Appendix A: Simulation word / nonword sets.

The following pages contain simulation word / nonword sets for:

The Word Similarity Simulation (Chapter 4).

The Monolingual Priming Simulation (Chapter 6).

- Word set.

- Orthographic and Semantic Codes

The Bilingual Single Network Model (Chapter 7).

- Languages A and B.

- Orthographic and Semantic Codes

The Word Similarity Simulation (Chapter 4).

La	ngu	lage	e O	ne (OG	is)		La	ngu	age	e Tv	vo (AY	s)	
Bases	Word Simila condi	,	Langi Inform condi	nation	Mapp Simila condi	arity	Mapping	Bases	Word Simila condi	arity	Lang Inforr condi	nation	Mapp Simila condi	arity	Mapping
Words	Input	Output	Input	Output	Input	Output	Туре	Words	Input	Output	Input	Output	Input	Output	Туре
dog	-	-	A	A	A	A	au	tay	-	-	В	в	В	В	au
fog	-	-	A	A	Α	A	au	gay	-	-	В	В	В	В	au
lip*	-	-	Α	A	A	A	au	fop*	-	-	В	В	В	В	au
log	-	-	Α	A	Α	A	au	day	-	-	В	В	В	В	au
bog	-	-	A	A	Α	A	au	zay	-	-	В	В	В	В	au
get*	-	-	A	A	А	A	au	zat*	-	-	В	В	В	В	au
mog	-	-	A	A	Α	В	tr	fay	-	-	В	В	В	A	tr
sog	-	-	A	A	Α	В	tr	bay	-	-	В	В	В	A	tr
bep*	-	-	Α	A	Α	В	tr	set*	-	-	В	В	В	A	tr
tog	-	-	Α	A	Α	В	tr	lay	-	-	В	В	В	A	tr
gog	-	-	Α	Α	А	В	tr	may	-	-	В	В	В	A	tr
bat*	-	-	Α	Α	А	В	tr	biy*	-	-	В	В	В	A	tr

Mapping Types:

au = "Autoassocation" tr = "Transformation"

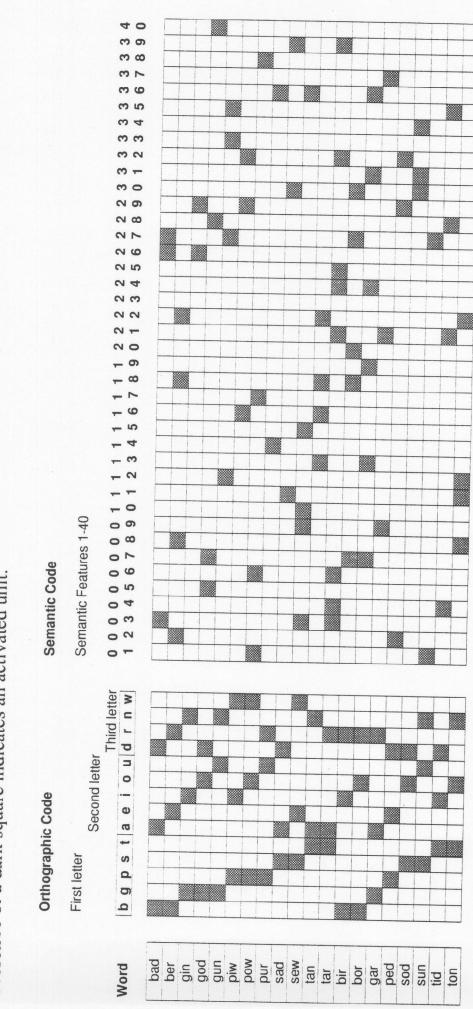
Starred words do not conform to their language's predominant regularity.

Words	Nonw	ords		
bad	ban		ala	
ber	bar	gen	pin	SOW
bir	baw	ger	pir	sud
bor		gew	pod	sur
	bed	gid	pon	SUW
gar	ben	gir	por	tad
gin	bew	giw	pud	taw
god	bid	gon	pun	ted
gun	bin	gor	puw	ten
ped	biw	gow	san	ter
piw	bod	gud	sar	tew
voo	bon	gur	saw	tin
pur	bow	guw	sed	tir
sad	bud	pad	sen	tiw
sew	bun	pan	ser	tod
sod	bur	par	sid	tor
sun	buw	paw	sin	tow
tan	gad	pen	sir	tud
ar	gan	per	siw	tun
id	gaw	pew	son	tur
on	ged	pid	sor	tuw

The Monolingual Priming Simulation (Chapter 6).

Orthographic and Semantic Codes for the monolingual priming wordset.





Bilingual Single Network simulation (Chapter 7).

Key.

V = Vowel (a, e, i, o, u) C = Consonant (b, f, g, s, t) HF = High Frequency LF = Low Frequency (30% of HF learning rate).) SINGLE = Word form exists in only one language BOTH = Word form exists in both languages. LL = Legal in both languages LN/NL = Legal in one language, Non-legal in the other.

COG = Cognate homograph (same form and meaning in each language)

NCG = Non-cognate homograph (same form, different meaning in each language).

-								
1	11	12	13	word	lang	rule	frequency	word type
1	b	u	u	buu	A	CVV	HF	SINGLE_LL
2	9	е	a	gea	A	CVV	HF	SINGLE_LL
3	1	U	e	tue	A	CVV	HF	SINGLE_LL
	g	i	i	gii	A	CVV	HF	SINGLE_LL
5	g	i	0	gio	A	CVV	HF	SINGLE_LL
67	9	0	i	goi	A	CVV	HF	SINGLE_LL
8	f	ei	a	fea	A	CVV	HF	SINGLE_LL
9	g	0	e	gie	A	CVV	HF	SINGLE_LL
9	f	i	0	SOO	A	CVV	HF	SINGLE_LL
÷.,	-	i	U	fiu		CVV	HF	SINGLE_LL
1	S		0	SIO	A	CVV	LF	SINGLE_LL
23	b	0	u	bou	A	CVV	LF	SINGLE_LL
3	1	1	u	biu	A	CVV	LF	SINGLE_LL
4	f		u	tiu		CVV	LF	SINGLE_LL
5	S	u	0	fuo	A	CVV	LF	SINGLE_LL
7	1	a	e	sae		CVV	LF	SINGLE_LL
8	b	a	u	tau	A	CVV	LF	SINGLE_LL
9	S	u	e	bue	A	CVV	LF	SINGLE_LL
0	b	a	0	suo	A	CVV	LF	SINGLE_LL
1	S	a u	eu	bae	A	CVV	LF	SINGLE_LL
2	b	u	u o	suu buo	A	CVV	HF	BOTH_COG
3	f	a	a	faa	A	CVV		BOTH_COG
4	f			fia		CVV	HF	BOTH_NCG
5	1	1	a		A	CVV	HF	BOTH_NCG
6			e	tie	A	CVV	LF	BOTH_COG
7	g	a	0	gao	A	CVV	LF	BOTH_COG
8	b	0 a	a	boa	A	CVV	LF	BOTH_NCG
9	1	a	b	tab	A	CVV	LF	BOTH_NCG
0		i			A	CVC	HF	SINGLE_LL
1	g	u	g	gig	A	CVC	HF	SINGLE_LL
2	b	a	S	sub bas	A	CVC	HF	SINGLE_LL
3	1	a		tas	A	CVC	HF	SINGLE_LL
4	f	a 0	S	fos	A	CVC	HF	SINGLE_LL
5	b	i	f	bif	A	CVC	HF	SINGLE_LL
5	S	0	1	sot	A	CVC	HF	SINGLE_LL
7	b	e	f	bef	A	CVC	HF	SINGLE_LL
B	S	i	1	sit	A	CVC	HF	SINGLE_LL
9		i	f		A	CVC		SINGLE_LL
)	g			gif	A	CVC	LF	SINGLE_LL
1	1	a e	S	fas	A	CVC	LF	SINGLE_LL
2	g s	u	g	geg	A	CVC	LF	SINGLE_LL
3	1		S	SUS		CVC	LF	SINGLE_LL
		1	f 1	tif	A	CVC	LF	SINGLE_LL
5	g	u		gut		CVC	LF	SINGLE_LL
+	b	a	9	bag	A	CVC	LF	SINGLE_LL
5	b	u	1	buf	A	CVC	LF	SINGLE_LL
	S	u	1	sut	A	CVC	LF	SINGLE_LL
3	f	0	1	fot	A	CVC	LF	SINGLE_LL
)	f	a	1	fat	A	CVC	HF	BOTH_COG
	b	a	f	baf	A	CVC	HF	BOTH COG

