Corpora & Corpus-Informed Learning in EAP

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October 1, Building 10, 4F, Komaba Campus, University Of Tokyo, 15:30 ~ 17:00

What are corpora & what can they tell us?

- * A *principled* collection of texts (spoken or written), stored on computer, which can be analyzed both *quantitatively* and *qualitatively* with the help of analytical software.
- * *Representative* of whatever the corpus is designed to show:

e.g. Corpus of classroom language would need to capture design variables such as student age & gender, location, level, class size, teacher characteristics (gender, qualifications, experience, nationality), etc. Size?

What are corpora & what can they tell us?

- Synchronic corpora ~ show what a target language is like at a given time (usually present)
- Diachronic corpora ~ show how a target language changes through time (historical linguistics)

Wordlists & word frequency

DRD

Top 40 most frequent words: 5m written

2 TO 22 AT 3 AND 23 BUT 4 OF 24 BE 5 A 25 HAVE 6 IN 26 FROM 7 WAS 27 NOT 8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE 20 HAD 40 BEEN	1	THE	21	AS
4 OF 24 BE 5 A 25 HAVE 6 IN 26 FROM 7 WAS 27 NOT 8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERH 19 HIS 39 ONE	2	ТО	22	AT
5 A 25 HAVE 6 IN 26 FROM 7 WAS 27 NOT 8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	3	AND	23	BUT
6 IN 26 FROM 7 WAS 27 NOT 8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	4	OF	24	BE
7 WAS 27 NOT 8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	5	Α	25	HAVE
8 IT 28 THEY 9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	6	IN	26	FROM
9 I 29 BY 10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	7	WAS	27	NOT
10 HE 30 THIS 11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	8	IT	28	THEY
11 THAT 31 ARE 12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	9	Ι	29	BY
12 SHE 32 WERE 13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	10	HE	30	THIS
13 FOR 33 ALL 14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	11	THAT	31	ARE
14 ON 34 HIM 15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	12	SHE	32	WERE
15 HER 35 UP 16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	13	FOR	33	ALL
16 YOU 36 AN 17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	14	ON		HIM
17 IS 37 SAID 18 WITH 38 THERE 19 HIS 39 ONE	15	HER	35	UP
18 WITH 38 THERE 19 HIS 39 ONE	16	YOU	36	AN
19 HIS 39 ONE	17	IS	37	SAID
	18	WITH	38	THERE
20 HAD 40 BEEN	19	HIS	39	ONE
	20	HAD	40	BEEN

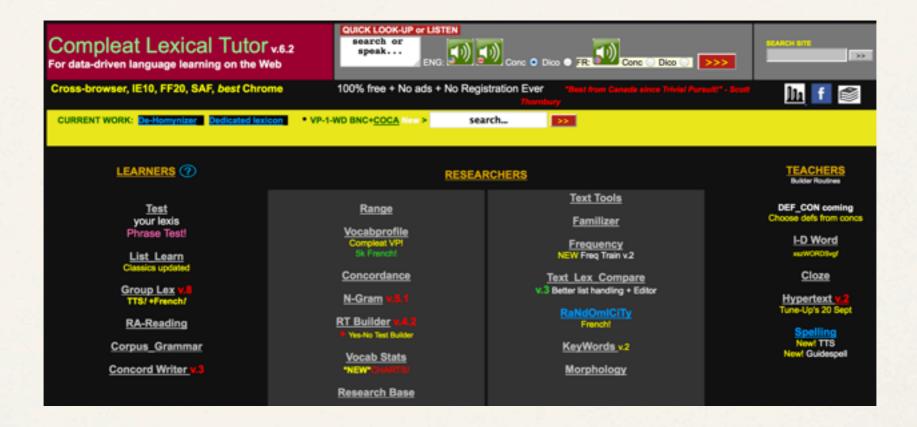
Top 40 most frequent words: 5m spoken

1	THE	21	ON
2	I	22	OH
3	AND	23	WE
4	YOU	24	HAVE
5	IT	25	NO
6	ТО	26	LAUGHS
7	Α	27	WELL
8	YEAH	28	LIKE
9	THAT	29	WHAT
10	OF	30	DO
11	IN	31	RIGHT
12	WAS	32	JUST
13	IT'S	33	HE
14	KNOW	34	FOR
15	MM	35	ERM
16	IS	36	BE
17	ER	37	THIS
18	BUT	38	ALL
19	SO	39	THERE
20	THEY	40	GOT

Wordlists & word frequency

- Written corpus:
- (a) *a* & *the* = high incidence of noun phrases
- (b) *of* = post-modified noun phrases (The house *of* his aunt)
- (c) *that* = subordinator (X reports *that...;* It is claimed *that...*) *that* = relative pronoun (The house *that* Jack built)
- (d) *to, for, in* = prepositional phrases (They drove *to* Scotland *for* a few days)
- Spoken corpus:
- (a) Markers of interactivity: I, you, yeah, know, mm, right, er, laughs

