

Quasi-Equifrequent Group Generation and Evaluation

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The frequency of occurrence and other statistical results derived thereupon from unique items in collections such as letters, words and records has recently formed the basis for the design of optimal information structures. A fundamental theorem of information science states that the information representing capability of a set of symbols is maximized when the probability of occurrence of any symbol in the available set becomes the same. Equifrequency however is very rarely encountered in real applications and it is in many cases desirable to have sets of items or symbols which are equifrequent within a certain deviation i.e. quasi-equifrequent. This paper presents an algorithm for generating equifrequent sets and evaluates and compares the efficiency and accuracy of (a) the entropy and (b) the variance concepts for measuring the degree of quasi-equifrequency in a set. Tests are carried out on the occurrence of the letters A–Z (out of a total of 7,908,100 letters) and on 244 unique subfields (out of a total of 1,113,447 bibliographic record subfields) and an absolutely equifrequent set of subfields is presented.

INTRODUCTION

Basic works on communication and information theory provide simple generalizations regarding efficiency in transmission and storage of information. Although the mathematical theory of communication appeared nearly thirty years ago,¹ it is only recently that an attempt was made to reinterpret the theory and to investigate its implications for information science.² Lynch² reasserts that Shannon's¹ first statement about the equifrequency of symbols and therefore about rectangular frequency distributions, stands as the ideal.

The aim of this paper is to investigate methods of equifrequent set generation and in particular to compare the efficiency and accuracy of the use of (a) the entropy and (b) the variance concepts for measuring the degree of quasi-equifrequency among a set of groups of symbols. Since absolutely equifrequent groups are rarely encountered in real applications, the term 'quasi-equifrequent' is used here to describe all intermediary arrangements prior to the one that can be characterized as optimal.

For testing purposes two different sets of 'symbols' are used: (1) The 26 letters of the English alphabet and their frequencies as calculated by Yannakoudakis out of a total of 7,908,100 letters.³ (2) The frequency of occurrence of 244 different MARC (Machine Readable Catalogue) record subfields. A BNB (British National Bibliography)⁸ file of 31,369 records was used and the frequency of occurrence of all unique fields and subfields was calculated out of a total of 1,113,447 subfields on the lines described by Ayres and Yannakoudakis.⁴

It is by no means unrealistic to consider a MARC subfield as a symbol since it is the basic element from which records are built, in more or less the same manner as words are made from letters of an alphabet. It is believed that this assumption will lead to the design of optimal record structures and hence efficient file structures.

PROBLEM FORMULATION

'Grouping' is defined here as the mapping of the alphabet

$$A = \{\alpha_1, \alpha_2, \dots, \alpha_M\}$$

(where $M = 26$ for the letters and $M = 244$ for the MARC subfields) onto another alphabet

$$G = \{g_1, g_2, \dots, g_N\}$$

such that the groups

$$g_1 = \{\alpha_i, \dots, \alpha_j\}$$

$$g_2 = \{\alpha_k, \dots, \alpha_l\}$$

$$g_N = \{\alpha_m, \dots, \alpha_M\}$$

are equiprobable within an acceptable deviation such that

$$c(g_i) = c(g_j) + \delta_{ij} \quad (1)$$

where $c(g_i)$, $c(g_j)$ represent the cumulative frequencies of groups g_i and g_j respectively, and ideally, δ_{ij} is at a minimum. The problem then is how to calculate the degree of quasi-equifrequency among the members of G so that comparisons between alternative arrangements of all $\alpha_i \in G$ can be made in order to choose the optimal. One criterion would be to minimize the variance of all $g_i \in G$, another to maximize the entropy of the distribution. Nugent and Vegh formulate the problem similarly but do not consider the use of entropy in their experiments.⁵