Language B

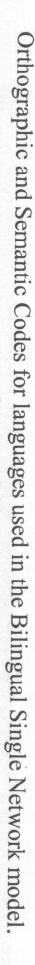
#	11	12	13	word	lang	rule	frequency	word type
1	S	a	u	sau	B	CVV	HF	SINGLE_LL
2	t	е	u	teu	В	CVV	HF	SINGLE_LL
3	1	0	u	tou	В	CVV	HF	SINGLE_LL
4	g	0	0	goo	В	CVV	HF	SINGLE_LL
5	1	i	0	tio	В	CVV	HF	SINGLE LL
6	g	a	е	gae	B	CVV	HF	SINGLE_LL
7	S	u	е	sue	B	CVV	HF	SINGLE LL
8	1	a	i	tai	В	CVV	HF	SINGLE_LL
9	S	i	е	sie	В	CVV	HF	SINGLE_LL
10	1	a	0	tao	В	CVV	HF	SINGLE_LL
11	b	е	i	bei	В	CVV	LF	SINGLE_LL
12	g	0	u	gou	В	CVV	LF	SINGLE_LL
13	1	е	0	teo	В	CVV	LF	SINGLE_LL
14	f	u	i	fui	B	CVV	LF	SINGLE_LL
15	g	i	a	gia	В	CVV	LF	SINGLE_LL
16	S	е	0	seo	В	CVV	LF	SINGLE_LL
17	S	0	i	soi	В	CVV	LF	SINGLE_LL
18	f	0	a	foa	В	CVV	LF	SINGLE_LL
19	t	0	İ	foi	В	CVV	LF	SINGLE_LL
20	t	е	е	tee	В	CVV	LF	SINGLE_LL
21	S	u	u	suu	B	CVV	HF	BOTH_COG
22	b	u	0	buo	B	CVV	HF	BOTH_COG
23	f	a	a	faa	В	CVV	HF	BOTH_NCG
24	f	1	a	fia	B	CVV	HF	BOTH_NCG
25	1	i	9	tie	В	CVV	LF	BOTH_COG
26	g	a	0	gao	B	CVV	LF	BOTH_COG
27	b	0	a	boa	B	CVV	LF	BOTH_NCG
28 29	b	a	a	baa	B	CVV	LF	BOTH_NCG
	g	a	1	gat	B	CVC	HF	SINGLE_LL
30	1 b	a	f	taf	B	CVC	HF	SINGLE_LL
32	D S	i	b f	bib	B	CVC	HF	SINGLE_LL
33	b			sif	B	CVC	HF	SINGLE_LL
34	f	0 i	g f	bog fif	B	CVC	HF	SINGLE_LL
35	1	e	S	tes	B	CVC	HF	SINGLE_LL
36	f	6	f	fef	B	CVC	HF	SINGLE_LL
37	b	i	1	bit	B	CVC		SINGLE_LL
38	1	0	f	tof	B	CVC	HF	SINGLE_LL
39	b	u			B	CVC		SINGLE_LL
40	g	e	g 1	bug	B	CVC	LF	SINGLE_LL
40	1	e		get	B	CVC	LF	SINGLE_LL
42	bi	u	9	teg but	B	CVC CVC	LF	SINGLE_LL
43		a	S		B			SINGLE_LL
43	g s	e		gas	B	CVC	LF	SINGLE_LL
45	1	u	g	seg fug	B	CVC	LF	SINGLE_LL
45	1	u	g	tus	B	CVC	LF	SINGLE_LL
40	b	0	S	bos	B	CVC CVC	LF	SINGLE_LL
47	S	0	b	sob	B		LF	SINGLE_LL
40	f	a	1	fat	B	CVC	HF	SINGLE_LL
50	b	a	f	baf	B	CVC CVC		BOTH_COG
50	5	a	1	Dai	D	CVC	HF	BOTH_COG

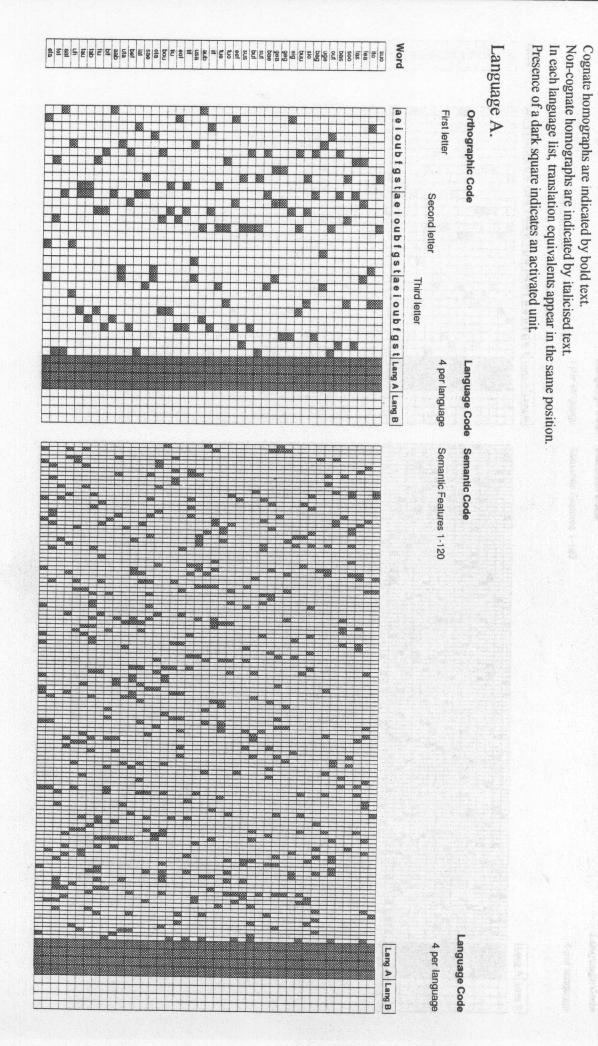
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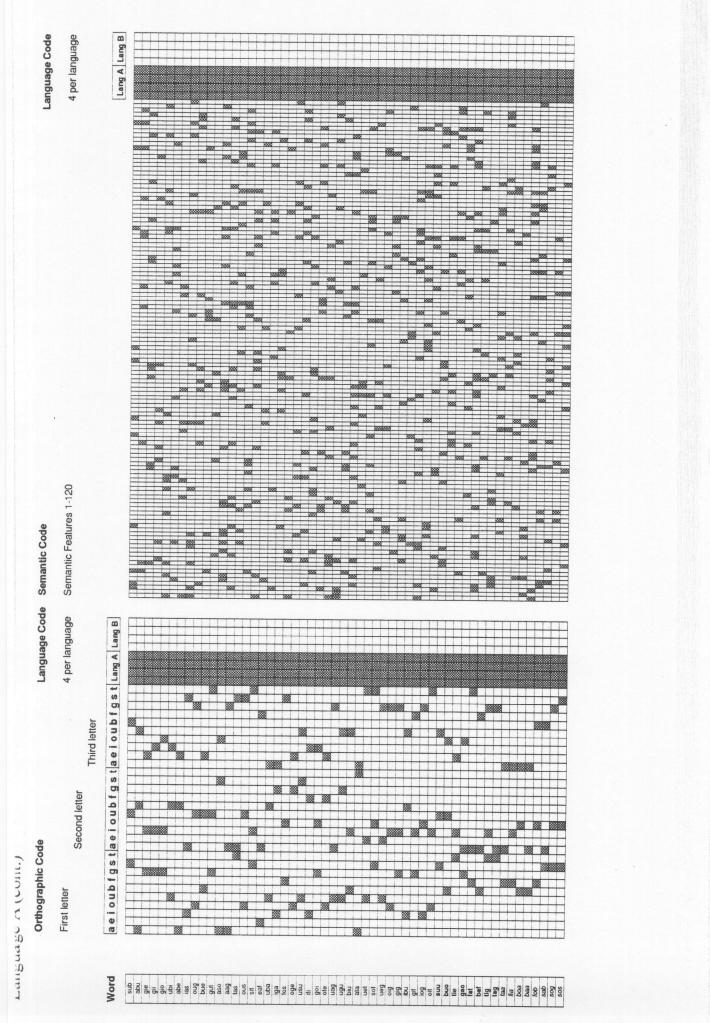
Language A