Wordlists & word frequency



- Compleat Lexical Tutor 'Vocabprofile' tool allows teachers to get a frequency breakdown for words in a target text, based on BNC's 1,000 -20,000 word frequency levels
- Engineering journal article vs. Graded reader (Intermediate)

Vocabprofile tool: word frequency breakdown

Journal article

Freq. Level	Families (%)	Types (%)	Tokens (%)	Cumul. token %
K-1 Words :	378 (44.42)	549 (42.04)	7097 (62.86)	62.86
K-2 Words :	175 (20.56)	239 (18.30)	994 (8.80)	71.66
K-3 Words :	66 (7.76)	79 (6.05)	271 (2.40)	74.06
K-4 Words :	72 (8.46)	89 (6.81)	408 (3.61)	77.67
K-5 Words :	40 (4.70)	47 (3.60)	181 (1.60)	79.27
K-6 Words :	24 (2.82)	28 (2.14)	276 (2.44)	8171
K-7 Words :	17 (2.00)	19 (1.45)	192 (1.70)	83.41
K-8 Words :	12 (1.41)	12 (0.92)	187 (1.66)	85.07
K-9 Words :	13 (1.53)	16 (1.23)	81 (0.72)	85.79
K-10 Words :	14 (1.65)	17 (1.30)	220 (1.95)	87.74
K-11 Words :	5 (0.59)	5 (0.38)	9 (0.08)	87.82
K-12 Words :	13 (1.53)	14 (1.07)	153 (1.36)	89.18
K-13 Words :	5 (0.59)	5 (0.38)	15 (0.13)	89.31
K-14 Words :	10 (1.18)	10 (0.77)	24 (0.21)	89.52
K-15 Words :				
K-16 Words :	1 (0.12)	1 (0.08)	1 (0.01)	89.53
K-17 Words :				
K-18 Words :	4 (0.47)	5 (0.38)	30 (0.27)	89.80
K-19 Words :	2 (0.24)	2 (0.15)	4 (0.04)	89.84
K-20 Words :				
Off-List:	77	242 (18.53)	1147 (10.16)	100.00
Total (unrounded)851+?	1306 (100)	11290 (100)	100.00

Words in text (tokens):	11290
Different words (types):	1306
Type-token ratio:	0.12
Tokens per type:	8.64
Pertaining to onlist only	
Tokens:	10143
Types:	1064
Families:	851
Tokens per family:	11.92
Types per family:	1.25

Pertaining to whole text

Graded Reader

Freq. Level	Families (%)	Types (%)	Tokens (%)	Cumul. token %	P
K-1 Words :	310 (72.77)	391 (73.08)	1803 (85.73)	85.73	W
K-2 Words :	65 (15.26)	73 (13.64)	115 (5.47)	91.20	D
K-3 Words :	27 (6.34)	29 (5.42)	41 (1.95)	93.15	Ту
K-4 Words :	9 (2.11)	9 (1.68)	11 (0.52)	93.67	To
K-5 Words :	2 (0.47)	2 (0.37)	2 (0.10)	93.77	
K-6 Words :	4 (0.94)	4 (0.75)	12 (0.57)	94.34	
K-7 Words :	2 (0.47)	2 (0.37)	2 (0.10)	94,44	P
K-8 Words :	2 (0.47)	2 (0.37)	2 (0.10)	94.54	To
K-9 Words :	2 (0.47)	3 (0.56)	29 (1.38)	95.92	T)
K-10 Words :	2 (0.47)	2 (0.37)	7 (0.33)	96.25	F
K-11 Words :	1 (0.23)	1 (0.19)	2 (0.10)	96.35	To
K-12 Words :					T)
K-13 Words :					
K-14 Words :					
K-15 Words :					
K-16 Words :					
K-17 Words :					
K-18 Words :					
K-19 Words :					
K-20 Words :					
Off-List:	22	39 (7.29)	77 (3.66)	100.00	
Total (unrounded)426+?	535 (100)	2103 (100)	100.00	

