gb gw	
f	S							f	s							f	s				
\ \ \	z							. V	Z							٧	z				
В	1	j	¥		W			å	1	j	¥.	W				m	n	'n	ŋ	gm ⁽³⁾	
m	n	'n	ŋ					m	n	n	J,	ŋw		•		B	I	j	¥	, W	
1							1								1						

⁽¹⁾ Voir section 2.1.4.4.

⁽²⁾ En Koyo, [Y] n'apparaît que dans quelques lexèmes dont la plupart sont des emprunts.

^{(3) &}quot;Le gm résulte d'une assimilation nasale devant les consonnes labio-vélaires" (Siméon, et. al. 1980:107).

- SYSTEMES CONSONANTIQUES DE QUELQUES LANGUES KRU OCCIDENTALES

	woll q d	t	C	20 c k	nk, 6) k kp gb	kw		Gu P b	t d	(2) c		ishe . 96 kp gb		976	p b		С	(Duit k k 9			p b	depo t		k		c.p.) ⁽⁴⁾	p		19	75,	man, p. kp	92).	al,
	f	S					W		s z l ɗ	j				W	f	s		w			f m	s I n		h	ŋm			f	s	j		w		
	m	n	n	1	ŋm	ka	ו יטע	m	n	n		ŋm	km	ŋw	m	n	Ŋ											m	n	Ŋ	Ŋm	i		
																	731																	
- 60	Nia	abo	ua	(Be		ck,	1975		Dew	oin	(1	Welm	ers)		Ва	assa	())	(Bert	kau,	et al.	. 1	Gre	ebo	(In	nes,	, p. :	14		Tép	<u>o</u> (Daw	son,	MS)	
109 -	Ni:		ua c			ck, kw	1975		Dew t	<u>oin</u>	(1		ers) kw		Ba P		c		kau, kp	et al.	p	Gre t		(In		, p.	14	P		<u>o</u> (son, kw	MS)	
- 109 -	р			p.	8)		1975		t	oin j					Р				kp	et al.			С		kр	, p. :	14						MS)	
- 109 -	р	t d		p. k	8) kp	kw	1975	p b	t	oin j		kp	kw		p b	t	С	k	kp	et al.	р	t .	С	ķ	kр	, p.	14	P b	t		k	kw	MS)	
- 109 -	þ	t d s		p. k	8) kp	kw	1975	p b	t d s	oin j		kp	kw		p b f	t d	С	k	kp	et al.	p b	t .	С	ķ	kр	, p.	14	P b	t ø		k	kw		
- 109 -	p b f	t d s		p. k	8) kp	kw	1975	P b f	t d s	oin j j		kp	kw		p b f	t d s	С	k	kp	et al.	p b	t .	С	ķ	kр	, p. :	14	P b	t ø		k	kw		
- 109 -	p b f	t d s z	c j	p. k	8) kp	kw gw	1975	P b f v &	t d s	j j	k g	kp	kw gw		P b f v	t d s	c j	k	kp gb	et al.	p b	t .	c j	k g	kр		14	P b	t ø		k	kw gb		
- 109 -	p b f v	t d s z	c j	p. k	8) kp	kw gw	1975	P b f v &	t d s	į į	k g	kp	kw gw		P b f v	t d s z	c j	k	kp gb w	et al.	P b f	t d s	c j n	k g	kp gb	w		P b f	t d s		k	kw gb		

⁽¹⁾ La série de nasales en wobé, gueré, tépo et bassa n'est pas phonémique (voir section 2.2.1.2.)

⁽²⁾ D'après Fisher, il y a une opposition entre l et d'en guéré : jdí singes / jli gallons

⁽³⁾ Il semble y avoir une opposition entre j et dj en bassa. Cette opposition n'a pas été relevée dans d'autres langues kru.

⁽⁴⁾ Nous ne savons pas pourquoi les semi-voyelles y et w ne figurent pas sur les tableaux de cedepo et de bassa.

- SYSTEMES CONSONANTIQUES DES LANGUES KRU ISOLEES

(LIBERIA)	(HAUTE-VOLTA)	(COTE-D'IVOIRE)
kuwaa (Thompson, p. 12)	<u>Seme</u> (Prost, p. 346)	Aīzi (Hérault p. 10)
ptkkpkw	ptckkp	ptck kp
b d j	bd j ggb	b d j g gb
f s	f s (ǎ)	f s š
I J Y W	v	v z ž
	· IIj w	l j w
мпро	m n n gm	m n n n
mb nd n j ng nmgb	(h semble être un allophone	
	de f)	

A practical systematisation procedure for a machine-readable database:

Step 1: A word processor or spreadsheet or DBMS table.

Step 2: Export as <u>CSV</u> (character/comma/tab separated value) table.

