Mutual information and corpus-based approaches to English reduplication Shih-ping Wang Ming Chuan University and University of Nottingham <u>spwang@mcu.edu.tw</u>

The aim of this project is to use corpus-based approaches to investigate reduplicative fixed expressions in English, e.g., *sooner or later, first and foremost, part and parcel*, etc. The probabilistic relations between two adjacent words were examined. A reduplication corpus has been constructed and the frequency of each token was calculated based on its occurrence in the British National Corpus (BNC). Then a word list with 116 items was proposed for related research in terms of SARA software, the built-in tool of BNC. Frequency is often considered as the main factor to decide the word order in a conjoined phrase (Fenk-Oczlon, 1989, 2001), but the frequency-based argument is shaky and not always reliable. Mutual information (MI) was therefore employed to calculate the probability of collocation and assess collocational significance. MI can be used to decide what to look for in a concordance (Church and Hanks, 1990). The higher the mutual information, the more genuine the association between two words.

Keywords: BNC, corpus, reduplication, frequency, mutual information, and z-score

1. INTRODUCTION

The aim of this project, grant-funded by the National Science Council (NSC) is to use corpus-based approaches to investigate reduplicative fixed expressions in English, e.g., *sooner or later*, *first and foremost*, *part and parcel*, etc. Reduplication is important in language studies. Its word order at the phrasal level is explored in the present study.

A reduplication corpus (1,700 items) had been accumulated. The frequency of each token was calculated based on its occurrence in the British National Corpus (BNC). A reduplicative wordlist with 232 items was established based on the self-constructed corpus. Then a questionnaire with 116 items, extracted from the wordlist, was proposed for related searches using the built-in SARA software.

Frequency is important in creating a word list. It is also considered as the main factor to decide which item goes first in a conjoined phrase (Fenk-Oczlon, 1989, 2001). According to Fenk-Oczlon's rule, high frequency comes before low frequency in binomials. However, frequency is not the only criterion to decide the word order in fixed expressions. The probabilistic relations between two adjacent words should be examined as well. Mutual information (MI) and z-score are therefore employed to assess collocational significance (Church and Hanks, 1990). Both MI and z-score may provide useful insights into direction of collocability. Therefore the primary question of current researches is to explore fixed reduplications and collocational significance. Three aspects, *frequency*, *MI* and *z-score*, are discussed to evaluate which method is more appropriate to decide the word order in binomials and how they shed new light on word order and collocational analysis in terms of corpus-based approaches.

2. LITERATURE REVIEW

2.1 Multiword units and reduplicative fixed expressions

In the research of word usage, linguists have recently turned their attention to multiword units (MWUs), which are strings of words acting as a unitary lexical item with a single meaning (Carter, 1998; Moon, 1998). MWU includes compound words, phrasal verbs, fixed phrases, idioms and proverbs (Schmitt, 2000, pp. 99-100). Freezes or fixed expressions consist of irreversible conjoined phrases and fixed reduplicatives (Pinker and Birdsong, 1979). Reduplicative MWUs can be compounds, fixed expressions and other types, e.g., *first and foremost* (248)¹, *deaf and dumb* (276), *bits and bobs* (49) and the like.

Reduplicated word-formation varies in English; examples consist of ablaut and rhyming terms: 'i- α ' (riprap), 'i-o' (ping-pong), *super-duper* and *hocus-pocus*. Terminology to describe the phenomenon also shows a discrepancy and includes: *fixed expressions, freezes, binomials* and *frozen locutions* (Pinker and Birdsong, 1979; McCarthy, 1990; Landsberg, 1995; Moon, 1998). Binomials or trinomials are usually irreversible combinations with other conjunctions whose order may be different from language to language, e.g., *sooner or later (503)*² (McCarthy, 1998, 130-131).