Words in text (tokens): 2103 Different words (types): 535 Type-token ratio: 0.25		
Different words (types): 535 Type-token ratio: 0.25 Tokens per type: 3.93 Pertaining to onlist only Tokens: 2026 Types: 496 Families: 426 Tokens per family: 4.76	Pertaining to whole text	
Type-token ratio: 0.25 Tokens per type: 3.93 Pertaining to onlist only Tokens: 2026 Types: 496 Families: 426 Tokens per family: 4.76	Words in text (tokens):	2103
Tokens per type: 3.93 Pertaining to onlist only Tokens: 2026 Types: 496 Families: 425 Tokens per family: 4.76	Different words (types):	535
Pertaining to onlist only Tokens: 2026 Types: 496 Families: 425 Tokens per family: 4.76	Type-token ratio:	0.25
Tokens: 2026 Types: 496 Families: 425 Tokens per family: 4.76	Tokens per type:	3.93
Tokens: 2026 Types: 496 Families: 425 Tokens per family: 4.76	Pertaining to onlist only	
Families: 426 Tokens per family: 4.76	Tokens:	2026
Tokens per family: 4.76	Types:	495
	Families:	425
Types per family: 1.16	Tokens per family:	4.76
	Types per family:	1,16

BNC-6,000 types: [fams 4 : types 4 : tokens 12]

VP-negative: k-6

blossom_[1] greenhouse_[2] housekeeper_[8] malaria_[1]

BNC-7,000 types: [fams 2 : types 2 : tokens 2]

VP-regative: k-7 moreover_[1] shrug_[1]

BNC-8,000 types: [fams 2 : types 2 : tokens 2]

VP-regative: k-8 indie_[1] malaysia_[1]

BNC-9,000 types: [fams 2 : types 3 : tokens 29]

W-regative k-9 orchid_[27] thermometer_[2]

BNC-10,000 types: [fams 2 : types 2 : tokens 7]

vP-negative: k-10 hothouse_[4] leech_[3]

BNC-11,000 types: [fams 1 : types 1 : tokens 2]

batten_[2]

BNC-13,000 types: [fams 5 : types 5 : tokens 15]

nordic_[2] outcrop_[2] phosphorus_[7] synoptic_[3] uv_[1]

BNC-14,000 types: [fams 10 : types 10 : tokens 24]

climatology_[4] con_[1] contour_[1] geophysics_[1] glacial_[1] hereinafter_[4] hydrological_[3] runoff_[2] seawater_[2] southernmost_[1] subsurface_[4]

BNC-15,000 types: [fams : types : tokens]

BNC-16,000 types: [fams 1 : types 1 : tokens 1]

labile_[1]

BNC-17,000 types: [fams : types : tokens]

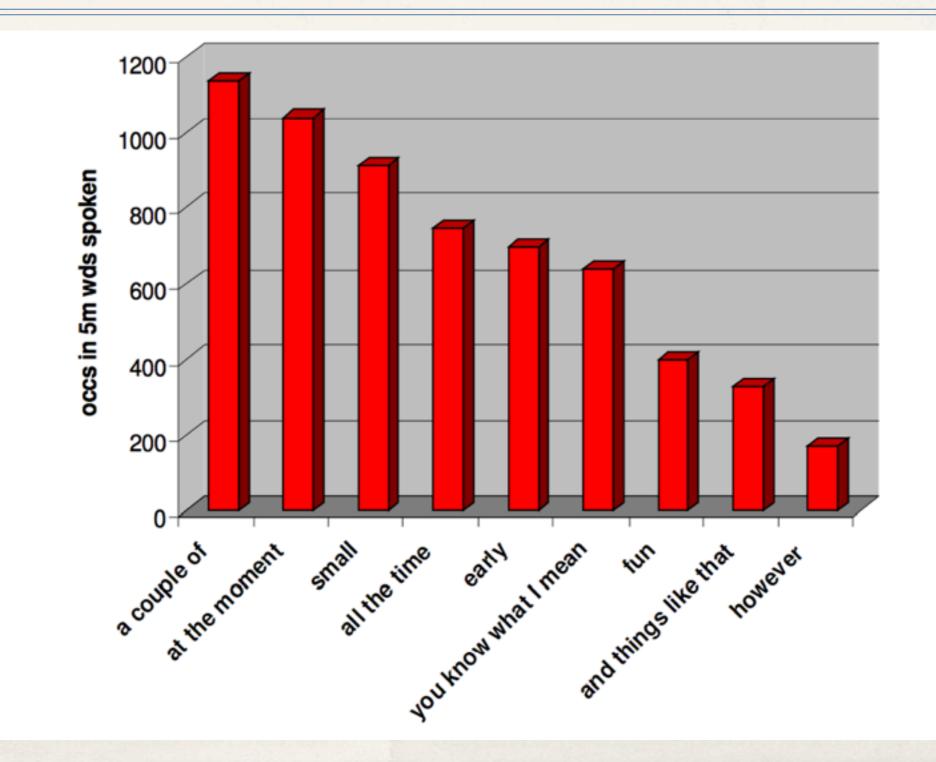
BNC-18,000 types: [fams 4 : types 5 : tokens 30]

denitrification_[11] gyre_[12] hydrographic_[6] isotherm_[1]

BNC-19,000 types: [fams 2 : types 2 : tokens 4]

biogenic_[2] fluvial_[2]