The variance method utilizes the basic distributional properties of the data set. When the items are arranged in groups, the variance of the distribution is at minimum if the groups are so arranged that their total frequencies are most evenly distributed. If we denote the mean frequency of the group set by \bar{f} then the variance of the distribution is

$$\sigma^2 = 1/N \sum (c(g_i) - \bar{f})^2 \quad (2)$$

Thus the variance method aims to minimize the function

$$\sum (c(\mathbf{g}_i) - \bar{f})^2 \quad (3)$$

The entropy method utilizes Shannon's expression

$$-H = \sum_{i=1}^N P(\mathbf{g}_i) \log_2 P(\mathbf{g}_i) \quad (4)$$

where $P(\mathbf{g}_i)$ is the probability of occurrence of group \mathbf{g}_i . If the groups are absolutely equifrequent then we have a maximum entropy

$$-H_{\max} = N[1/N \log_2 (1/N)] = \log_2 1/N \quad (5)$$

Therefore the relative entropy can be obtained as the fraction $-H/(-H_{\max})$ or relative entropy

$$r = \frac{1}{\log_2(1/N)} \sum_{i=1}^N P(\mathbf{g}_i) \log_2 P(\mathbf{g}_i) \quad (6)$$

Thus the entropy method aims to maximize r ($0 \leq r \leq 1$).

Brack *et al.* used the relative entropy to measure the quasi-equifrequency of character strings (digrams, trigrams, tetragrams etc.) obtained from a number of bibliographic record files.⁶ Although each measure has in the past been used in one application or another, a direct comparison of the efficiency of the two has not been carried out, and apart from Nugent and Vegh,⁵ no detailed description of an algorithm to generate alternative quasi-equifrequent groups is available.

EXPERIMENTAL RESULTS

Given a set of M items in a collection, the algorithm to generate a number of quasi-equifrequent groups will require the following input: (a) identification of each item; (b) frequency of each item; (c) starting number of groups and (d) finishing number of groups. Regardless of the measure used the algorithm will terminate, optionally, upon the fulfilment of one of the following conditions, whichever appears first: (1) finishing number of groups is reached or (2) an absolutely equifrequent group set is generated.

Following a number of considerations and empirical investigations the algorithm was designed and implemented as described below. Although the method cannot guarantee an optimum solution, it will always converge to a near optimal solution. Total enumeration of all possible arrangements in order to choose the optimum would in any case be impractical due to the time constraint involved.

The algorithm

- (1) Sort items by frequency of occurrence in descending order.
- (2) Allocate appropriate storage areas/slots for cumulative frequencies and initialize to zero. (A slot thus becomes synonymous to a group).
- (3) Perform the following steps until all frequencies in the sorted list have been exhausted: (i) Go through all storage slots and identify the slot with minimum cumulative frequency; (ii) Add next frequency in sorted list to the slot identified in step (i) above.

- (4) Calculate the variance or relative entropy for the groups formed.
- (5) Tentatively switch the items of each group with all items of every other group and calculate the resultant variance or entropy immediately after each switch. The best improvement, if any, subject to Eqns (3) or (6), from all switches made is then recorded and the actual switch then takes place. If an improvement is made then step (5) is repeated else step (6) is entered. With the aid of an algorithmic language step (5) becomes:

```

for all  $g1 \in G$  do  $\nexists g1, g2$  are subsets within  $G$ 
for all ( $g2 \in G$  and  $g1 \neq g2$ ) do
for all  $a \in g1$  do  $\nexists a, b$  are elements within  $g1, g2$ 
for all  $b \in g2$  do
  begin  $gt1 := g1 - a + b$ ;
     $gt2 := g2 + a - b$ ;
     $Gt := G + gt1 - g1 + gt2 - g2$ ;
     $v := 1 - \text{entropy}(Gt)$ ;  $\nexists$  or  $v := \text{variance}(Gt)$ 
    if appropriate  $\nexists$ 
if  $v < v_{\min}$  then begin  $v_{\min} := v$ ;
      record ( $a, b, g1, g2$ )
    end if
  end od od od od;
   $g1 := g1 - a + b$ ;  $g2 := g2 + a - b$ ;

```

- (6) If variance becomes zero or entropy reaches one or the finishing number of groups is reached, then stop. Else increment the number of groups by one and return to step (2).

(Note. The switching of items is made subject to the following rules which help to improve the efficiency of the program: (a) items with equal frequencies are not switched; (b) items in single item groups are not

Table 1. (MARC fields) An arrangement into six absolutely equifrequent groups (A three digit code identifies the field and a letter the subfield)