Step 3: Process manually or automatically: analyse and format as desired.

```
ptckkpkw_bdC_ggb_fs_vz___ B_ljxwmnJNNw_ _ _ _ _ _ _ _ _
Bete
         p t c k kp kw _ b d C _ g gb gw f s _ v z _ _ _ B _ l j x w m n J N Nw _ _ _ _ _ _ _ _ _ _ _
Godie
         ptckkpkwkjbdC_ggb_fs_vz___ B_ljxwmnJN_ _ _ _ _ _ _ _ _ _
Kovo
         ptckkpkw_bdC_ggb_fs_vz___ B_ljxwmnJN_ _ _ _ _ _ _ _ _ _ _ _
Nevo
DidaDeLozoua ptckkpkw_bdC_ggbgwfs_vz___ B_ljxwmnJNNw_ _ _ _ _ _ _ _ _
         ptckkpkw bdC ggbgwfs vz B ljxwmnJN Nm
DidaF
         p t c k kp kw _ b d C _ _ gb _ _ f s _ _ _ _ _ _ _ _ _ w m n J _ Nw Nm km _ _ _ _ _ _ _ _ _ _
Wobe
         p t c k kp kw _ b d C _ g gb gw f s _ v z _ _ _ B D l j _ w m n J _ Nw Nm km _ _ _ _ _ _ _ _
Guere
         ptck_ kw_bdC__gb__fs______l__wmnJ______
Krahn
         p t c k kp kw _ b d C _ _ gb _ f s _ _ _ h _ _ _ l _ _ _ m n J _ _ Nm _ _ _ _ _ _ _ _ _
Cedepo
         p t c k kp kw _ b d C _ _gb _ f s _ _ _ _ _ l j _ w m n J _ _ Nm _ _ _ _ _ _ _ _
Klao
         p t c k kp kw _ b d C _ g gb gw f s _ v z _ _ _ B _ l j _ w m n J _ _ _ _ _ _ _ _ _ _ _ _ _
Niaboua
         Dewoin
         ptckkp__ bdCdjggb__ fs_vz_h hw B_1__w m n J_ Nw__ _ _ _ _ _ _ _ _ _
Bassa
         ptckkp__bdC_ggb_fs___hhw__lj_wmnJNNwNm__hmhnhl___
Grebo
         ptck_ kw_bdC_ggb_fs___h___lj_wmnJN_ Nm_____
Теро
         p t _ k kp kw _ b d C _ _ _ _ f s _ _ _ _ _ _ 1 j x w m n J N _ _ _ _ _ _ mb nd nC Ng Nmgb
KuwaaLiberia
Seme Haute Volta\ p\ t\ c\ k\ p\ \_\ b\ d\ C\ \_\ g\ gb\ \_\ f\ s\ S\ v\ \_\ h\ \_\ \_\ l\ j\ \_\ w\ m\ n\ J\ \_\ \_\ \_\ gm\ \_\ \_\ \_\ \_\ \_\ \_\ \_\ 
         ptckkp__bdC_ggb__fsSvzZ____lj_wmnJN_____
AiziCdI
```