*	11	12	13	word	lang	rule	frequency	word type
51	ſ	0	b	fob	A	CVC	HF	BOTH_NCG
52	S	a	b	sab	A	CVC	HF	BOTH_NCG
53	t	i	g	tig	A	CVC	LF	BOTH_COG
54	t	a	g	tag	A	CVC	LF	BOTH_COG
55	S	0	g	sog	A	CVC	LF	BOTH_NCG
56	S	0	S	SOS	A	CVC	LF	BOTH_NCG
57	i	1	0	ito	A	VCV	HF	SINGLE_LN
58	a	1	a	ata	A	VCV	HF	SINGLE LN
59	a	b	u	abu	A	VCV	HF	SINGLE LN
60	u	1	a	uta	A	VCV	HF	SINGLE_LN
61	0	g	e	oge	A	VCV	HF	SINGLE LN
62	u	f	i	ufi	A	VCV	HF	SINGLE LN
63	u	9	e	uge	A	VCV	HF	SINGLE LN
64	u	S	a	usa	A	VCV	HF	SINGLE LN
65	i	b	u	ibu	A	VCV	HF	SINGLE LN
66	a	S	u o	aso	A	VCV	HF	SINGLE_LN
67	e	f	a	efa	A	VCV	LF	SINGLE LN
68		b		uba	A	VCV	LF	SINGLE LN
69	u	-	a		A	VCV		SINGLE_LN
	u	S	u	usu			LF	
70	0	1	a	ota	A	VCV		SINGLE_LN
71	u	b	i	ubi	A	VCV	LF	SINGLE_LN
72	u	g	U	ugu	A	VCV	LF	SINGLE_LN
73	0	f	e	ofe	A	VCV	LF	SINGLE_LN
74	a	b	6	abe	A	VCV	LF	SINGLE_LN
75	i	g	a	iga	A	VCV	LF	SINGLE_LN
76	i	f	i	ifi	A	VCV	LF	SINGLE_LN
77	i	i	f	iif	A	VVC	HF	SINGLE_LN
78	i	0	g	iog	A	VVC	HF	SINGLE_LN
79	u	e	1	uet	A	VVC	HF	SINGLE_LN
80	0	i	1	oit	A	VVC	HF	SINGLE_LN
81	e	e	f	eef	A	VVC	HF	SINGLE_LN
82	a	u	b	aub	A	VVC	HF	SINGLE_LN
83	u	e	g	ueg	A	VVC	HF	SINGLE_LN
84	0	u	g	oug	A	VVC	HF	SINGLE_LN
85	0	i	g	oig	A	VVC	HF	SINGLE_LN
86	0	u	1	out	A	VVC	HF	SINGLE_LN
87	a	a	1	aat	A	VVC	LF	SINGLE_LN
88	е	i	g	eig	A	VVC	LF	SINGLE_LN
89	u	a	g	uag	A	VVC	LF	SINGLE_LN
90	a	a	g	aag	A	VVC	LF	SINGLE_LN
91	0	u	S	ous	A	VVC	LF	SINGLE_LN
92	e	a	f	eaf	A	VVC	LF	SINGLE_LN
93	i	a	1	iat	A	VVC	LF	SINGLE_LN
94	e	0	f	eof	A	VVC	LF	SINGLE_LN
95	a	a	b	aab	A	VVC	LF	SINGLE_LN
96	i	a	S	ias	A	VVC	LF	SINGLE LN

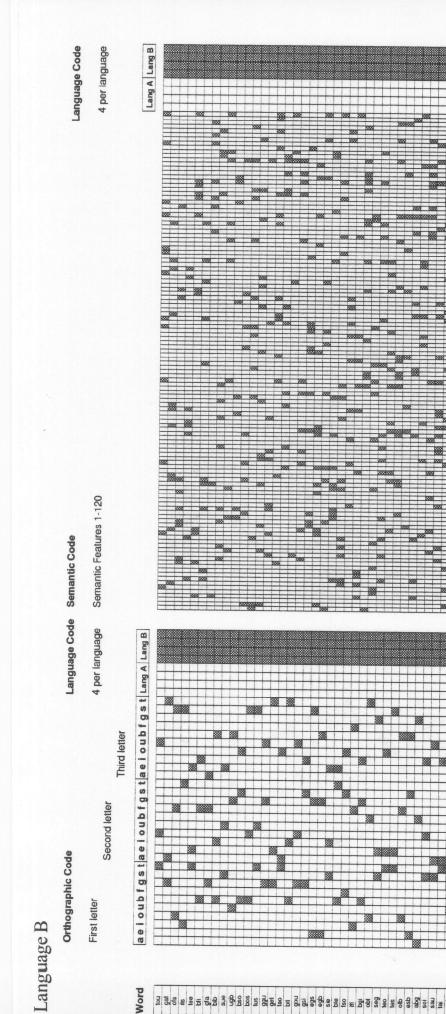
				V		0	e E	
#	11	12	13	word	lang	rule	frequency	word type
51	T	0	b	fob	B	CVC	HF	BOTH_NCG
52	S	a	b	sab	B	CVC	HF	BOTH NCG
53	1	i	g	tig	B	CVC	LF	BOTH_COG
54	1	a	g	tag	B	CVC	LF	BOTH_COG
55	S	0	g	sog	B	CVC	LF	BOTH_NCG
56	S	0	S	SOS	B	CVC	LF	BOTH_NCG
57	g	g	u	ggu	B	CCV	HF	SINGLE NL
58	b	S	a	bsa	B	CCV	HF	SINGLE_NL
59	S	S	I	ssi	B	CCV	HF	SINGLE_NL
60	b	S	0	bso	B	CCV	HF	SINGLE NL
61	S	g	i	sgi	B	CCV	HF	SINGLE NL
62	S	1	0	sto	B	CCV	HF	SINGLE NL
63	b	1	e	bte	B	CCV	HF	SINGLE NL
64	f	f	0	ffo	B	CCV	HF	SINGLE NL
65	b	g	i	bgi	B	CCV	HF	SINGLE NL
66	S	g	e	sge	B	CCV	HF	SINGLE NL
67	S	1	u	stu	B	CCV	LF	SINGLE NL
68	1	1	a	tta	B	CCV	LF	SINGLE NL
69	g	f	a	gfa	B	CCV	LF	SINGLE NL
70	f	S	0	fso	B	CCV	LF	SINGLE NL
71	b	f	i	bfi	B	CCV	LF	SINGLE NL
72		f	0	gfo	B	CCV	LF	SINGLE NL
73	g	S	u	ssu	B	CCV	LF	SINGLE NL
74	b	1	a	bta	B	CCV	LF	SINGLE NL
75		S	i	gsi	B	CCV	LF	SINGLE NL
76	g	b	u	sbu	B	CCV	LF	SINGLE_NL
77	9	S	b	esb	B	VCC	HF	SINGLE NL
78	0	b	1	obt	B	VCC	HF	SINGLE NL
79	0	S	b	osb	B	VCC		SINGLE_NL
80	U	1	f	utf	B	VCC	HE	SINGLE_NL
81	u		b	unuugb	B	VCC	HF	SINGLE_NL
82	e	g	b	¥	B	VCC	HF	SINGLE_NL
83	0	g	b	egb ofb	B	VCC	HF	SINGLE NL
84	i		b	igb	B	VCC	HF	SINGLE NL
85	i	g	f	iff	B	VCC	HF	SINGLE NL
86	e		S	egs	B	VCC	HF	SINGLE NL
87		g b	f	ubf	B	VCC	LF	SINGLE NL
88	u	S	f	usf	B	VCC	LF	SINGLE NL
89	U O	f	S	ofs	B	VCC	LF	SINGLE_NE
90	i	1	S	its	B	VCC	LE	SINGLE_NL
		-			B	VCC	LF	SINGLE_NL
91	e	1	1	ett	B	VCC	LF	SINGLE_NL
92	0	g	+	ogt	B		LF	SINGLE_NL
93	0	b	g	obg	B	VCC	LF	
94	e	g	1	egt		VCC	LF	
95 96	a	b	g	abg itg	B	VCC	LF	SINGLE_NI