Word-cluster frequency



Sinclair's 'idiom principle' vs. 'open choice principle'

- The production of texts involves alternation between word-for-word combinations (open choice principle) & pre-constructed, multi-word combinations (idiom principle).
- * A large amount of language (~50%) is prefabricated
- * (Chunks larger than 6 or 7 rare)

(Erman & Warren, 2009)

Concordance lines (Key Word in Context - KWIC)

Node

delivery of key speeches in the Supreme Soviet, for example, there was no evidence that prominent groups were gaining proportionately more influence vis-a-vis the party, which may demonstrate latter's consent, unless there is a specific arrangement to the contrary, or evidence that the parties intended otherwise. This draft also has a separate provision relating to in Butcher v Dowlen 1981RTR 24. 1.7 If the judge is undecided on the other evidence after he has heard both sides then the conviction is conclusive, but the judge fait accompli of socialist' reforms to enlist their total support. However the evidence suggests (see pages 75-6) that it is more likely that this decision was the appearance of the national curriculum legislation in England and Wales can be interpreted as evidence of the government's determination to bring the educational system of Scotland into close alignment life is his appointment to the Muftilik, it is necessary here to mention the evidence behind the tradition, while leaving till later the discussion of the validity of the Any officer, servant or agent of the Bank may, on producing if required evidence of his authority, enter any premises occupied by any person on whom a notice consistently recorded higher levels of total approval. At the same time there was some evidence that those who had a more general interest in politics, as distinct from those of maintaining the standards reached in the new services. Some are beginning to show evidence of decay in the activity levels achieved and the associated staff performance. Other work (The problems these women faced as workers are considered in Part II.) Evidence suggests that working class wives were prepared to put up with occasional drinking bouts by the matter should be dealt with -- that is to say, by hearing oral evidence. Accordingly, Lord Meston submits that the mother's removal of the child on a correlation between the presence of active inflammation and PEG absorption. There was little evidence to support the presence of a primary defect in the colonic barrier in patients with model requires that general practitioners and managers develop new skills particularly in contracting. The evidence that contracting is an efficient mechanism in the NHS is still limited. General practitioners minstrel songs of the tenth and eleventh centuries we know exceedingly little. There is evidence, however, of a steady development in oral tradition between the death of Count and others' related to a time when he was in secure accommodation. The evidence in support of that is to be found in the evidence of Helen Taylor. own bodies. # (Griffin 1984: 74-5) Now this is certainly intriguing evidence of the ability of some species to adapt to mirror images (problems gleefully exploited acquired distinctiveness and equivalence solely in terms of associative mechanisms. Is there any positive evidence that might prompt us to adopt the more complex position that differentiation occurs as well of the employment status of either mother or father and without proof of need or evidence of contributions. The acceptance of such a scheme meant that the government had accepted the ruins of the buildings. A more likely suggestion would be that it is evidence of the estate workers continuing to live on the site and work the land long

(BNC academic sub-corpus)

Concordance lines

- Read from the key word (node) outwards, rather than from left to right
- Identify the central group of words which form a phrase or can stand alone
- Sorting alphabetically left or right of the node facilitates identification of common lexico-grammatical patterns in the concordance lines

Concordance lines: Adjectives commonly used with *evidence*

Type/Quality	Quantity	Time sequence
conclusive	enough	present
experimental	some	recent
scientific	little	
clear	no	
available		

Collocation & colligation patterns

Lexical items may be primed to co-occur with other words (collocation), e.g. lean + meat:

, red and green peppers, onions, chickpeas, kidney beans, tomatoes and **lean** diced ham, wholemeal roll with low-fat spread. Dinner Chicken casserole, boiled potatoes sliced tomatoes, fruit with virtually fat-free fromage frais. Dinner Bolognese sauce made with **lean** mince, and wholemeal spaghetti. Virtually fat-free yoghurt for dessert. # FRIDAY cottage cheese, wholemeal toast with low-fat spread and jam or yeast extract. Lunch **Lean** roast meat (visible fat removed), jacket-baked potato, lightly steamed green vegetables # very low-fat cottage cheese # low-fat hard and soft cheeses. # MEAT # Very **lean** minced beef # Meat for roasting (very lean, with visible fat removed) soft cheeses. # MEAT # Very **lean** mince to pan, stirring until browned. Sprinkle 15ml (1 tbsp) flour or yeast extract, very-low-fat curd or cottage cheese, mashed banana. Two rashers **lean** back bacon (visible fat removed) plus two tomatoes and 50g (2oz)

Lexical items may be primed to co-occur with a grammar words (colligation), e.g. possesive adj. + true feelings:

. Hope smiled to himself: the smile broadened, and to disguise his true feelings he turned the smile on Mr Crump; who was greatly encouraged as he had

before the Armistice was signed, she realised that she had been denying her true feelings for years. She had loved Connor from the day she set eyes on him