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```
ptckkpkw bdC ggb fs vz B ljxwmnJNNw
Bete
         ptckkpkw_bdC_ggbgwfs_vz___ B_ljxwmnJNNw_ _ _ _ _ _ _
Godie
         ptckkpkwkjbdC_ggb_fs_v ljxwmnJN_____
Kovo
         ptckkpkw bdC ggb fs v
Nevo
                                 Then: 171 language comparisons to do:
DidaDeLozoua ptckkpkw bdC ggbgwfs v
         ptckkpkw bdC ggbgwfs
                                   (n^2-n)/2 = (19^2-19)/2 = 171
DidaF
         ptckkpkw bdC gb fs
Wobe
                                    for 44 features each time: 7524.
         ptckkpkw bdC ggbgwfs
Guere
Krahn
         ptck kw bdC gb fs
         ptckkpkw bdC gb fs
Cedepo
                                         That's a helluva lot.
         ptckkpkw bdC gb fs
Klao
                                    So in comes the software ...
         ptckkpkw bdC ggbgwf
Niaboua
Dewoin
         pt kkpkw bdC ggbgwf
                                   How is the comparison done?
         ptckkp bdCdjggb
Bassa
         ptckkp bdC ggb
Grebo
         ptck_kw_bdC_ggb_fs___h
Теро
KuwaaLiberia pt_k kp kw _ b d C _ _ _ f s _ _ _ _ 1 j x w m n J N _ _ _
                                                                      g Nmgb
SemeHauteVolta p t c k kp _ _ b d C _ g gb _ f s S v _ _ h _ _ _ 1 j _ w m n J _ _ _ _ gm _ _ _ _