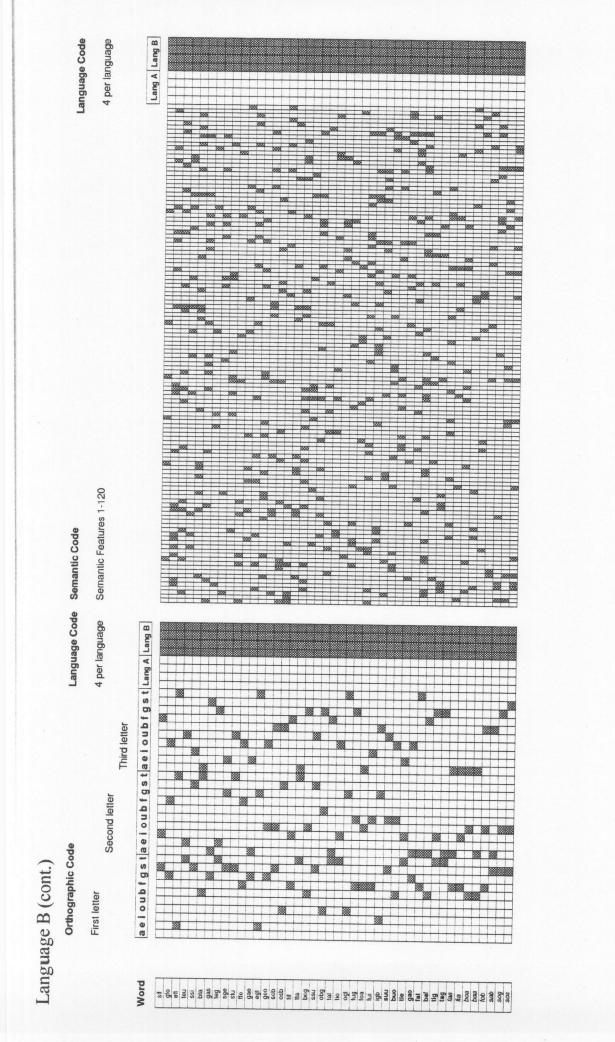




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Word



Appendix B: Experimental Stimulus Sets.

The following pages contain Experimental Stimulus sets for:

The Bilingual Priming Study.

The first Bilingual Language Switching Study.

The second Bilingual Language Switching Study.

Test stimuli used in the Bilingual Priming Study.

 Negative of log(mean probability of occurrence) over all gram
 Trigram
 Count
 Trigram
 <th 7rench 3.16 3.24 3.257 3.257 2.577 2.577 2.577 3.41 3.41 3.41 3.41 3.43 3.43 3.43 2.94 0.43 2.19 3.09 3.01 3.27 0.47 T Nonwords legal in both languages. (see Appendix C for explanation of these values). 2.94 3.21 2.11 2.00 0.17 0.18 2.17 0.27 2.18 0.19 ANTER BARET DESPATE DESPATE DEVIN FAND FAND FAND GODE HOUL LENCE MOIT MONE NUNCE PLUGE Mean Mean Imag 519 567 261 6667 6667 6600 6000 6000 6000 6000 6574 5779 5579 5579 5598 5598 5537 351 455 487 130 KF-Freq (of translation equivalent) 22 832 61 769 64 9 9 9 54 394 110 211 1016 2 44 1125 434 735 391 133 133 133 133 133 133 133 133 10 202 202 202 26 9 9 9 9 9 9 9 9 60 60 316 393 375 French Singles PLAGE IRES PROCHAIN VIDE PEINE CHER PROCHAIN MATIN AVANT SABOT SABOT SABOT FAIT COUT SOIR SOIR HAUT DUR GLACE HIVER HIVER HIVER CLAT CUAND BAIN MALADE NON TANTE Mean Mean Mean Imag 480 401 556 609 462 463 635 572 572 572 619 619 603 485 535 535 535 535 518 442 French meaning of Non-Cognate Homographs KF-Freq (of translation equivalent) 16 833 834 834 834 9 9 9 9 828 9328 9328 9489 22 22 3201 3001 733 3 410 202 202 89 89 11 11 11 11 11 11 2216 28 36 36 6397 63 1,021 2,383 877 2,099 229 BILLET CAR CAR CAMENT COMMENT DENT DENT FOND HATE LENT LOIN MINCE POUR RIDE SELZE TOUT CHOSE COIN COIN FIN FIN HABITS LIT HABITS LIT HABITS SABLE SABLE SALE SALE SALE Mean Mean Mean Imag 479 89 543 634 492 634 463 545 545 545 545 545 545 507 507 507 546 448 495 463 KF-Freq 48 66 17 33 33 33 33 33 195 31 37 3 49 48 59 59 88 43 English Singles MOUSE APRON TAIL HIDE BURN FORMER FORMER SPARE NINE GILT GILT BUTON BUTTON BUTLER GIRL GIRL PLAIN RULER ALOIE PIER PIER BRIBE ALONE SLICE SEVER CLOVE Mean Mean Mean Std Imag 476 65 464 470 KF-Freq 172 9 49 5 42 42 274 37 10 10 11 11 11 11 11 11 10 10 10 41 75 48 45 83 English meaning of Non-Cognate Homographs outh languages? Same WP in ZZ ZZZ zz Z ZZ CAR CHAT CHAT DENT FOND HATE LENT LOIN MINCE PAYS POUR RIDE RIDE RIDE RIDE COIN CRANE FORT FORT HABITS LIT HABITS LIT NAIN MAIN MAIN MAIN MURS PAUN SABLE SON BILLET CHOSE Mean Mean Std Overall Mean Group 2 Group 1 Group 1 Group 2

Imag = Paivio and Gilhooly-Logie Imagibility rating from Quinlan (1994). Trigram and Bigram frequencies from author's corpus counts. KEY: KF-Freq = Kucera and Francis (1967) frequency rating, in counts per million.

NB The trigram _OVE_ does not appear in the French corpus.

Test stimuli used in the first Bilingual Language Switching Study.

ar in the t.	Imag	543	634	492	463	543	378	488	545	541	425	480	507	202	439	457	446	200	100	615	200	221	004	430	140	210	010	424	000	117	538	RDF	580	001	50 ⁺	201	001	493	59	490	170	110	516	010
es to appea age contex	KF-Freq	4	220	9	48	3	17	48	3	-	1	195	31	57	e	53	-	30	00	2		50	00	15	121	101	81	63	30	202	184	60	10	AG 1	24	24	5	43	65	12	101	43	30	70
English Singles to appear in the correct language context.		BUTLER	GIRL	COIL	PLAIN	RULER	CLEVER	ANGER	PIER	DIAL	BRIBE	ALONE	SLICE	SPOT	SEVER	BILLER	CLOVE	IMPROVE	MOLISE	APRON	TAIL	CHAIN	HIDE	BLIBN	FORMER	SPARE	NINE	SAVE	GILT	DROP	SPACE	NOSE	BUTTON	<u> </u>								1		
S	Imag	436	638	453	453	486	363	462	556	000	413	204	001	403	410	490	430	299	603	562	551	559	418	542	309	289	502	347	419	422	560	610	582	475	06	485		477	63	473	473	111	522	
homograph ng).	KF-Freq	1	274	n S	42	2	13	42	0 +		170	0	01	40	77	+	_	37	10	2 2	2	55	23	17	119	29	88	72	2	44	166	66	10	45	61	20		44	2	1	47	45	33	
Non-cognate homographs (English reading).		BILLET	CAH	CHAI	DENIT	COND	LATE	IENT	ION	MINCH	PAVS	POLIR	RIDE	SFIZE	I AID	TUIT		CHOSE	COIN	CRANE	FIN	FORT	HABITS	LIT	MAIN	OURS	PAIN	PLUS	SABLE	SALE	SON	CHAIR	VENT	Mean	Std	Median		Mean	DIC	Median	Mean	Std	Median	1
		Group 1																Group 2																Both Gps				aroup 1			Group 2			

English Singles to appear in the

		1	1		1	-	1	-		-		-			-	-	-		1	-	T	1	1		-					-		-		_			-
	Imag	506	643	467	424	463	362	425	572	577	393	446	533	445	406	474	448	319	597	568	545	567	404	553	321	315	519	393	446	373	580	618	530	476	06	465	474
ge context.	KF-Freq	1	167	4	42	10	18	. 65	5	-	-	177	12	30	S	58	0	33	12	13	3	33	24	10	98	29	95	61	-	46	148	70	11	40	48	21	37
wrong language		GUTTER	ISLAND	FOUL	PRIDE	UNE	A	TASTE	OTTER	SNAIL	TOIL	LIVES	BOIL	SPLIT	HAUL	SAFE	CUE	SCHEME	SAIL	BLADE	PIER	BONE	BORE	BITE	CLAIM	FALSE	-	CLOSER	SPOUT	TREND	HAIR	RAIN	TIDE				L

465	474	73	456	478	105	525
21	37	55	11	43	40	31

Cognate homographs to appear in an English context.