; you do whatever you wish.' Stephen's controlled voice disguised his true feelings, but Christina sensed his jealousy and changed the subject.' Why don't

you're probably a very good husband, but you like to hide your true feelings." Oh, don't be so serious, Basil,' smiled

a statement of fact. Dexter was surprised that the TV presenter revealed her true feelings towards Nicola so quickly: most people in his experience, when first interviewed by

'd ever seen, but which she always felt quite at odds with her true feelings. In fact, there were lots of things she'd like to change about

turned, looking at the girl, smiling at her reassuringly, keeping his true feelings from showing. Games. It was all one big game to DeVore. He

. It had been then, perhaps, that he had first realised his true feelings for her. Then that he had first articulated it inside his head. I

What else will happen?' she mumbled.' You'll discover my true feelings for you,' he said in a low tone.' Sounds ominous.

Collocation & colligation patterns

Blonde

1	HAIR	141
2	GIRL	19
3	WOMAN	14
4	HEAD	14
5	CURLS	9
6	STRANDS	5
7	HAIR-DYE	3
8	BOMBSHELL	3

(BNC fiction sub-corpus: common collocates of ???)

Part of speech (POS)-tagging

^ JOURNAL_NN1 OF_IO GEOPHYSICAL_JJ RESEARCH_NN1 :_: OCEANS_NN2 ,_, VOL._NN1 118_MC ,_, 16251644Export_FO of_IO nutrients_NN2 from_II the_AT Arctic_JJ Ocean_NN1 [_(1_MC1]_) This_DD1 study_NN1 provides_VVZ the_AT first_MD physically_RR based_VVN mass-balanced_JJ transport_NN1 estimates_NN2 of_IO dissolved_JJ@ inorganic_JJ nutrients_NN2 (_(nitrate_NN1 ,_, phosphate_NN1 ,_, and_CC silicate_NN1)_) for_IF the_AT Arctic_JJ Ocean_NN1

> CLAWS POS-tagging for a journal article: NN1 = singular common noun IO = of (as preposition) JJ = general adjective

Key words (e.g. WordSmith)

- Compares a target corpus with a (larger) reference corpus to identify unusually frequent words
- * Usually identifies 'what a text is about':

📶 ot	hello,	the mo	or of veni	ce.kws							
Eile	<u>E</u> dit	⊻iew	<u>C</u> ompute	<u>S</u> ettings	Window	rs Help					
N			ł	<ey th="" word<=""><th>Freq.</th><th>%</th><th>. Freq.</th><th>RC. %</th><th>Keyness</th><th>P</th><th>emma 🔼</th></ey>	Freq.	%	. Freq.	RC. %	Keyness	P	emma 🔼
1				CASSIO	113	0.43	113	0.01	479.84	0.000000000	
2				IAGO	60	0.23	60		254.67	0.0000000000	
3				MOOR	56	0.21	78		212.19	0.0000000000	
4			DESD	EMONA	40	0.15	40		169.75	0.0000000000	
5			HANDKE	RCHIEF	N 28	0.11	32		113.79	0.0000000000	
6			RO	DERIGO	15 27	0.10	28		113.26	0.0000000000	~
<]										>
KWs	plot	links	clusters	filenames	notes	source tex	đ				

Dispersion

* Visual insights into where key words crop up in a text (WordSmith):

🐔 bible	12.kws					×
File E	dit View Compute Settin	ngs Help				
N	Key word	sion	Links	Hits	Plot	^
283	FEAST	.727 2.01	44	123		
284	CUT	.844 1.63	101	320		
285	AFRAID	.869 1.62	77	193		
286	JEHOSHAPHAT	.321 1.32	25	84		
287	ASHAMED	.6260.55	41	122		
288	WRITTEN	.684 0.36	65	277		
289	WOE	.542 9.13	45	106		
290		.615 8.65		97		
291	PROPHESY	.391 8.03	25	90		
292		.600 7.63		76		
293		.635 7.60		86		
294	FLEE	.816 7.59	30	105		
295		.853 6.42		89		
296		.767 6.36		81		
297		.349 5.95		75		
298		.736 5.77		92		
299	CALLED	.809 5.60	192	625		~
KWs	plot links clusters fil	enames notes	source text			

Type/token ratios

- * A measure of the lexical variety (or vocabulary load) in a text
- TTR = number types / number tokens x 100
- TTRs for speech (~ 50%) tend to be lower than those for writing (~ 70%) because of limited processing time

Semantic prosody

 The way in which seemingly neutral words can take on positive or negative associations through frequent co-occurrence with particular collocations (BYU-BNC):