         ptckkp__bdC_ggb_fsSvzZ____lj_wmnJN_____
AiziCdI
```

A practical systematisation procedure for a machine-readable database:

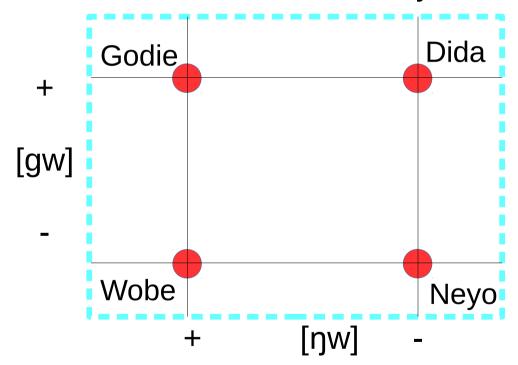
Step 1: A word processor or spreadsheet or DBMS table.

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Step 3: Process manually or automatically: analyse and format as desired.

Gtop G.	. 100000 <u>.</u>		
Bete	ptckkpkw_	Probably the most well-known algorithm for	
Godie	ptckkpkw_	pairwise comparison of sequences:	
Koyo	ptckkpkwk		
Neyo	p t c k kp kw _	Levenshtein Edit Distance	
DidaDeLozoua	p t c k kp kw _	Levensillem Eun Distance	
DidaF	ptckkpkw_		
Wobe	ptckkpkw_	Definition:	
Guere	ptckkpkw_		
Krahn	ptck_ kw_	The minimum number of <u>deletions</u> ,	
Cedepo	p t c k kp kw _	insertions and substitutions needed to	
Klao	ptckkpkw_	change one sequence into another.	
Niaboua	ptckkpkw_	change one sequence into another.	
Dewoin	pt_k kp kw_		
Bassa	ptckkp	Other distance measures can be used.	
Grebo	ptckkp		hn hl
Геро	ptck_ kw_	In stylometry, numerical distances are used	
KuwaaLiberia	pt_k kp kw_	(Comparision of MFW, most frequent	mb nd nC Ng Nmgb
SemeHauteVolt		words, in texts).	
AiziCdI	ptckkp		

- But: the phoneme data matrix is deceptively 2-dimensional: 19 languages x 44 consonants:
- The 19 objects are actually located in a 44 dimensional quality space. Here are 2 of these dimensions, applied to the 4 languages Godie, Dida, Wobe and Neyo:



- Even distinctive features involve around 12 dimensions.
- How to visualise all 44 dimensions in 2 dimensions?

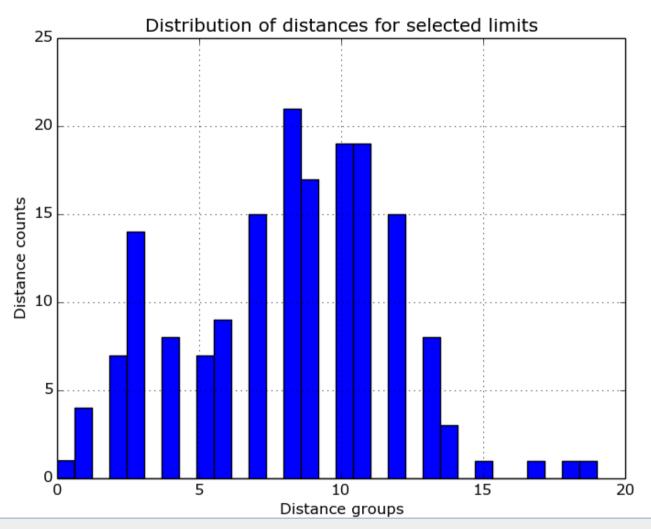
_																			
Bete	Q	1	2	1	1	3	10	6	9	11	8	4	4	7	11	8	12	9	6
Godie	1	Q	3	2	0	2	11	5	10	12	9	3	3	8	12	9	13	10	7
Koyo	2	3	Q	1	3	3	12	8	9	11	8	4	4	9	13	8	12	9	6
Neyo	1	2	1	Q	2	2	11	7	8	10	7	3	3	8	12	7	11	8	5
DidaDeLozoua	1	0	3	2	Q	2	11	5	10	12	9	3	3	8	12	9	13	10	7
DidaF	3	2	3	2	2	Q	11	5	10	10	7	3	3	10	12	7	13	10	7
Wobe	10	11	12	11	11	11	Q	8	6	6	4	10	12	12	11	8	14	11	12
Guere	6	5	8	7	5	5	8	Q	11	11	8	4	6	9	13	10	18	11	10
Krahn	9	10	9	8	10	10	6	11	0	4	3	7	9	10	12	5	11	8	9
Cedepo	11	12	11	10	12	10	6	11	4	Ø	3	9	11	10	10	5	13	8	11
Klao	8	9	8	7	9	7	4	8	3	3	Q	6	8	11	9	4	10	7	8
Niaboua	4	3	4	3	3	3	10	4	7	9	6	0	2	7	13	8	14	7	6
Dewoin	4	3	4	3	3	3	12	6	9	11	8	2	Ø	9	13	8	12	9	6
Bassa	7	8	9	8	8	10	12	9	10	10	11	7	9	0	10	11	19	8	9
Grebo	11	12	13	12	12	12	11	13	12	10	9	13	13	10	0	7	17	10	11
Теро	8	9	8	7	9	7	8	10	5	5	4	8	8	11	7	0	12	7	8
KuwaaLiberia	12	13	12	11	13	13	14	18	11	13	10	14	12	19	17	12	0	15	14
SemeHauteVolta	9	10	9	8	10	10	11	11	8	8	7	7	9	8	10	7	15	0	5
AiziCdI	6	7	6	5	7	7	12	10	9	11	8	6	6	9	11	8	14	5	0