Basically fixed expressions can be divided into the concise formal types as shown in Table 1 (Carter, 1998; Moon, 1998):

Types of freezes	Examples	
irreversible conjoined phrases	wear and tear	
	hook, line and sinker	
	first and foremost	
fixed reduplicatives		
• vowel alternations (ablaut)	pitter-patter	
	ping-pong	
• rhyming terms	super-duper	
	razzle-dazzle	
	hocus-pocus	

Table 1 Formal types of fixed expressions

2.2 Freezes, word order and frequency-based arguments

Fenk-Oczlon (1989, 2001) argued that the frequency-based approaches can be simply used to solve some old questions. A new rule was proposed for the decision of word order in freezes: *high frequency before low frequency*, which implies 'more frequent tokens come before less frequent ones.' For example, the frequency of 'plus' (7,767) in the fixed expression, 'plus or minus' is higher than that of the second one, 'minus' (1,776); the frequency of occurrences for their collocation is 66. According to Fenk-Oczlon, the new rule achieves the highest accuracy with 84% correct predictions in his corpus. Frequency is considered

¹ The Arabic numerals mean the frequency of each token in the British National Corpus.

² In Mandarin Chinese the word order is opposite, i.e., *chi-tsao* ('late-early') \rightarrow 'sooner or later'.

as the main factor to decide which item goes first in a conjoined phrase (Landsberg, 1995). However, *is frequency the only criterion to decide the word order in fixed expressions*? It is often maintained that the probabilistic relations between two adjacent words should be considered as well when dealing with the fixed expressions (Moon, 1998; Schmitt, 2000).

2.3 Mutual information, z-score and frequency in the British National Corpus

There are three major kinds of scores frequently proposed to assess the collocational significance of each co-occurrence: i.e., mutual information (MI) score, t-score and z-score (McEnery and Wilson, 1996; Hunston, 2002). MI-score is probably the best known among them. T-score and z-score are most similar in terms of how they are calculated. MI and z-score are the two formulae mostly used to calculate the relationship of significant collocations. MI-score and z-score can be calculated using the SARA software, a built-in tool in the BNC, by which the z-score is generally recommended. Therefore, only both *MI* and *Z*-scores together with *frequency* will be disussed in the current research.

MI is a measure of the strength of collocation, provides a summary of what company words keep and thus is used for assessing collocational significance (Aston and Burnard, 1998). The higher the MI score, the more genuine the association between two words (Oakes, 1998), which can be calculated in terms of the following formula (Church and Hanks, 1990; Stubbs, 1995):

I = log2 ((f(n, c) x N) / (f(n) x f(c)))

- I = MI = mutual information; n = node; c = collocate
- f(n, c) is the collocation frequency
- f(n) is the frequency of node word (the query focus)
- f(c) is the frequency of the collocate
- N is the number of words in the corpus (corpus size):
 - If I (n; c) > 3, then the pairs tend to be significant or 'interesting'³.
 - If I (n; c) \sim 0, then the pairs are less interesting.
 - If I (n; c) \leq 0, then x and y are in complementary distribution.

The z-score is the number of standard deviations from the mean frequency. It is used to measure how likely it is that the focus/node and collocate are related. The higher the z-score is for an item related to the node word, the greater is its degree of collocability with that word (McEnery and Wilson, 1996).

³ An MI score greater than (or equal to) 3 may indicate a significant collocational link (Church and Hanks, 1990: 24; Hunston, 2002: p. 71).

2.4 Problems and Research Questions

Frequency is commonly assumed to influence the word order in fixed expressions. However, is frequency the main factor to decide which item goes first in a conjoined phrase? In addition, the current topic of reduplication and binomial freezes is often neglected because it presents problems for those theorists. Therefore the research questions of the present study will focus on the following topics:

- Reduplication at the lexical level;
- Frequency explored to see whether it is the major factor to influence word order in freezes:
- The frequency and probability of collocation for reduplicative freezes, calculated in terms of MI (Church and Hanks, 1990; Moon, 1998);
- MI and z-score employed to calculate the probability of collocation and assess collocational significance.

Frequency, *MI* and *z*-*score* are explored to see how they shed new light on word order and collocational analysis in terms of an integrated methodology.