	Imag	498	590	470	427	481	359	471	537	588	423	513	502	470	455	459	477	341	601	564	585	578	467	557	354	324	526	395	454	391	610	564	588	488	81	479	483	56	474	494	UUF		246
JIIIEAL.	KF-Freq	2	267	9	24	9	6	34	5	3	3	191	11	52	3	76	2	33	6	16	6	39	28	17	131	33	86	55	2	53	143	46	11	44	60	21	43	74	8	44	41	33	3
		BARON	COLLEGE	MOLECULE	BRAVE	BRUTE	REGRET	CRIME	MAT	CANAL	CASTE	NATURE	FATIGUE	SILENCE	AMUSE	REGION	FABLE	EXAMINE	CAVE	LECTURE	CAGE	CONCERT	PASSION	DOME	CHANCE	PRECISE	PATIENT	THEME	шı	CONTENT	SQUARE	AVENUE	BHONZE		1		 			-			L

Cognate homographs to appear in a French context.

		1		1	T	1	T	1	1	T	T	1	1	-	-	-		-																		
Imag	587	249	606	297	630	491	370	340	399	518	578	266	571	473	531	341	0.00	356	555	599	481	440	626	575	582	578	587	334	619	501	298	593	55	A75	C/+	1 001
KF-Freq	35	373	33	91	38	497	12	40	134	85	72	57	2	53	15	369	FUC	504	611	4	395	204	23	68	198	10	37	246	33	103	148	82	99	111	001	
	H	POSSIBLE	LAINCE	NOILION	PIANO	£	EPISODE	TENS		CAPITAL	VILLAGE	APPARENT	BISCUIT	MINUTE		IMPORTANT	EVIDENCE	ALIDIENICE	BOIDIFT	BOWEL	INIOL	MILLION	CHANGE	ANIMAL	TIME	I UNNEL	MALE	MOMENT	GHAVE	DALE	DIFFERENCE	IHAIN	PAGE			

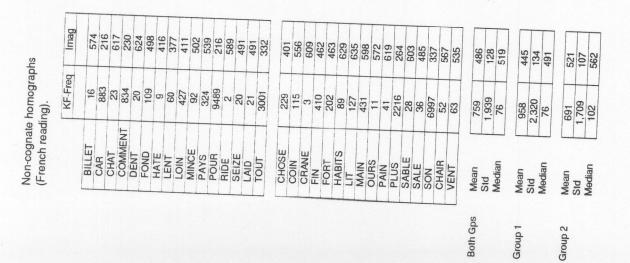
017	139	525	453	125	482	486	151	565	The second secon
	123	67	119	147	55	121	103	93	T

French Singles to appear in the wrong language context. KE E.

Imag	570	37.4	ROA	367	597	495	448	334	387	501	544	253	613	477	506	352	383	576	613	425	433	613	606	909	592	595	262	607	522	338	593	597
KF-Freq	-	1			22			1			1	1					162	160	8	319	283	22	95	591	6	34	4393	72	37	1171	94	70
	SAC	BON	MENTON	DERNIER	POULE	BORD	VOLER	UILE	IAHD	HUI IN	PLANCHER	LAS	reche	JEIEH	SAUT	COMME	 NIOS	MUH	CHEVILLE	OUVERT	AGIN	חות	MEH	MAISON	CHOMAGE	SOUTE	INDO	CULLINE	LAVER	MEME	DIAO	PIEU

497	117	533	464	106	486	523	112	593
462	1,035	75	508	1,117	68	470	1,053	95

r in the	Imag	567	261	667	243	600	479	429	374	386	495	579	255	598	537	517	351		383	559	599	463	460	635	621	593	596	638	231	601	456	361	608	200	497	124	542	459	130	487	FOA	111	C11	689
es to appear age context.	KF-Freq	22	832	61	769	22	64	6	54	394	110	211	1016	2	44	28	1125		100	133	10	497	202	45	83	424	6	49	2331	26	51	1022	20	00	362	592	63	298		63		716	+	-
French Singles to correct language		TANTE		PLAGE	INES	SAVON	VIDE	PEINE	1:	PROCHAIN		MAIN	AVAN		SENTIER	PRIX	FAIT	TINC	SOIR	POLICE	HALIT		HUU	GLACE	TLTT		CAN	CAL	DAINI	MALADE	· · · -	PONT	LUNE											



Test stimuli used in the second Bilingual Language Switching Study.

becific	Imag	617	595	588	581	581	529	529	490	521	255	443	383	181	4/1	180	+00	632	594	599	585	585	551	532	506	475	298	451	371	390	495	574	624	514	97	531	512	100	529	210	CR CR	542
cally Non-sp jles	KF-Freq	17	50	9	63	162	286	10	14	123	1016	88	162	51	60	187	-	18	52	7	58	197	130	2	21	109	308	102	174	143	72	54	117	127	178	87	157	235	87	20	21	01
Orthographically Non-specific English Singles		NURSE	ENGINE	NAIL	RIFLE	PICTURE	FIVE	PASTE	RIPE	GAME	BEFORE	COVER	CARE			FIRE		LEMON	GOLD	OVEN	STONE	RED	HOT	LICE	MATE	INCOME	MATTER	HISE	INSIDE	IRADE	MASIEH	SEAL	HORSE	Mean	Std	Median	Mean	Std	Median	Mean	DIO	Mediari
		Group 1																Group 2																Both Gps			Group 1	-		2 dnoin		

Orthographically Non-specific French Singles

KF-Freq 9 32 33 160 686 173 173 173 173 173 173 173 173 173 173
--

										-						
638	607	595	580	574	576	511	541	542	498	247	446	344	420	480	612	635
274	13	28	10	60	139	808	15	7	277	552	51	129	234	195	75	60
VOITURE	RENARD	NUAGE	BOUTON	GRAS	SELON	ETAT	BRULER	BASSIN	MORT	CEPENDENT	TIRER	TORT	FERMER	SEUL	ROCHER	JARDIN

510	109	534	508	115	526	513	102	542
186	251	79	206	282	66	166	215	68

Orthographically Specific English Singles

512	106	533	515	104	533	508	108	533
170	340	72	141	164	94	199	450	64

Orthographically Specific French Singles

	KF-Freq	Imag
GATEAU	13	624
Z	60	605
SEAU	7	586
E E	57	570
IEVEU	148	580
UNE	385	521
DLEUR	8	529
NE	12	510
LIGNE	298	491
UR	1069	264
EU	318	461
IAI	231	365
NGUEUR	116	395
UF	81	494
VERRE	66	585
AU	47	638

614	597	614	588	569	518	511	500	489	278	448	324	385	493	585	639
31	59	7	39	143	255	4	7	226	750	130	157	213	70	88	112
OISEAU	NEIGE	AGNEAU	MEUBLE	BLEU	BUREAU	LUEUR	PROIE	VOIX	auoi	MOIS	PAREIL	NIVEAU	DOUX	ROI	SOLEIL

6	10				
6	52	510	103	515	
256	90	143	175	100	
		256 99 90 525			

Orthographically Specific French Singles

Orthographically Specific English Singles K-Fred

Imag

K-Freq

Orthographically Non-specific French Singles

mag	20	00	14	23	33	26	2	267		488	1	0
5	9	9 9			48	45	31	26		48	131	510
hail-u	121	59	15	212	967	155	53	312		237	290	128
	BLOOD	TREE	SHOWER	TOWN	BACK	GROWTH	SLIGHT	THUS	L			

624 612 611 554 458 336 339 286 286

117 11 11 105 760 380 380 438

CHEVAL NAVIRE LAPIN GARE TRAVAIL COTE PIRE MOINS

484 126 506

243 244 111

Singles presented in the wrong language context.

Non-specific	
aphically	English Singles

Imag	8	2	0	7	6	0	4			0	
E	61	61	61	54	48	42	33	368	501	108	518
K-Freq	134	58	38	59	226	707	81	368	209	215	108
	SUMMER	SMILE	SISTER	NOVEL	VOICE	UNDER	EVENT	SENSE	Mean	Std	Median

Appendix C: Orthographic Characteristics.

The following pages contain an explanation of the orthographic characteristics measure, and then measurements for the experimental stimuli. These are initially summarised for all the experiments, and then shown individually for:

The Bilingual Priming Study.

Language Switching Study 1.

Language Switching Study 2.

Orthographic characteristics of Experimental Stimuli.

The following pages contain tables listing the stimuli used in the empirical studies, along with values indicating how characteristic they are of the English and French language. The values indicate the log of the mean probability of occurrence of the bigrams / trigrams comprising each item, according to English and French corpus counts. All bigram and trigram log probabilities are negative, since they are less than one. However, for convenience, they are shown in the tables as positive values.

A frequently occurring bigram or trigram will generate a small negative log probability - thus the higher the value, the rarer the letter string.