Côte d'Ivoire: Manding languages – phonological rules

TABLEAU Nº 2

CALCUL DE DISTANCE ENTRE LES 23 PARLERS MANDING de COTE D'IVOIRE

N°S des parlers	1	. 5	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23
l		10	7	8	6	3	4	6	6	4	6	3	4	6	7	8	5	7	8	7	8	6	4
2	10	•	8	11	9	4	5	7	7	4	6	4	5	5	3	6	3	4	4	3	3	5	4
3	7	8		11	13	6	5	4	4	6	3	6	6	9	3	6	5	6	2	5	6	5	3
4	8	11	11		12	3	4	5	5	5	4	5	4	6	2	5	4	5	3	2	3	4	3
5	6	9	13	12	•	5	4	3	3	4	2	5	5	8	2	5	4	5	1	4	4	4	2
6	3	4	6	3	5		13	11	11	9	8	12	9	. 7	4	6	3	4	4	5	5	8	12
7	4	5	5	4	4	13	•	12	12	9	9	11	9	6	3	5	5	5	5	3	4	8	13
8	6	7	4	5	3	11	12	•	14	9	10	11	9	4	3	5	3	3	5	2	2	7	11
9	6	7	4	5	3	11	12	14	0	9	10	11	9	4	3	5	3	3	5	2	2	7	11
10	4	4	6	5	4	9	9	9	9		10	12	9	7	5	6	6	5	4	4	4	7	10
11	6	6	3	4	2	8	9	10	10	10		9	7	3	3	4	4	3	6	2	2	6	11
12	3	4	6	5	5	12	11	11	11	12	9		11	7	4	6	4	4	4	3	3	7	10
13	4	5	6	4	5	9	9	9	9	9	7	11	•	7	2	6	5	5	5	3	4	8	8
14	6	5	9	6	8	7	6	4	4	7	3	7	7		7	9	8	9	6	9	10	5	5
. 15	7	3	3	2	2	4	3	3	3	5	3	4	2	7		11	7	8	10	12	11	7	3
16	8	6	6	5	5	6	5	5	5	٤	4	6	6	9	11		7	7	9	1.1	10	7	4
17	5	3	5	4	4	3	5	3	3	6	4	4	5	8	7	7		9	10	10	10	7	6
18	7	4	6	5	5	4	5	3	3	5	3	4	5	9	8	7	9		8	10	9,	5	5
19	8	4	2	3	1	4	5	5	5	4	6	4	5	6	10	9	10	8		10	10	7	5
20	7	3	5	2	4	5	3	2	2	4	?	3	3	9	12	11	10	10	10		13	7	4
21	8	3	6	3	4	5	4	2	2	4	2	3	4	10	11	10	10	9	10	13		7	4