3 METHOD & PROCEDURE

3.1 Data and Analytical Tools

The data for the present study include the author's own reduplication corpus and the BNC. The author's data have been gathered in the field, from research papers, dictionaries, websites, newspapers, advertisements, slogans, etc. since 1986. The ongoing collection (about 1,700 tokens) has undergone two stages for this research. The criteria for the establishment of the reduplication corpus are mainly based on the *pattern* (Wang, 2002a):

- The form of each token should be *reduplicated* in various types (onset, rhyme, etc.): e.g., *first and foremost* (247), *this and that* (202), and *town and gown* (8).
- The pattern may be reduplicative binomial or trinomial expressions:
 - **Full copy**, <u>X₁ {conj., prep} X₁</u>, e.g., so and so, and all in all;
 - **Partial reduplication**, <u>X{conj., prep, art}Y</u>, e.g., wine and dine (10), tit for tat (92), trick or treat (15), and bric-a-brac (52);
 - *Triplet and others*: $X_1 X_2 X_3$, e.g., tic tac toe (3) and Milly Molly Mandy (2).

Only the second pattern, partial reduplication, is explored for this current study. The following are the main foci to be investigated in order to demonstrate how reduplication is used pervasively in day-to-day discourse:

- Corpus: using the BNC and Constructing the author's own corpus;
- Calculating the frequency for each selected token in the BNC;
- New power extracting data to be used in the questionnaire;
- Using Google to surf general websites to download examples of reduplication;
- Instruments: using SPSS and SARA for statistical analysis;
- Using SARA searches for the frequency of each token, MI and z-score.

3.2 Procedure and Method

The initial step is to collect data, analyze it and then create the compiled reduplication corpus (1,700 items). The first 232 items were extracted as <u>wordlist 1</u>, and then 116 items were selected as a questionnaire (wordlist 2). All tokens are chosen according to the following principles:

- •All types are mainly based on the patterns, [X] + and/or + [Y], including MWU, fixed expressions and idioms.
- •All tokens identified in the corpus as reduplications undergo SARA searches for their frequency of occurrence, MI-score and z-score.
- •ANOVA and Sheffé tests were used to investigate which approach underwent significant difference in terms of frequency, MI-score and z-score.

Figure 1 summarizes the basic procedure for the current studies:

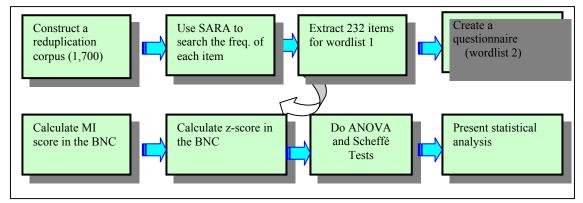


Figure 1 Flowchart for the data-processing

4. RESULTS and SAMPLE ANALYSIS

4.1 Frequency counts and frequency-based method

4.1.1 Procedure 1: a word list with the frequency

Fenk-Oczlon (1989) proposes frequency-based arguments to investigate freezes or fixed expressions. Therefore, the frequency for each token in the BNC was calculated first. The abridged ranking order in Table 2 is based on the results of frequency searches, i.e. the collocational F(x,y). For example, the following F(x) represents 'either' (27152), and F(y), 'or' (367981). The frequency of their collocation, F(x, y) 'either...or...' is 22111. The MI-score of 22.30 and the z-score of .1178 are calculated respectively.