Cause

	•	CONTEXT	ALL 🗖
1		PROBLEMS	453
2		DEATH	370
3		CONCERN	298
4		DAMAGE	291
5	Θ	EFFECT	210
6	Θ	ACTION	208
7	Θ	TROUBLE	160
8	Θ	PROBLEM	141
9	Θ	HARM	117
10	0	SERIOUS	102
11		INJURY	90
12	Θ	DISEASE	90

Provide

		CONTEXT	ALL 🔍
1	0	INFORMATION	847
2	0	SERVICES	468
3	Θ	SERVICE	430
4		SUPPORT	423
5	0	EVIDENCE	337
6	0	BASIS	284
7	Θ	OPPORTUNITY	228
8	Θ	CARE	209
9		DETAILS	205
10	Θ	USEFUL	192
11		ADVICE	188
12		TRAINING	180

Semantic prosody: wave

BNC: Fiction sub-corpus

		CONTEXT	ALL 🗖
1		HAND	31
2		SWEPT	13
3	Θ	WAVE	12
4		NAUSEA	10
5		GOODBYE	9
6		BROKE	9
7		ARMS	9
8		WASHED	7
9		DIZZINESS	6
10		PANIC	6
11		ANGER	6
12		HIT	6

BNC: Newspaper sub-corpus

		CONTEXT	ALL 🔍
1		VIOLENCE	9
2		STRIKES	8
3		ATTACKS	6
4	Θ	WAND	5
5		SWEPT	5
6		BUYING	5
7		BOMBINGS	4
8		MAGIC	4
9		POLITICAL	4
10		WATER	4
11		JOB	4
12	Θ	SUPPORT	4

Corpora accessible (free) online

British National Corpus (BNC): 100 million words of spoken (10%) & written (90%) language, representing a wide cross-section of 20th century British English

http://www.natcorp.ox.ac.uk/ or *http://corpus.byu.edu/bnc/

Corpus of Contemporary American English (COCA): 450 million word million words of spoken (21%) & written (79%) language, representing a wide cross-section of contemporary American English

http://corpus.byu.edu/coca/

 Michigan Corpus of Academic Spoken English (MICASE): 1.8 million words of academic speech, recorded at University of Michigan

http://quod.lib.umich.edu/m/micase/

 Hong Kong Engineering Corpus (HKEC): 9.2 million words of (mostly written) English from the engineering sector in Hong Kong

http://rcpce.engl.polyu.edu.hk/HKEC/

 Compleat Lexical Tutor: Provides a variety of corpus analysis tools for learners, researchers & teachers, using BNC, COCA & Brown corpora, or user's own text files

http://www.lextutor.ca/

Exploring the BYU corpora

BYU-BNC: BRITISH NAT		ACCESS: 1/						
100 MILLION WORDS, 1980s-	: 7	history lists profile logo						
DISPLAY	SEE CONTEXT: CLICK ON WORD (ALL SECTIONS), NUM	BER (ONE SECTION), OR	CONTEXT] (SELECT) [HEI]	(COMPARE 💽 ? SI	DE BY SIDE 💌
LIST O CHART O KWIC O COMPARE	CONTEXT	ALL 🗖	SPOKEN FICTIO	DN MAGAZINE	NEWSPAPER	NON-ACAD	ACADEMIC	MISC
WORD(S) dog 2		7764	132.98 124.8	3 150.51	84.46	20.31	26.61	83.18
COLLOCATES								0.607 seconds
OS LIST 2								
ANDOM SEARCH RESET								
CTIONS SHOW								
IGNORE 2 IGNORE SPOKEN FICTION FICTION								
MAGAZINE MAGAZINE NEWSPAPER NEWSPAPER NON-ACAD NON-ACAD	KEYWORD IN CONTEXT DISPLAY					Help	Help / information / contact	
						< 1/14 > >>		
ORTING AND LIMITS					SAMPLE: 100	200 500 1000		
	CLICK FOR MORE CONTEXT	C [?] SAVE LI	ST CHOOSE LIST	CREATE NEW LIS	SAMPLE: 100 3	200 500 1000		
ORTING FREQUENCY				(SP:PS0JF) got a dog, on a	ST [?]		PS0JF) that, yeah (S	SP:PS0JB) No t
ORTING FREQUENCY INIMUM FREQUENCY S		A B C me and	(SP:KD1PSUNK) (unclear)		of that, that dog (SP:	PS0JA) Do you? (SP:		SP:PS0JB) No t
ORTING FREQUENCY INIMUM FREQUENCY S C C C C C C C C C C C C C C C C C C	1 KD1 S_conv	A B C me and A B C it? (paus	(SP:KD1PSUNK) (unclear)	(SP:PS0JF) got a dog, on a	o do that, that <u>dog</u> (SP: le <u>dog</u> and she loves to	PS0JA) Do you? (SP: o roll in this. So I hav	ve to tie it up	
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SORTING FREQUENCY MINIMUM FREQUENCY S IDE OPTIONS # HITS FREQ 100 KWIC 100 GROUP BY WORDS DISPLAY PER MIL	1KD1S_conv2G3XS_demonstratn3KE0S_conv4KNWS_conv5KD2S_conv6KPGS_conv	A B C me and A B C it? (paus A B C I (SP:PS) A B C up, has A B C up, has A B C ispect of the second of the se	(SP:KD1PSUNK) (unclear) (SP:KD1PSUNK) (unclear) (SX) (laugh) (SP:PSOSY) (augh) (SP:PSOSY) (SP:PSOJ1) There's the do unclear) (SP:PS555) (laug (re contracted to (pause))	(SP:PS0JF) got a dog, on a t this (unclear) I have a litt A joke? A joke? (SP:PS0SX) e and then says something ggy look! (SP:PS0J7) Oh ay h) Do remember weedy Rol	ST [?] do that, that <u>dog</u> (SP: de <u>dog</u> and she loves to) (laughing) It was a <u>do</u> like guide <u>dog</u> , guard re! Here's the <u>dog</u> . (SP: y? (SP:PSSSA) No. (SP: uation of <u>dog</u> eat dog,	PS0JA) Do you? (SP: o roll in this. So I hav og, yeah! (unclear) (S dog, rabbiting or wha PS0J3) All wrapped o PS555) Roly, my do dare, dare I say. Wh	ve to tie it up SP:PSOSX) (unclear) atever it is. Ok? So th up with him, it's a wo g. (SP:PSSSA) Oh ye at we are endeavour	(SP:PSOSX) W hat is onder she don' tah. (SP:PS55 ing to (pause)