690Z	015A	510A	043A	245B	260D	490A	250A	511A	945X
840A	410U	410V	240R	710M	111A	710L	640A	910D	400H
111I	700V	400V	513A	610R	521A	110I	111F	410T	740P
710I	710H	610U	243S	900B	111C	810A	911C	910J	
TOTAL FREQUENCY =				19605C					
690A	001=	100A	245D	245E	500A	651X	690C	110A	021A
018A	690H	410A	245F	700C	900H	080A	410C	250D	400A
400Y	110E	690E	600C	610B	711A	711J	911J	240Q	710K
740Q	911I	610H	710G	410E	411V	810T	611J	611X	710V
TOTAL FREQUENCY =				19605C					
692A	245A	300C	100H	700A	504A	503A	010A	710A	900X
017A	610A	600H	240A	100C	022A	690F	041R	110H	745A
640S	710E	690H	700T	910B	240P	911Z	740S	243A	740R
100D	710U	690K	910B	400C	745V	640Q	410W	243P	410G
TOTAL FREQUENCY =				19605C					
082B	082A	008=	690D	700H	650X	650Z	245G	945A	945Z
790A	690W	245H	710C	651Y	600X	710B	111J	110L	100E
410W	002=	900C	600E	240S	910F	110K	900F	240D	610Z
690D	110G	400U	110D	740V	700D	645X	610V	710T	490W
610I	411J								
TOTAL FREQUENCY =				19605C					
260A	083A	350A	083B	300B	300F	910Z	440A	900Z	910C
021B	600A	041A	110C	610C	650Y	110B	111K	651Z	600F
900F	440I	640X	690G	910W	700F	110J	740A	911K	740D
710D	110H	645A	411A	610G	410D	245C	640Z	411U	910K
910I									
TOTAL FREQUENCY =				19605C					
650A	260B	260C	300A	050A	910X	900A	690I	651A	900H
690V	690V	440V	610X	690X	910H	250C	518A	504A	910E
100F	600T	911A	911X	610F	610T	700F	711V	840W	700H
711I	710J	600Z	640R	711F	243R	611A	911F	611K	610J
640P									
TOTAL FREQUENCY =				19605C					

Table 2. (MARC fields) An arrangement into nine absolutely equifrequent groups

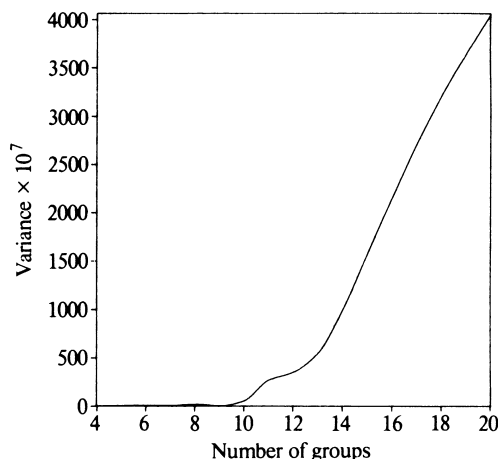
690Z	043A	651X	710A	790A	690U	110C	690X	900U	505A
250P	440U	700T	690G	911Z	610B	111E	740P	740P	710H
610U	900D	111C	810A						
TOTAL FREQUENCY =				130706					
690A	910A	910X	440A	900X	021A	018A	100C	700C	250C
110H	651Z	410W	110E	600C	610D	740S	911J	100D	710J
690K	243S	411V	810T	911C	010J				
TOTAL FREQUENCY =				130706					
692A	015A	050A	245E	690I	900Z	021B	610A	410U	910U
710B	410C	600F	900F	600T	400V	513A	711J	840W	410T
110G	610H	710G	640Q	611J	611X	710V			
TOTAL FREQUENCY =				130706					
082B	300C	300B	504A	500A	651A	900H	945X	840A	245F
022A	111K	110L	745A	002=	640X	700V	110K	110J	243A
740U	645A	110D	410E	700B	410H	610K	490W		
TOTAL FREQUENCY =				130706					
260A	001=	100H	700A	245B	010A	490A	690V	041A	410A
610C	110B	111J	640S	710F	600E	911X	240P	900E	521A
711I	110H	710I	740V	645X	243P	610I	411J		
TOTAL FREQUENCY =				130706					
650A	008=	100A	245D	650Z	245G	910C	511A	600H	240A
410V	111A	518A	640A	100F	690I	910H	910G	700F	610Z
240Q	600Z	010B	400C	410D	410G	610Y	411U		
TOTAL FREQUENCY =				130706					
260B	350A	690D	700H	503A	260D	110A	017A	245H	710C
600X	710I	080A	910E	400T	900C	240S	700F	711K	240U
700U	710U	411A	640R	245C	710T	910K	910I		
TOTAL FREQUENCY =				130706					
083A	260C	300A	650X	900A	945A	945Z	600A	440V	651Y
650Y	710L	100E	400A	111I	610F	610T	740A	110I	600D
710K	911I	745V	610G	640Z	911E	610J			
TOTAL FREQUENCY =				130706					
082A	245A	083B	300E	910Z	690C	250A	690V	690H	610X
240R	690F	041B	910D	400M	911A	690E	910F	711A	911K
710D	740Q	400U	711E	243R	611A	611K	640P		
TOTAL FREQUENCY =				130706					