No.groupi $F(x) > F($ $F(x)$ $F(y)$ $F(x, y)$ MIZ scoreTo1High27152 367981 22111 22.30 $.1178$ either X or Y2High+ 193179 197273 1152 30.50 $.4021$ in and out (2000)3High1710 38424 503 30.40 $.6051$ sooner or la4High+ 36609 6086 458 32.70 1.4491 once or twice5High+ 2629 724 276 34.00 3.2693 deaf and du6High+ 120825 604 247 33.90 3.3861 first and for7High+ 5328 12249 156 28.70 $.5973$ upper and lo9High+ 97273 192000 152 24.50 $.1436$ out and abo10High+ 51557 717 134 32.10 2.2891 part and part	•					
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	okens					
$\begin{array}{cccccccccccccccccccccccccccccccccccc$						
$\begin{array}{cccccccccccccccccccccccccccccccccccc$						
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	X in and X					
$\begin{array}{cccccccccccccccccccccccccccccccccccc$						
5 High + 2629 724 276 34.00 3.2693 deaf and du 6 High + 120825 604 247 33.90 3.3861 first and for 7 High + 120825 604 247 33.90 3.3861 first and for 8 High 5328 12249 156 28.70 .5973 upper and lo 9 High + 197273 192000 152 24.50 .1436 out and abor	ater					
6 High + 120825 604 247 33.90 3.3861 first and for 7 High + 120825 604 247 23.00 .0686 this and that 8 High 5328 12249 156 28.70 .5973 upper and lo 9 High + 197273 192000 152 24.50 .1436 out and abord	ce					
7 High 454441 111952 202 23.00 .0686 this and that 8 High 5328 12249 156 28.70 .5973 upper and lo 9 High + 197273 192000 152 24.50 .1436 out and abord	ımb					
1 28 High 5328 12249 156 28.70 .5973 upper and log 9 High + 197273 192000 152 24.50 .1436 out and abord	remost					
8 High 5328 12249 156 28.70 .5973 upper and lo 9 High + 197273 192000 152 24.50 .1436 out and abo	t					
9 High + 197273 192000 152 24.50 .1436 out and abo						
9 High + 197273 192000 152 24.50 .1436 out and abo	ower					
10 High \pm 51557 717 134 32.10 2.2801 part and part	out					
10 Iligii - 51557 717 154 52.10 2.2691 part and par	rcel					
11 High + 8335 4715 112 29.10 .8123 positive or r	negative					
12 High + 24401 9171 83 27.30 .5036 mother and	daughter					
13 High 78 111 69 32.50 4.1748 comings and	d goings					
14 High + 7767 1776 66 29.00 1.0207 plus or minu						
15 High 1187 2048 67 28.80 .9577 fame and fo	ortune					
16 High + 74666 30303 53 24.30 .1978 last but not	least					
17 High 1490 4324 54 27.20 .5917 odds and en	nds					
18 High 3424 9978 48 25.60 .3671 rough and ro	eady					
19 High + 3282 92 49 31.80 3.8643 bits and bob						
20 High + 23864 10059 46 25.50 .358 black and bl						
21 High 5124 19891 44 24.40 .2489 birth and de						
22 High 2663 9361 43 25.40 .3588 mix and ma						
N=116; $n=76$ Correct rate for Fenk-rule $(n/N) = 65.5\%$						

Table 2 Abridged results for frequency-based grouping with MI and Z score

4.1.2 Procedure 2: "High frequency before low frequency"

Fenk-Oczlon's rule (1989; hereby *Fenk-rule*), "high frequency before low frequency," indicates that the frequency of F(x) is higher than that of F(y), i.e., F(x) > F(y). There are only 76 items in the wordlist (total = 116) whose F(x) is bigger than F (y). This means that the correct rate of *Fenk-rule* is 65.5% in our test, not so accurate as 84% correct predictions in Fenk-Oczlon's statement. For example, the frequency of the first part (1710) of 'sooner or later' is lower than that of the second one (38424). It is evident that Fenk rule fails to come up consistently with the correct prediction.

4.1.3 Procedure 3: Frequency-based grouping

Using the ANOVA test, there is no significant difference for frequency-based groups and Z-score groups. Only MI between groups shows a significant difference (*p<.01; see Table 3). Based on the ranking of collocational frequency of occurrences, all the items were divided into three groups, higher (1), middle (2) and lower (3) groups.