Approaches to corpus-informed language learning

- Indirect approaches: Use corpora to inform materials design & ensure that language models presented are both naturalistic & pedagogically useful
- Direct approaches: Use corpora in the classroom & allow teachers & learners to discover language patterns for themselves

Indirect approaches

* "[The ELT profession] has been rather slow to incorporate corpus methods into its working practices. It is still the case that the majority of ELT materials-writers rely on a combination of their own intuitions and teaching experience, and a well-established canon of apparently self-evident 'facts' about the language which have, more or less, the status of tradition." (Rundell 1996)

Burton (2012), *Corpora* 7.1

- Questionnaire: only 8 out of 13 professional textbook writers had ever used corpora to inform their materials design (extent not reported)
- Decision to query corpora came from authors themselves, rather than publishers
- Reasons cited for *not* referring to corpora: lack of: (a) expertise; (b) access; or (c) time

(Burton, G. (2012). 'Corpora and coursebooks: destined to be strangers forever?' Corpora 7.1, 91-108)

Direct approaches

- Widening gap between corpus-linguistics research and classroom teaching (Römer 2006; McCarthy 2008; Zhang 2008; Aijmer 2009).
- This is despite the recognized pedagogic value of 'data driven learning' (DDL) (e.g. Johns 1991):
- (i) Encourages learners to discover language patterns for themselves (inductively) => greater cognitive processing/deeper learning
- (ii) More learner-centred, allowing learners to focus on items that best fit their particular stage of interlanguage development

(iii) Allows learners to notice language patterns omitted from textbooks or reference books (too complicated or overlooked?)

Reasons for poor take-up of DDL?

- Lack of technical resources or corpus analysis tools (use paper-based worksheets instead?)
- Time-consuming nature of inductive learning
- Destroys learners' world of clear cut grammatical rules
- Unfamiliar terminology (parsing, tagging, type-token ratios, etc.)
- Unpredictability engendered by DDL approach

Example of DDL: *Exploring* Academic English

research Exploring Academic English w mediated filling for student essay writing ortance is ss a particular currer size y tools in the critical *analysis* of television. It is the n social and economic *factors* which together create a ous literature on the *concept* of freedom has sproute eresting that there is evidence that the brain is remar see whether further *research* shows an extension of to the Olympiads. According to the historian Polybiu ertainly not a common *factor* in the evolution of anir As a contentious issue in Cana Jennifer Thurston and the

Example of DDL: *Exploring* Academic English

- Innovative concordance-based workbook for use in EAP classes or for independent learning (see Thurstun & Candlin 1998)
- Deals with 6 important rhetorical functions in EAP: (a) stating the topic; (b) referring to the literature; (c) reporting others' research; (d) discussing processes; (e) hedging; (f) drawing conclusions
- 3 or 4 high frequency words identified from each rhetorical category (indirect approach) & concordance lines analysed by students (direct approach)