In the difference column:

- A positive number corresponds to a string that comprises grams more probable in French.
- A negative number corresponds to a string that contains grams more probable in English.

The corpuses were obtained from a set compiled on the Oxford University Central Computing Services, available by anonymous ftp at ermine@ox.ac.uk, in the /languages/ directory. The gram probabilities are not scaled according to frequency of usage of the words in spoken or written language.

Example 1: VOICE.

The word VOICE decomposes into the bigrams VO, OI, IC, and CE; and to the trigrams VOI, OIC, and ICE. The probabilities of these grams occurring in each language are as follows:

English	French
0.000967	0.001654
0.001369	0.002771
0.009640	0.002998
0.005205	0.004856
0.000071	0.000380
0.000102	0.000005
0.000902	0.000643
	0.000967 0.001369 0.009640 0.005205 0.000071 0.000102

The mean bigram probability is thus 0.004295 in English and 0.003070 in French. The mean trigram probability is 0.000358 in English and 0.000343 in French.

The negatives of the logs of these values gives the following indexes for this word.

Bigrams		Trigrams		Difference	
Mean English	Mean French	Mean English	Mean French	Bigram	Trigram
2.37	2.51	3.45	3.46	-0.15	-0.02

This word has a fairly similar characteristics in French and English.

Example 2: POOR.

The word POOR decomposes into the bigrams PO, OO, and OR; and into the trigrams POO and OOR. The probabilities of these grams occurring in each language are as follows:

	English	French
PO	0.004113	0.003768
00	0.003205	0.000153
OR	0.010666	0.006237
POO	0.000173	0.000005
OOR	0.000157	0.000032

The mean bigram probability is thus 0.005995 in English and 0.003386 in French. The mean trigram probability is 0.000165 in English and 0.000019 in French.

The negatives of the logs of these values gives the following indexes for this word.

Bigrams		Trigrams		Difference	
Mean English	Mean French	Mean English	Mean French	Bigram	Trigram
2.22	2.47	3.78	4.73	-0.25	-0.95

This word is much more characteristic of English than of French. Note that not all words that appear characteristic of just one language will necessarily have this fact reflected in their bigram and trigram stats: these differences may only arise in higher order statistics.

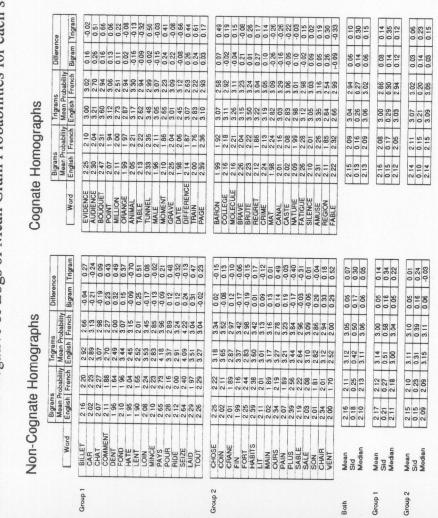
Bilingual Priming Study.

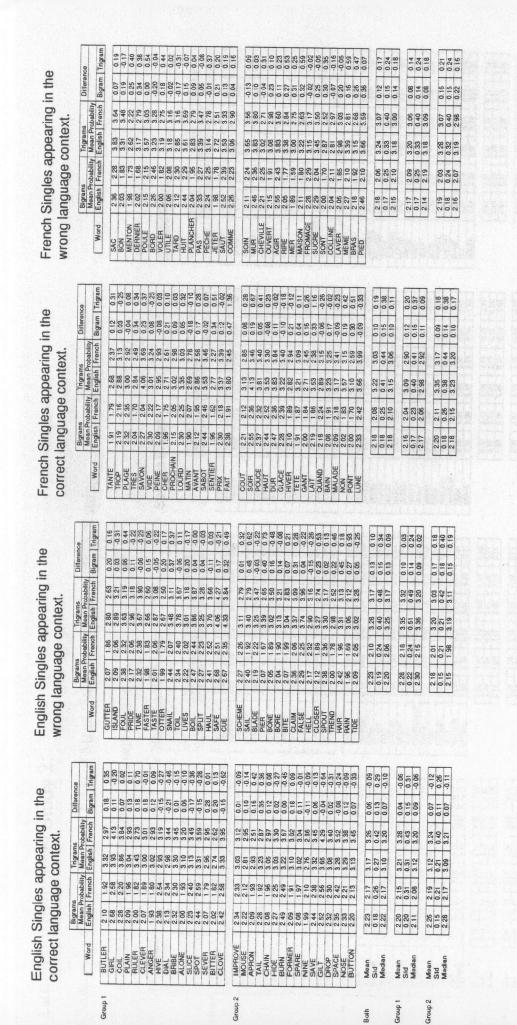
Table showing the orthographic characteristics in English and French, for the filler stimuli in the Bilingual Priming Study. Values show the negative of the logs of the mean gram probabilities of occurrence averaged over all the grams in each string, and over all strings. The larger the number, the rarer the components of a string.

		Mea	an Bigram	Frequence		Mear	n Trigram	Frequenc	ies
		English Mean	Corpus SD	French Mean	Corpus SD	English _{Mean}	Corpus SD	French	Corpus
Words	English French	2.17 2.15	0.17 0.15	2.12 2.04	0.25 0.23	3.11 3.18	0.33 0.32	3.19 3.04	0.44 0.41
Nonword	ds	2.17	0.33	2.07	0.23	3.14	0.44	3.10	0.55

Language Switching Study 1.

Scores show the Negative of Logs of Mean Gram Probabilities for each string.





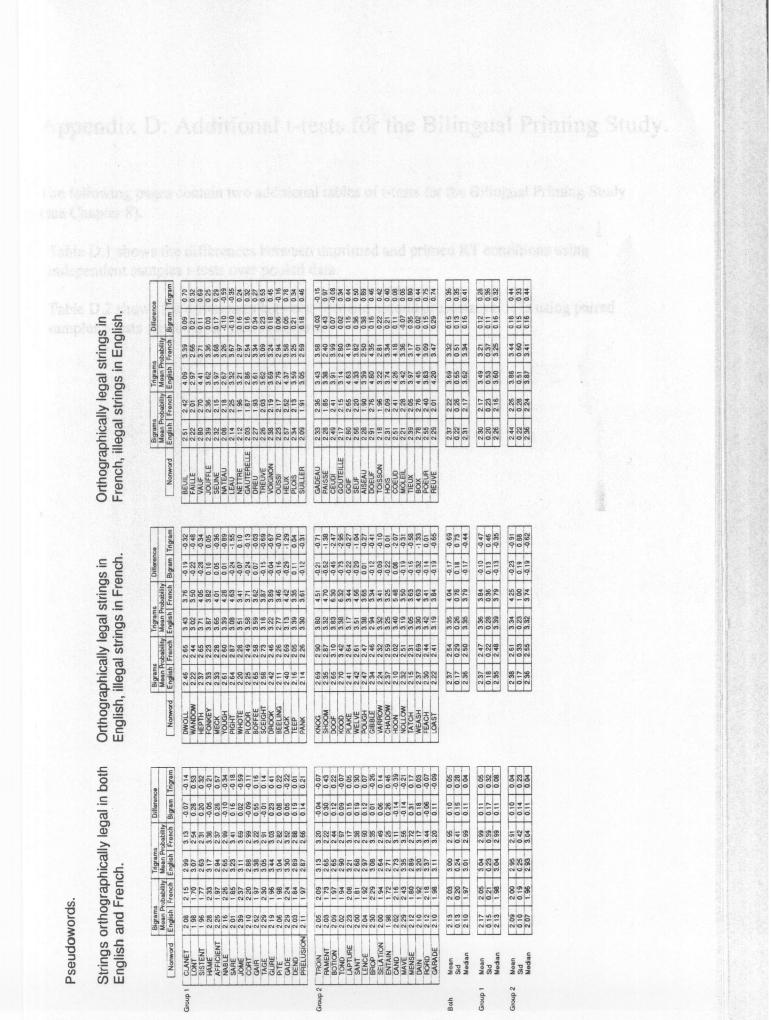
Language Switching Study 2.

Scores show the Negative of Logs of Mean Gram Probabilities for each string.

Singles presented in the Correct Language Context.