ΑΝΟΥΑ								
		Sum of Squares	df	Mean Square	F	Sig.		
FXY.0	Between Groups	22662708	2	11331354	2.069	.133		
	Within Groups	4.60E+08	84	5475709.9				
	Total	4.83E+08	86					
MI.0	Between Groups	600.342	2	300.171	16.437	.000		
	Within Groups	1533.961	84	18.261				
	Total	2134.303	86					
ZSCORE.0	Between Groups	7251390.2	2	3625695.1	2.585	.081		
	Within Groups	1.18E+08	84	1402822.6				
	Total	1.25E+08	86					

Table 3 ANOVA for frequency-based grouping

The Scheffé test further indicates that only MI between groups exhibits significant difference (Table 4). It is evident that something else is needed to make up the shortcomings of frequency-based argument.

Multiple Comparisons

Scheffe				i			
			Mean			95% Confiden	ce Interval
			Difference			Lower	Upper
Dependent Variable	(I) GROUP	(J) GROUP	(I-J)	Std. Error	Sig.	Bound	Bound
FXY.0	1.00	2.00	1169.0814	613.382	.169	-359.4983	2697.661
		3.00	1184.0000	705.544	.250	-574.2500	2942.250
	2.00	1.00	-1169.0814	613.382	.169	-2697.6611	359.498
		3.00	14.9186	613.382	1.000	-1513.6611	1543.498
	3.00	1.00	-1184.0000	705.544	.250	-2942.2500	574.250
		2.00	-14.9186	613.382	1.000	-1543.4983	1513.661
MI.0	1.00	2.00	2.8205*	1.120	.047	2.902E-02	5.612
		3.00	7.2955*	1.288	.000	4.0845	10.506
	2.00	1.00	-2.8205*	1.120	.047	-5.6120	-2.902E-0
		3.00	4.4749*	1.120	.001	1.6835	7.266
	3.00	1.00	-7.2955*	1.288	.000	-10.5064	-4.084
		2.00	-4.4749*	1.120	.001	-7.2664	-1.683
ZSCORE.0	1.00	2.00	70.9095	310.465	.974	-702.7845	844.603
		3.00	707.7545	357.112	.147	-182.1876	1597.696
	2.00	1.00	-70.9095	310.465	.974	-844.6035	702.784
		3.00	636.8450	310.465	.128	-136.8490	1410.539
	3.00	1.00	-707.7545	357.112	.147	-1597.6967	182.187
			-636.8450	310.465	.128	-1410.5390	136.849

Table 4 Scheffé test for frequency-based grouping

he mean difference is significant at the .05 level.

4.2.1 Procedure 4: Mutual Information (MI-based grouping)

The MI-based grouping is based on the results of calculating the association of mutual information between two items (e.g., *binomials*) in terms of using the Sara tool. It can be divided into three new groups, i.e., higher group (the first 23%), lower group (the last 23%), and middle group (the rest). Again, the ANOVA test shows there is no significant difference for frequency-based ranking between groups. On the other hand, both MI and Z- score groups indicate significant differences between groups.

Table 5 ANOVA for MI-based grouping

ANOVA

		Sum of Squares	df	Mean Square	F	Sig
N 41			-		-	Sig.
MI	Between Groups	1585.789	2	792.895	121.425	.000
	Within Groups	548.514	84	6.530		
	Total	2134.303	86			
ZSCOREMI	Between Groups	61326443	2	30663221	40.396	.000
	Within Groups	63762044	84	759071.958		
	Total	1.25E+08	86			
FXY.MI	Between Groups	4805388.3	2	2402694.2	.422	.657
	Within Groups	4.78E+08	84	5688297.0		
	Total	4.83E+08	86			

The Scheffé test pinpoints that MI-based grouping illustrates the significant differences only for the relations between MI and Z-score groups except for one comparison.