22	6	5	5	4	4	8	8	7	7	7	6	7	8	5	7	7	7	5	7	7	7		7
23	4	4	3	3	2	12	13	11	11	10	11	10	8	5	3	4	6	5	5	4	4	7	·



Spread of differences between 19 Kru consonant inventories for 44 features, which we want to visualise.

Useful strategy: interpret and map <u>differences</u> as <u>distances in quality space</u>.

Strategy #1:

Squash to 2 dimensions!

- Differences are interpreted as distances
- Distances are represented spatially as a distance map
- The dimensions are squashed like a system of springs into 2 dimensions
- Further dimensions may be represented by colours, etc.

Strategy #2:

Select elite features!

- Check the features for their importance in distinguishing objects
- Randomly start with an important feature and build a hierarchy of features distinguishing between sets of objects until all are distinguished
- Different choices lead to different results, different insights

Dealing with high orders of dimensionality

Strategy #1:

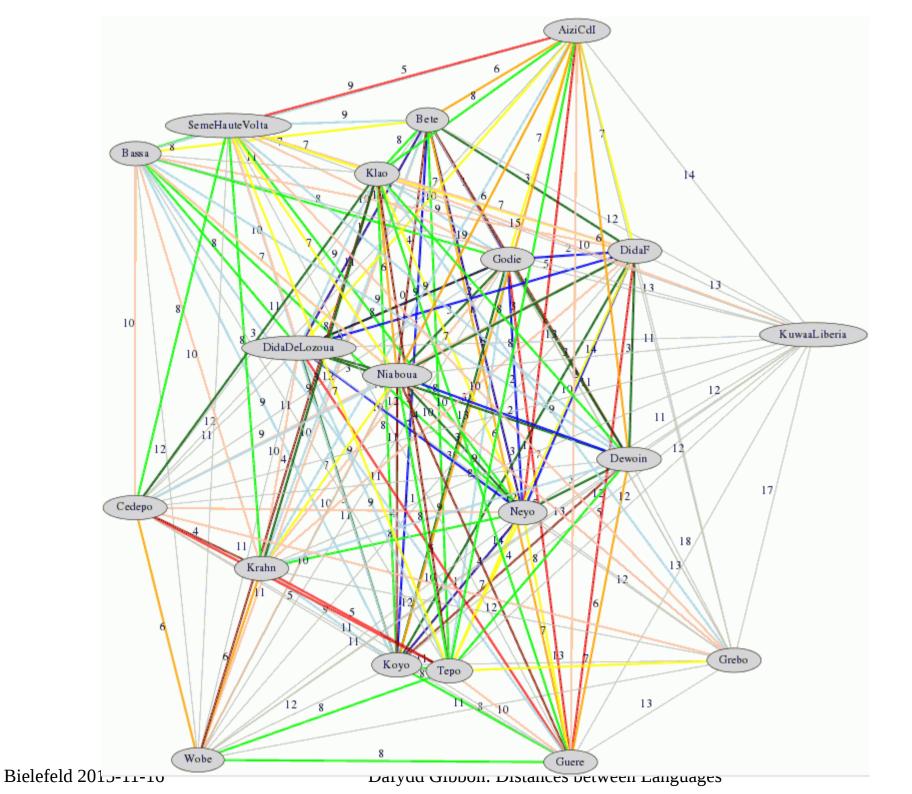
Squash to 2 dimensions!

- Differences are interpreted as distances
- Distances are represented spatially as a distance map
 The dimensions
- The dimensions squashed like a system of springs 2 dimensions
- Further dimensions may be represented by colours, etc.

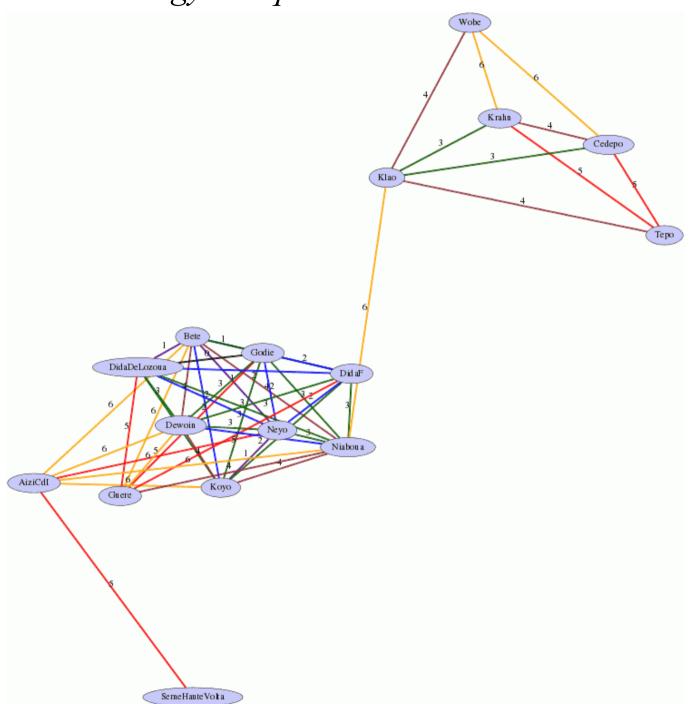
Strategy #2:

Select elite features!

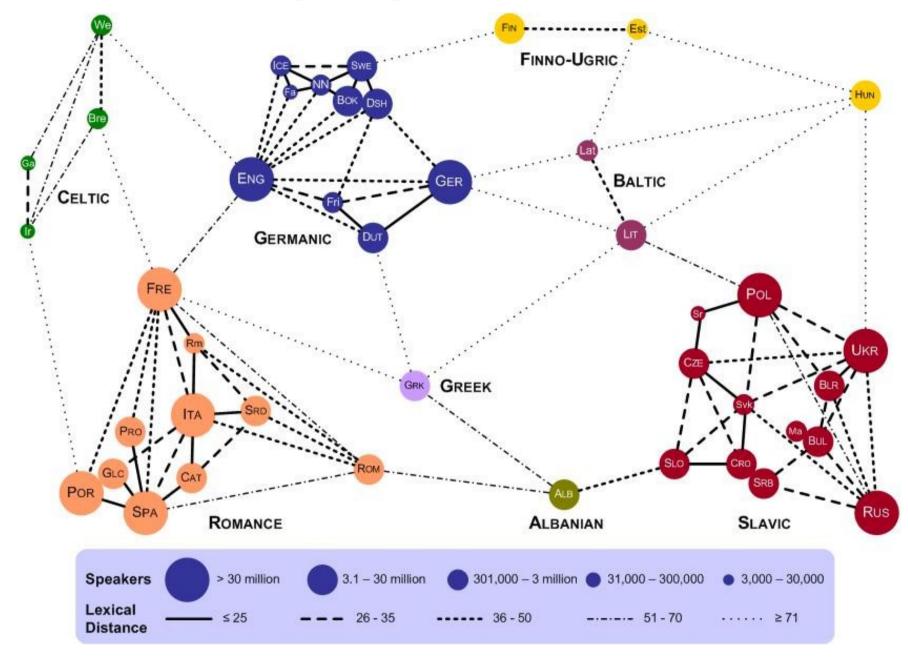
- Check the features for their importance in distinguishing objects Randomly start with an mportant feature and uild a hierarchy of eatures distinguishing between sets of objects until all are distinguished
- Different choices lead to different results, different insights



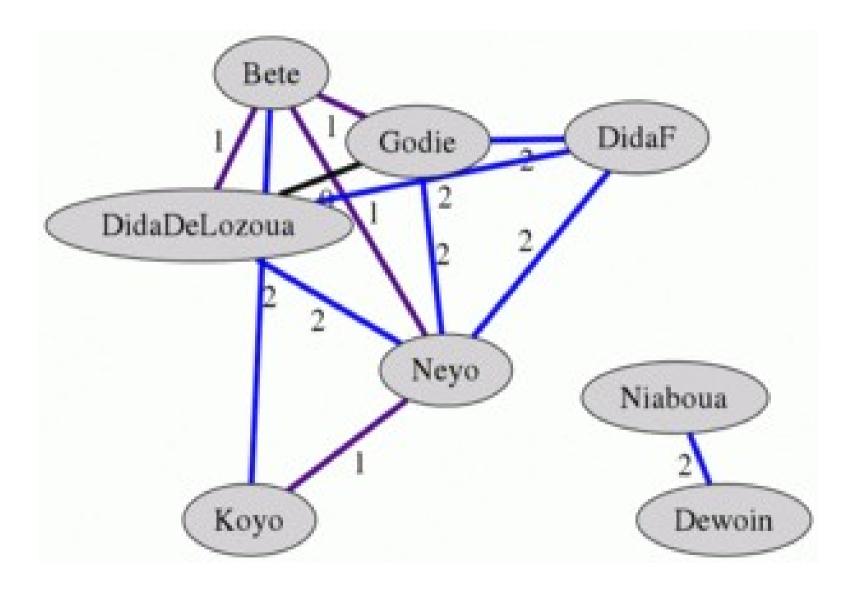
Strategy 1: Squash those dimensions!

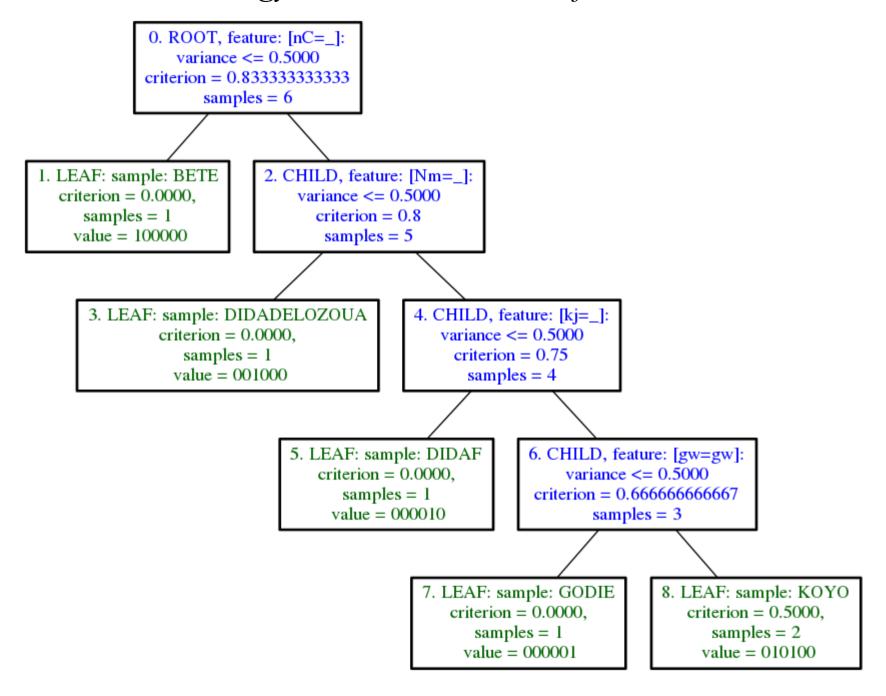


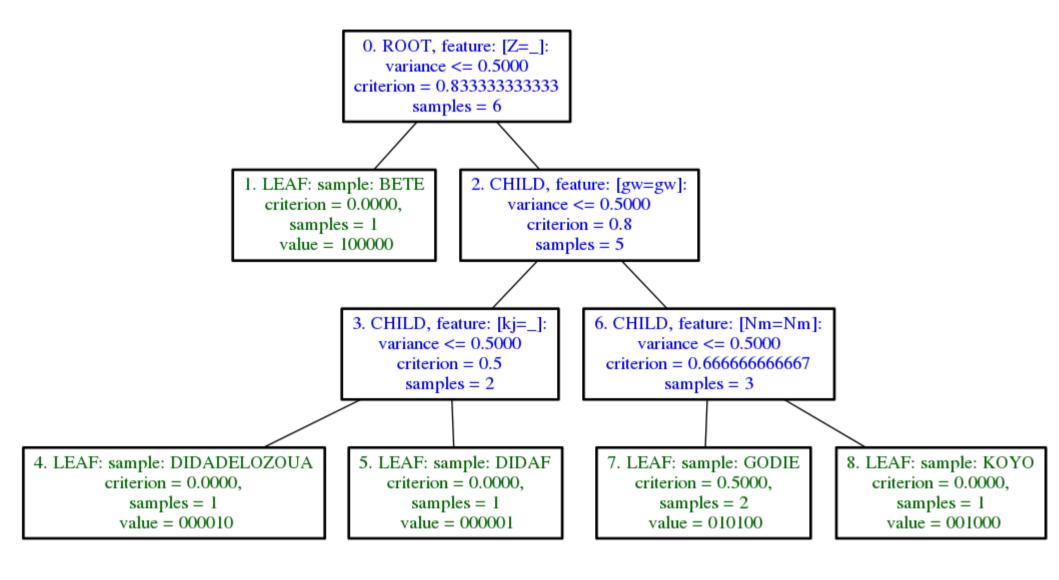
Strategy 1: Squash those dimensions!

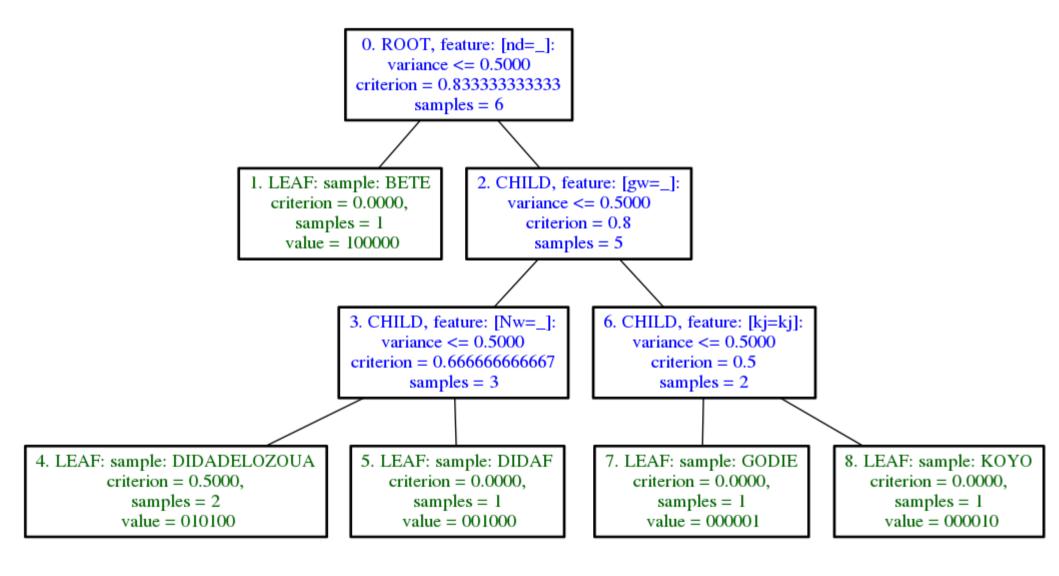


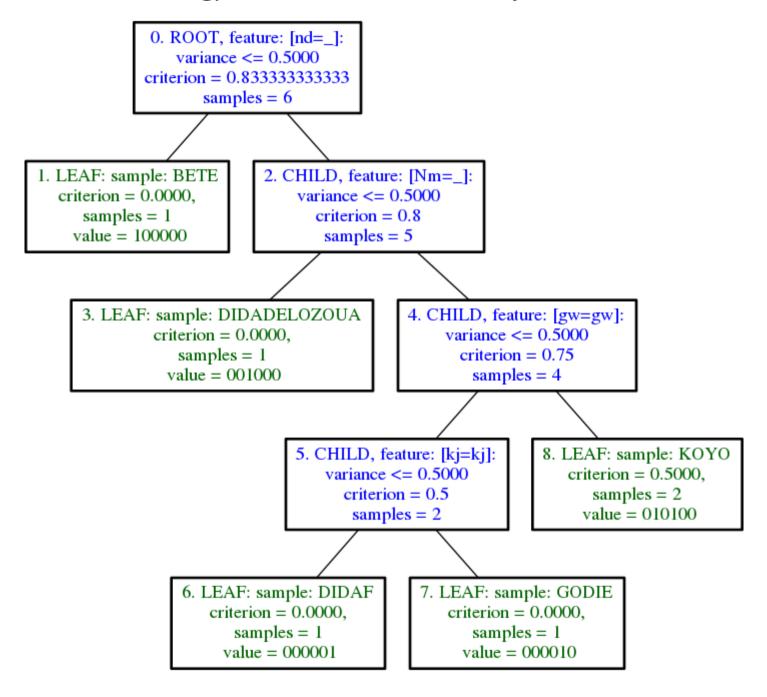
Strategy 1: Squash those dimensions!

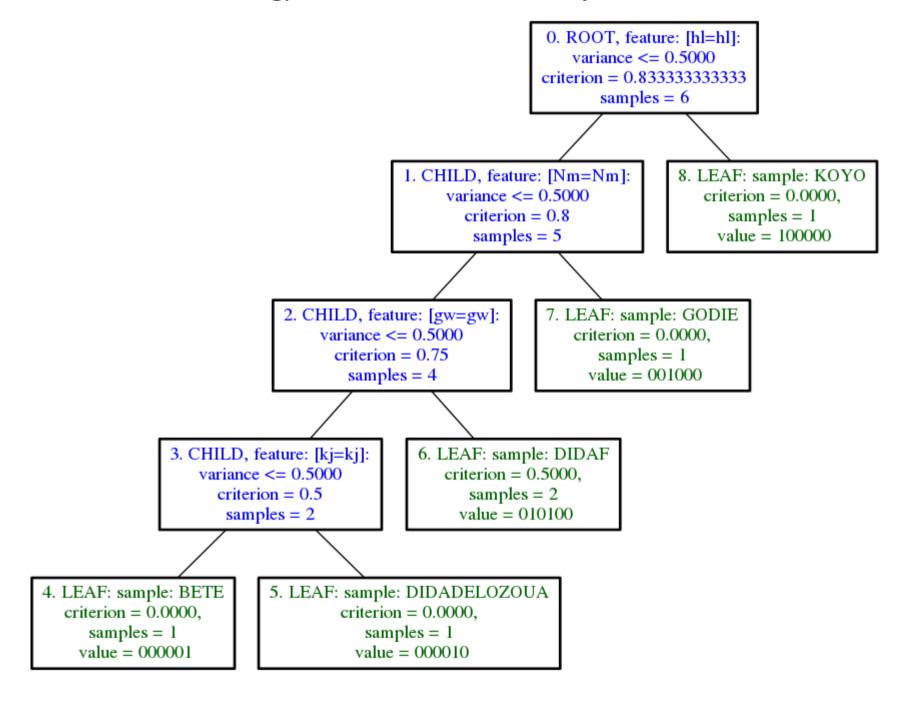


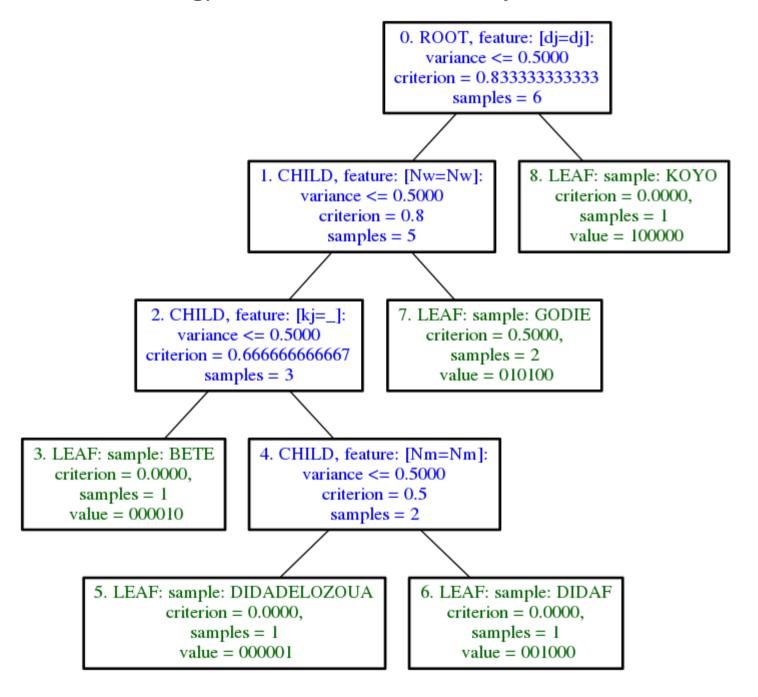


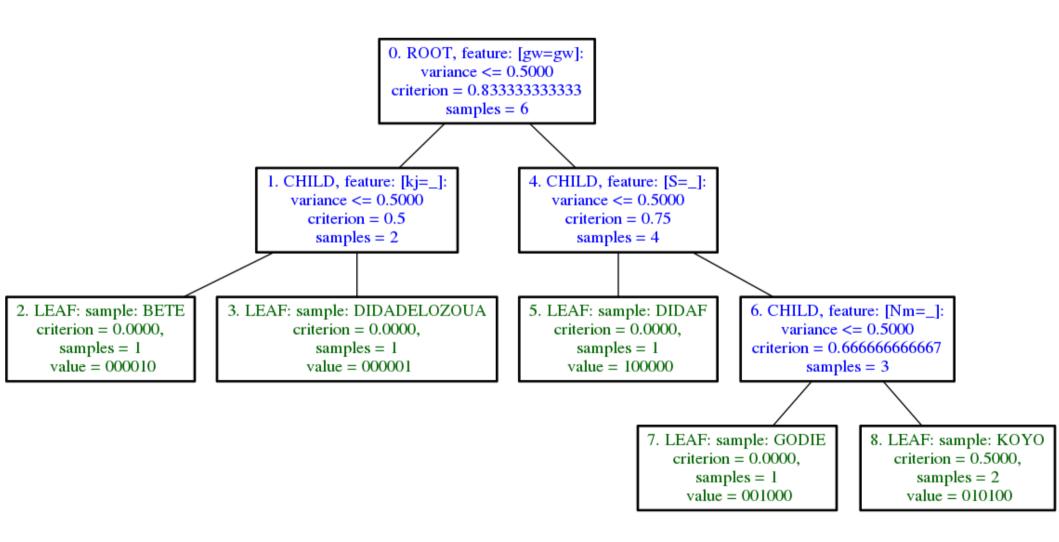












The moral of this story is that

Legacy data in linguistic atlases can be given a new lease of life and a solid quantitative foundation in addition to any further research on dialect relations and history which may be pursued.

Standard arrangements of quantitative information (e.g. tables) may be useful.

Graphical visualisations are helpful in either suggesting or underlining lines of investigation.

Demo: DistGraph

Many thanks – and looking forward to further discussions and future cooperation!