Specific French	Bigrams Mean Probability Word English French English French Bigram Trigram	GATEAU 211 221 271 331 -010 -061 NEZ 240 202 453 351 010 -061 SNE 231 221 353 351 010 -061 SNE 231 231 359 366 010 -038 112 SNE 248 182 445 247 359 361 001 031 CHEVEL 247 223 337 238 006 028 UONE 227 223 337 238 049 030 VOLEUR 227 223 337 238 049 030 VOLEUR 231 227 338 304 056 039 CIONE 231 227 338 304 056 039 VOLEUR 230 203 366 049 056 039 VOLEUR 231 227 338 340 056	OISEAU 2.31 2.08 3.88 2.78 0.23 0.59 MEIGE 2.37 2.32 3.61 3.48 0.05 0.13 AGNELU 2.41 2.42 3.58 3.56 0.05 0.13 AGNELU 2.41 2.42 3.58 3.56 0.05 0.13 AGNELU 2.17 2.26 3.58 3.69 0.16 0.05 MEUBLE 2.17 2.26 2.69 3.69 0.16 0.05 BLEU 2.51 2.30 3.67 3.03 0.03 0.07 UCHR 2.54 2.07 3.06 3.08 3.07 0.05 PIDEL 2.57 2.89 3.03 3.01 0.05 0.16 PIDEL 2.34 2.07 3.03 0.27 0.65 PIDEL 2.34 2.07 3.17 3.02 0.16 VXK 3.65 2.64 3.71 3.02 0.16	237 221 357 324 0.17 0.33 020 023 0.49 0.41 0.19 0.50 234 221 355 330 0.13 0.26	234 214 367 319 021 048 015 023 052 046 023 061 233 216 355 318 013 036	2 40 2 3 46 3 0 13 0 17 0 24 0 22 0 44 0 3 0 13 0 17 0 24 0 22 0 44 0 3 0 0 3 0 13 0 17 2 34 2 25 3 8 3 3 0 0 13 0 14
Specific English	Bigrams Trigrams Difference Mean Probability Mean Probability Mean Probability Word English French English French Bigram Trigram	AFROM 217 227 325 323 0.10 0.07 UAKE 2.37 266 377 36.0 29 -165 SUOAK 2.17 2.37 266 377 36.0 29 -165 SUOAK 2.17 2.26 3.17 3.75 0.10 0.058 COAST 2.17 2.26 3.17 3.75 0.10 0.658 COAST 2.17 2.26 3.17 3.75 0.10 0.61 COAST 2.17 2.26 3.17 3.75 0.10 0.071 UchT 2.46 2.97 3.01 3.82 0.10 0.011 0.081 BLCOM 2.33 2.60 3.61 4.16 0.22 0.055 0.055 BLCOM 2.33 2.11 3.61 4.16 0.22 0.055 0.055 BLCOM 2.28 3.71 3.76 3.71 0.22 0.055 0.055 0.055	FLOWER 2.21 2.01 3.35 4.34 0.20 0.99 RhOok 2.35 2.36 3.35 4.34 0.20 0.99 RhOok 2.40 3.85 5.82 0.06 1.94 MOON 2.12 2.06 3.84 5.72 0.12 2.06 MOON 2.12 2.05 3.84 5.72 0.12 2.06 MOON 2.12 2.06 3.84 5.72 0.12 2.06 MOON 2.12 2.06 3.84 5.04 0.06 1.74 MOON 2.12 2.06 3.83 3.14 0.15 0.15 RECH 2.21 2.94 3.26 3.26 0.05 0.71 HEAV 2.45 2.74 3.42 0.05 0.76 MAR 2.33 2.93 3.9 0.16 0.70 WAR 2.33 2.94 3.77 4.30 0.70 WAR 2.33	232 242 336 4.15 0.11 0.79 0.16 0.30 0.22 0.77 0.19 0.69 2.29 2.44 3.36 3.31 -0.10 -0.68	236 251 340 4.13 -0.14 -0.73 0.18 0.31 0.23 0.68 0.19 0.61 235 252 3.40 3.66 -0.11 -0.53	2 28 2.34 3.31 4.17 -0.07 -0.86 0.19 0.75 0.82 0.19 0.75 2.23 2.33 4.10 -0.08 -0.75 2.75 2.23 3.33 4.10 -0.08 -0.70 -0.86 -0.75 -0.75 -0.75 -0.75 -0.75 -0.75 -0.75 -0.75 -0.75 -0.75 -0.76 <t< th=""></t<>
Non-Specific French	Bigams Trigrams Difference Mean Probability Mean Probability Mean Probability Word English French Bigram	CANARD 206 2.22 2.96 3.28 0.16 -0.32 USINE 201 214 3.07 3.05 0.04 BERGER 201 214 3.07 3.06 0.04 BOLIMAT 2.56 2.49 3.43 3.73 0.23 0.01 BOLIMAT 2.56 2.49 3.43 3.73 0.23 0.31 BOLIMAT 2.56 2.49 3.43 3.73 0.23 0.31 BOLIMAT 2.56 2.93 3.63 0.53 0.32 0.31 URE 2.33 0.14 0.78 0.38 0.36 0.32 URE 2.39 2.87 0.26 0.32 0.49 LIFE 2.97 2.89 2.87 0.26 0.23 LIFE 2.97 2.89 2.87 0.26 0.23 CLAIK 2.97 2.89 2.86 0.26 0.25 CLIFE 2.97 2.98	REMARD 202 1 86 2 97 2 76 0 15 0 21 NUAGE 2 556 2 55 3 28 3 12 0 01 0 16 NUAGE 2 15 2 55 3 28 3 12 0 01 0 16 RAS 2 18 1 92 2 95 2 56 0 15 0 17 GAS 2 18 1 92 2 95 2 162 0 16 0 07 GAS 2 18 1 92 2 93 3 22 0 17 0 16 0 16 RELM 2 04 2 91 3 06 2 29 0 16 0 17 0 16 BASIN 2 18 2 18 3 47 2 83 0 17 0 16 0 19 MORT 2 18 3 06 2 91 3 00 0 20 0 09 0 16 MASIN 2 18 3 06 2 93 0 17 0 16 0 16 0 16 MASIN 2 18 3 16 2 93 0 17 0 105 0 10 0 10	2.15 2.02 3.15 3.00 0.13 0.16 0.16 0.24 0.33 0.34 0.20 0.27 2.11 1.96 3.07 2.93 0.16 0.15	216 1.96 315 2.99 0.18 0.16 0.17 0.22 0.32 0.39 0.20 0.25 2.12 1.96 3.06 2.91 0.25 0.19	2.13 2.06 3.16 3.00 0.07 0.15 0.15 0.25 0.34 0.28 0.18 0.29 2.11 2.01 3.07 2.99 0.14 0.13
Non-Specific English	Bigrams Trigrams Difference Mean Probability Mean Probability Mean Probability Word English French English	Group 1 NURSE 2.40 2.20 3.43 3.24 0.20 0.19 MAIL 2.30 2.01 2.03 2.01 2.03 2.01 0.01 0.08 MAIL 2.31 1.55 3.57 0.01 0.06 0.23 NAIL 2.33 2.10 3.66 3.85 0.06 0.23 PICTURE 2.00 2.10 3.12 3.12 0.10 0.06 PICTURE 2.46 2.10 3.12 3.07 0.01 0.00 FIVE 2.43 2.66 2.94 2.96 2.91 3.07 0.01 0.06 FIPE 2.22 2.16 2.93 3.31 3.35 0.01 0.06 BEFORE 2.06 1.84 2.96 3.33 0.11 0.02 0.01 0.06 COVER 2.96 1.84 2.96 3.33 0.11 0.02 0.01 0.06 COVER 2.93 3.33	Group 2 LEMON 2.03 1.91 3.13 3.13 0.12 0.00 OVEN 2.15 2.82 3.93 3.07 3.03 3.07 0.03 0.85 OVEN 2.15 1.99 1.95 2.79 3.06 3.07 0.72 0.03 0.48 FED 2.01 1.69 1.95 2.73 3.55 3.76 0.75 0.15 0.31 0.15 0.15 PED 2.07 1.69 2.73 3.55 3.76 0.75 0.05 0.06 MATE 1.27 2.203 3.95 3.76 0.75 0.05 0.06 MATE 1.97 2.09 2.09 2.09 2.06 0.06 0.01 0.05 0.06 MATE 2.09 2.09 2.04 2.93 2.94 0.05 0.06 0.01 0.05 0.06 0.01 0.05 0.06 0.06 0.01 0.05 0.06 0.06 0.01	Both Mean 215 210 3.08 3.22 0.05 -0.15 Std 0.16 0.26 0.30 0.40 0.17 0.33 Median 2.11 2.06 3.06 3.16 0.66 -0.15	Group 1 Mean 2.20 2.12 3.16 3.24 0.08 -0.08 Std 0.16 0.20 0.30 0.24 0.14 0.26 Median 2.20 2.10 3.21 3.20 0.07 -0.13	Group 2 Mean 2 10 2 08 2 99 3 21 0 02 -0 22 Sid 0.15 0.30 0.28 0.51 0.16 0.37 Median 2 07 2.00 3.00 3.14 0.04 -0.19