Table 6 Scheffé test for MI-based grouping

Scheffe							
			Mean			95% Confide	nce Interval
			Difference			Lower	Upper
Dependent Variable	(I) GROUP	(J) GROUP	(I-J)	Std. Error	Sig.	Bound	Bound
MI	1.00	2.00	6.5869*	.670	.000	4.9176	8.2561
		3.00	11.9773*	.770	.000	10.0572	13.8973
	2.00	1.00	-6.5869*	.670	.000	-8.2561	-4.9176
		3.00	5.3904*	.670	.000	3.7211	7.0596
	3.00	1.00	-11.9773*	.770	.000	-13.8973	-10.0572
		2.00	-5.3904*	.670	.000	-7.0596	-3.7211
ZSCOREMI	1.00	2.00	1716.5628*	228.377	.000	1147.4356	2285.6900
		3.00	2194.8227*	262.691	.000	1540.1837	2849.4617
	2.00	1.00	-1716.5628*	228.377	.000	-2285.6900	-1147.4356
		3.00	478.2599	228.377	.118	-90.8673	1047.3871
	3.00	1.00	-2194.8227*	262.691	.000	-2849.4617	-1540.1837
		2.00	-478.2599	228.377	.118	-1047.3871	90.8673
FXY.MI	1.00	2.00	-379.7368	625.176	.832	-1937.7064	1178.2329
		3.00	154.5909	719.109	.977	-1637.4650	1946.6468
	2.00	1.00	379.7368	625.176	.832	-1178.2329	1937.7064
		3.00	534.3277	625.176	.695	-1023.6420	2092.2974
	3.00	1.00	-154.5909	719.109	.977	-1946.6468	1637.4650
		2.00	-534.3277	625.176	.695	-2092.2974	1023.6420

Multiple Comparisons

* The mean difference is significant at the .05 level.

4.2.2 Procedure 5: Z Score-based grouping

Likewise, given the z-score-based grouping, the frequency-based ranking between groups displays insignificant difference. The Scheffé test verifies the observations that various groups based on MI grouping exhibit significant differences.

4.3 Summary

Given the ANOVA tests and multiple comparisons, the key grouping methods can be summarized in terms of Table 7:

	Ranking	Frequency	MI	Z Score
Grouping		F(x, y)		
Frequency-ba	sed			
MI-based				
Z Score-based	ł			
**p<.01				

Table 7 Nine Different grouping types and the results of ANOVA tests

It is apparent that no matter what kind of grouping method is used, there is no significant difference for frequency ranking groups. However, MI ranking always shows significant difference between groups. If frequency-based grouping is adopted, there is no significant difference for z-score ranking. Multiple comparisons such as Scheffé confirm the above summary. Therefore frequency is not the main factor to decide the word order and the collocations in frozen expressions. It is evident that MI is more important to assess the probabilistic collocation between two adjacent words.

5. CONCLUDING REMARKS

The aim of this research, using corpus-based approaches, has been to explore fixed expressions in terms of

frozen types, frequency, correct rate, collocation, and MI score. A corpus with a ranked wordlist was constructed. The frequency and probability of collocation for reduplication were calculated in terms of MI and z-score. Frequency was explored, but it is not the major factor to influence the word order in freezes. The MI-based method is instead proposed to confirm observations of word order and to revise Fenk-Oczlon's arguments (1989). In addition, ANOVA and post hoc tests, Scheffé, were employed to evaluate which grouping method is appropriate to come up with robust statistics for collocational significance. The key points for this research are summarized along the following lines:

- Given corpus-based approaches, Fenk's rule, high frequency before low frequency, is not always reliable. Among the 116 items in the present corpus, only 76 items meet this rule. The correct rate is 65.5%.
- MI-based approaches play a useful role in aligning, reinforcing or making up the shortcomings of frequency-based arguments.
- MI provides a quick guide to decide what to look for in the collocation pairs. It is used to calculate the probabilistic collocation of two observed items, f(x) and f(y). If the first item, f(x), is larger than f(y), the MI score is higher.

Statistics such as percentage coverage, and frequency of occurrences in a corpus are required to reinforce and constitute relevant arguments and research approaches. MI and z-scores are useful reference points while choosing fixed expressions to discuss. Further studies integrating probabilistic methods are definitely needed.