Example of DDL: *Exploring* Academic English

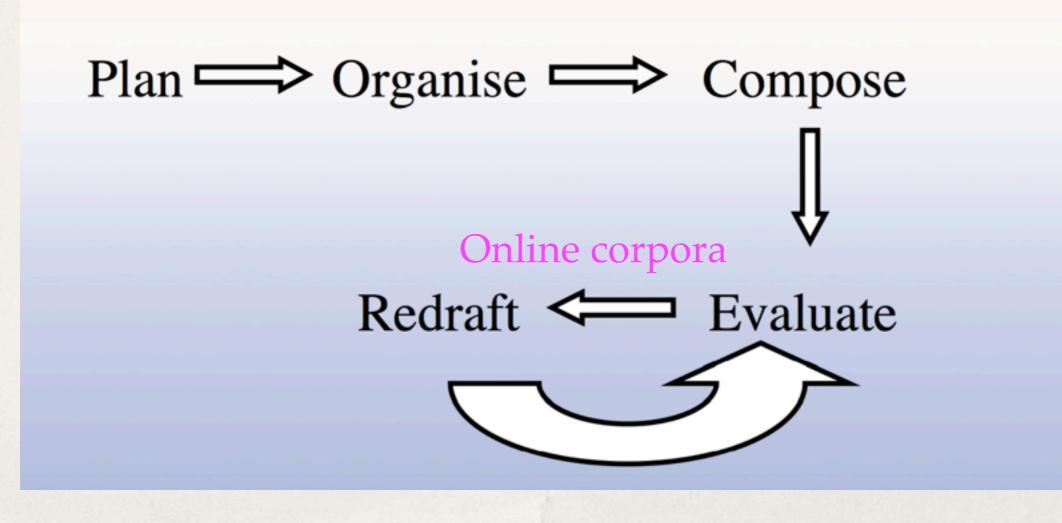
- * 4 stages in each unit:
- (i) 'Look' stage: concordance lines presented (right-sorted)
- (ii) 'Familiarise' stage: identification of lexico-grammatical patterns around key words
- (iii) 'Practice' stage: gap-fill & matching exercises
- (iv) 'Create' stage: students practice using key words in own texts

Exploring Academic English: according to...

- * Try the DDL tasks for the target phrase, according to
- Compare the workbook tasks with the information available for according to online (with the BYU-corpora)
- * Can you see any advantages or disadvantages for choosing paperbased vs. web-based modes of learning in the classroom?

Using online corpora to develop students' writing skills (Gilmore 2009)

Process writing approach



Can online corpora help learners in the re-drafting process?

- * 45 2nd year students in compulsory academic writing class
- 1st draft of factual report on the topic of 'obsession'
- Problem areas in essays highlighted by teacher
- * 30-minute training session on using online corpora
- * Students (in pairs) evaluated mistakes in 1st drafts using BNC & COBUILD corpora
- Redrafting completed out of class
- 1st & 2nd drafts assessed for naturalness (blind rated by 2 NS teachers)
- Students commented on usefulness of online corpora

Training session

- * Typical student errors analysed using online corpora:
- (i) Since then, he started to go.
- (ii)... but we cannot make it worth.
- (iii) My confidence changed.
- (iv) X died for a car accident.

Blind rating of students' writing

Student Writing Samples

The following extracts are taken from university students' academic essays. Please indicate which version you consider to be more natural by placing a cross next to it, for example:

I started to associate with my girlfriend a year ago. _____ I started going out with my girlfriend a year ago. X_____ No difference _____

If you do not consider one to be more natural than the other, please put a cross next to 'No difference \underline{X} '.

Thank you very much for your help with this research. KO (051324)

Today, his fantastic activity allow himself to be the children's hero. ______
 Today, his fantastic plays make himself be the children's hero. ______
 No difference ______

Results

- Total number modifications 1st => 2nd draft = 350 (Range 1-17 per essay)
- * Improved = 67.3%
- * No difference = 27.4%
- * Worse = 5.3%

Example modifications

Improved:

1st draft: He became popular in the USA not only Japan.2nd draft: He became popular not only Japan but also in the USA.

No difference:

1st draft: Her activity will attract people as before.

2nd draft: Her activity will attract people same as before.

Worse:

1st draft: Underage smoking was prohibited in Japan, so she couldn't avoid

fired.

2nd draft: Underage smoking was prohibited in Japan, so she couldn't evade displacement.

Student feedback

* Were online corpora useful for redrafting your essays?

• Yes = 95%

" I think it is very useful for me because I can know the native speaker's sentences. In my dictionary there are many sentences but they are not natural sentences".

✤ No = 5%

"If I don't know what I should type, I can't find out".

"Corpus is not useful for me because it is complicated".

Conclusion

- Corpora and corpus analysis tools can make a valuable contribution to EAP classes
- Online access and functionality has improved considerably in recent years
- Challenges exist but can be overcome with adequate training & motivation

Thank you!

ありがとうございました

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