Specific French	Anti- metric Tranum T
Specific English	Bigeruns Bigeruns Word English French Word English French ThEE 2 11 134 2 51 ShOwch 2 33 2 56 2 51 BACK 2 33 2 51 2 51 ShOwch 2 33 2 55 2 56 Median 2 33 2 52 2 55 Median 2 33 2 52 2 55
Non-specific French	Biganit Triganit Triganit Triganit Dilietence Word English Frendit Man Probability Man Probability Biganit Triganit Munic 2.28 2.36 2.35 2.36 0.17 0.33 Munic 2.28 2.93 3.10 2.93 0.11 0.33 Munic 2.02 2.03 0.11 0.33 0.11 0.33 Munic 2.02 2.03 3.10 2.03 0.11 0.33 Munic 2.01 2.01 2.01 0.33 0.11 0.33 Munic 2.01 2.01 2.01 0.01 0.01 0.02 Munic 2.01 2.01 0.01 0.02 0.03 0.01 Munic 2.01 2.01 2.01 0.01 0.02 0.02 Munic 2.01 2.01 0.01 0.01 0.02 0.01 Munic 2.01 2.01 0.01 0.01
Non-specific English	Bigame Togame Togame Dilence Word Exglish Tender English Tender Dilence Sum <probability< td=""> Mean Probability Mean Probability Mean Probability Bigan Tigame Simula 218 1.34 3.22 3.18 0.24 0.12 Simula 1.32 2.51 3.45 3.17 0.18 0.07 Nortic 2.37 2.51 3.45 3.17 0.19 0.07 Nortic 2.37 2.01 1.81 2.63 2.61 0.02 Work 2.16 2.11 1.177 3.09 0.01 0.01 Mean 2.16 1.88 3.15 3.09 0.00 0.00 Mean 2.16 1.88 3.15 3.09 0.00 0.00 Mean 2.16 0.33 0.00 0.00 0.00 0.00 Mean 2.16 1.88 3.15 3.09 0.00 0.00 </probability<>



Appendix D: Additional t-tests for the Bilingual Priming Study.

The following pages contain two additional tables of t-tests for the Bilingual Priming Study (see Chapter 8).

Table D.1 shows the differences between unprimed and primed RT conditions using independent samples t-tests over pooled data.

Table D.2 shows the differences between first and second presentation RTs using paired samples t-tests for subjects' individual data.

disadvantage of confusing inter-subject variability with variability in the size of the priming effect (only the latter is Table D.1: Results of the Cross-language priming study. Scores show the mean response time in msecs in the lexical decision across subjects. Mean unprimed and primed RTs were compared using independent samples t-tests. This has the task for a stimulus in the identical context, when unprimed and when primed. Responses are only included for (1st presentation - 2nd presentation) pairs that were both correct. Responses for each word type were pooled of interest here).

			Presentati	Presentation Context	Response Time (ms)	Time (ms)				
Stimulus Type	Example	Repetition Type	Prime	Target	Unprimed	Primed	Difference	S Error	SError t-test p-val p<0.05?	p<0.05?
Non-Cognate Homographs	MAIN	Within language	English	English	692	605	88	37	0.018	sig
		Between language	English	French	750	661	89	47	0.057	0
		Between language	French	English	692	742	-50	53	0.350	
		Within language	French	French	750	608	142	35	0.000	sig
English Singles	RAIN	Within language	English	English	630	574	56	26	0.031	sig
		Between language	English	French	899	606		55	0.844	
		Between language	French	English	630	646	-16	30	0.589	
		Within language	French	French	899	803	96	, 46	0.039	sig
French Singles	BAIN	Within language	English	English	807	748	59	34	0.081	
		Between language	English	French	699	617	53	. 25	0.035	sig
		Between language	French	English	807	834	-28	48	0.571	
		Within language	French	French	699	574	95	. 22	0.000	sig
Nonwords	SARE	Within language	English	English	744	696	49	37	0.191	
		Between language	English	French	720	832	-112	38	0.004	sig
		Between language	French	English	744	645	100	30	0.001	sig
		Within language	French	French	721	724	ကု	28	0.911	

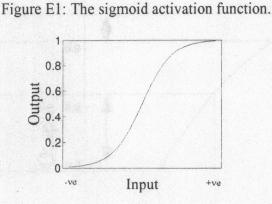
mean RTs for each subject were tested using a paired samples t-test. This has the advantage of factoring out inter-subject between language repetition conditions, there might appear to be a priming effect merely because the subject is faster in variability. However, it confuses priming effects with differences in base rate performance between context. Thus in Table D.2: Results of the Cross-language priming study. Scores show the mean response time in msecs in the lexical decision task for the first and second presentation of a stimulus, when both responses were correct. The difference between the second language than the first.

			Presentati	Presentation Context	Response Time (ms)	Time (ms)				
Stimulus Type	Example	Repetition Type	Prime	Target	1st Pres.	1st Pres. 2nd Pres.	Difference	S Error	s Error t-test p-val p<0.057	p<0.05?
Non-Cognate Homographs	MAIN	Within language	English	English	742	605	138	38	0.001	sig
		Between language	English	French	647	661	-15	36	0.691	
		Between language	French	English	793	742	51	55	0.355	
		Within language	French	French	717	608	109	26	0.000	sig
English Singles	RAIN	Within language	English	English	640	574	99	19	0.001	sig
		Between language	English	French	617	606	-293 ·	37	0.000	sig
		Between language	French	English	939	646	292	48	0.000	sig
		Within language	French	French	862	803	59	38	0.132	
French Singles	BAIN	Within language	English	English	789	748	41	20	0.046	sig
		Between language	English	French	827	617	210	34	0.000	sig
		Between language	French	English	660	834	-174	53	0.003	sig
		Within language	French	French	674	574	100	16	0.000	sig
Nonwords	SARE	Within language	English	English	754	696	58	23	0.014	sig
		Between language	English	French	734	832	66-	28	0.001	sig
		Between language	French	English	678	645	34	23	0.156	
	i i i	Within language	French	French	758	724	34	31	0.273	

Appendix E.

The frequency-by-priming interaction in 3-layer networks.

Hidden units and output units in 3-layer networks conventionally employ a sigmoid activation function (see Fig. E1) to convert the net input activation arriving at a given unit to its output activation state. Net input activations of large magnitudes cause the unit's resulting activation state to tend towards 0 for negative values and 1 for positive values.



For poorly trained mappings (e.g. low frequency words), the extra training involved in priming can move unit activations from intermediate values towards 0 or 1 as required by the mapping. For well trained mappings (e.g. high frequency words), unit activations will already be close to 0 or 1. Extra training will generally only serve to increase the magnitude of the net input activation to a unit, in an attempt to push it ever closer to 0 or 1. The increase in performance from a given number of training cycles will be less for well trained mappings than poorly trained mappings. This is the ceiling effect.

Fig. E.2 illustrates a frequency-by-priming interaction arising out of applying the asymptoting output activation function to the additive input contributions of frequency and priming. The identical contribution of priming at input has a much weaker effect on high frequency words than on low frequency words. Only the top half of the sigmoid function is shown in this figure.

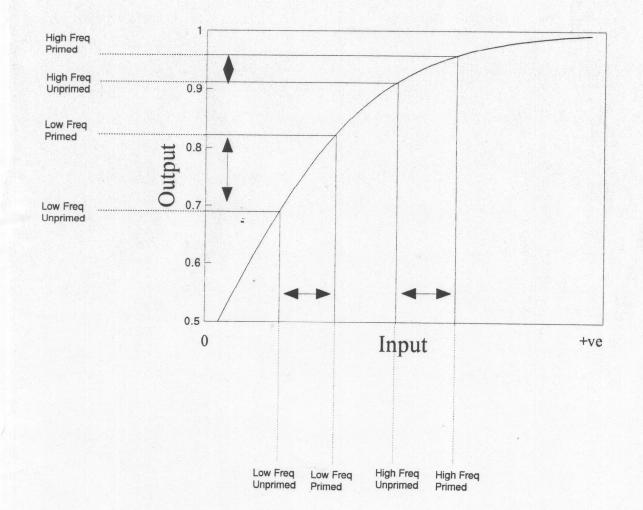


Figure E2: An example of the frequency by priming interaction for a unit involved in word recognition (after Plaut et al, 1996).