ACKNOWLEDGEMENT

The author is grateful to Michael McCarthy, John Ohala, Stuart Davis, Ming-wen Wu for their very useful comments. The research is partially supported by a grant, NSC 91-2411-H-130-008, from the National Science Council.

REFERENCES

Aston, G. & Burnard, L. 1998 The BNC handbook. Edinburgh: Edinburgh University Press.

- Biber, D., Conrad, S.&, Reppen, R. 1998 *Corpus linguistics: investigating structure and use.* Cambridge: Cambridge University Press.
- Birdsong, D. 1995 Iconicity, markedness, and processing constraints in frozen locutions. In M. Landsberg (Ed). Syntactic iconicity and linguistic freezes, pp. 31-45.
- Carter, R. 1998 Vocabulary: applied linguistic perspectives. London: Routledge.
- Church, K., &, Hanks, P. 1990 Word association norms, mutual information, and lexicography. Computational Linguistics, 16(1), 22-29.
- Cook, G. 2000 Language play, language learning. Oxford: Oxford University Press.

Cooper, W. & J. Ross 1975 World order. Chicago Linguistic Society, 11, 63-111.

Fenk-Oczlon, G. 1989 Word frequency and word order in freezes. Linguistics, 27, 517-556.

Fenk-Oczlon, G. 2001 Familiarity, information flow, and linguistic form. In J. Bybee & P. Hopper (Eds)

Frequency and the Emergence of Linguistic Structure, pp. 431-448. Amsterdam: John Benjamins. Hoey, M. 1991 *Patterns of lexis in text*. Oxford: Oxford University Press.

Hunston, S. 2002 Corpora in applied linguistics. Cambridge: Cambridge University Press.

Landsberg (Ed). 1995 Syntactic iconicity and linguistic freezes: the human dimension. New York: Mouton de Gruyter.

Landsberg, M. 1995 Semantic constraints on phonologically independent freezes. In Landsberg (Ed). Syntactic iconicity and linguistic freezes: the human dimension, pp. 65-78.

McCarthy, M. 1990 Vocabulary. Oxford: Oxford University Press.

McCarthy, M. 2001 Good listenership made plain: British and American non-minimal response tokens in everyday conversation. MS, University of Nottingham.

McEnery, T. & Wilson, A. 1996 Corpus linguistics. Edinburgh: Edinburgh University Press.

Moon, R. 1998 Fixed Expressions and Idioms in English. Oxford: Oxford University Press.

Nation, P. 1982 Beginning to learn foreign vocabulary: a review of the research. RELC Journal, 13 (1), 14-36.

- Oakes, M. 1998 Statistics for corpus linguistics. Edinburgh: Edinburgh University Press.
- Pinker S. & Birdsong, D. 1979 Speaker's sensitivity to rules of frozen word order. Journal of Verbal Learning and Verbal Behavior, 18, 497-508.
- Schmitt, N. 2000 Vocabulary in language teaching. Cambridge: Cambridge University Press.
- Schmitt, N. & McCarthy, M. 1997 Vocabulary. Cambridge: Cambridge University Press.
- Sinclair, J. 1991 Corpus, concordance, collocation. Oxford: Oxford University Press.
- Stubb, M. 1995 Collocations and semantic profiles: on the cause of the trouble with quantitative studies. Functions of Language, 2 (1), 23-55.
- Tao, H. & McCarthy, M. 2001 Understanding non-restrictive which-clauses in spoken English, which is not an easy thing. Language Sciences, 23, 651-677.
- Wang, S.P. 2001a Integrating corpus-based and vocabulary learning approaches into a linguistics project. Paper presented in 46th International Linguistics Association, New York: New York University.
- Wang, S.P. 2001b Reduplication and repetition in applied linguistics. The Proceedings of 2001 International Conference on the Application of English Teaching, 200~222. Taipei: Crane Publishing Co.
- Wang, S.P. 2002a Corpus-based approaches and discourse analysis to reduplication and repetition, paper submitted to journal.