switched because this can only decrease the degree of quasi-equifrequency within the group. This decrease will be due to the fact that all single item groups will involve items of higher relative frequency than any other item of a multi-item group.)

Test results have proved that in approximately 90% of the cases the terminating condition embodied in step (5) is fulfilled in one pass. The other 10% of the cases involve less than 9 loops in step (5).

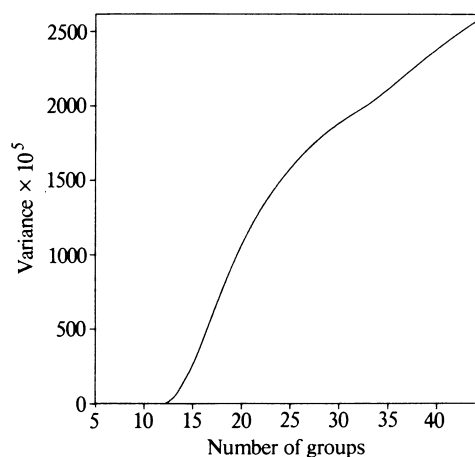
A number of programs were written to implement both methods and record statistical and other information which enabled comparative evaluation under the following main factors: (i) Accuracy of results, (ii) Time, (iii) Sensitivity. For our purpose it was considered appropriate to generate between 5 and 45 groups for all MARC subfields and between 4 and 20 groups for the letters A-Z. Experiments carried out proved that both methods

give similar results in terms of the actual measure used in each case. This is particularly obvious between 4 and 9 groups as shown in Figs 1 and 3 and between 5 and 12 groups as shown in Figs 2 and 4. In actual fact absolutely equifrequent groupings were obtained in 6 and 9 groups for the MARC subfields with both methods and the distributions are presented in Tables 1 and 2.

**Figure 1.** Variance vs number of groups (Letters A-Z).

It is interesting to note that with the frequencies of the letters A-Z no absolutely equifrequent groups could be achieved. An example of this is presented on Table 3 which contains the results for 9 groups.

Some interesting results were obtained when the time involved in each method was considered in our comparisons. Table 4 contains the results for the letters A-Z where it can be seen that as the number of groups

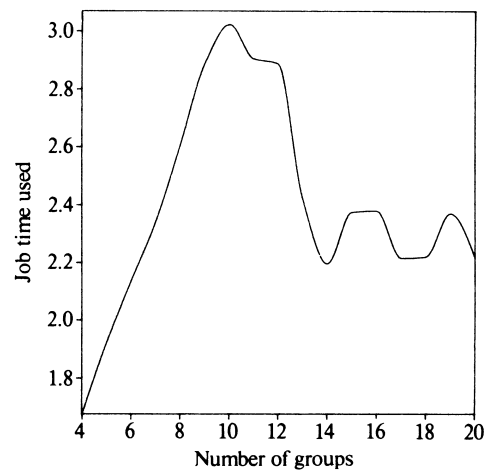
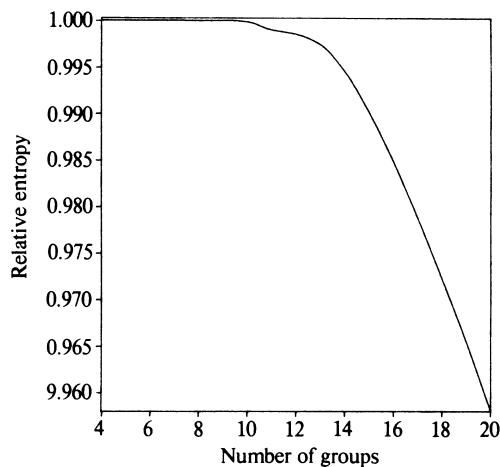
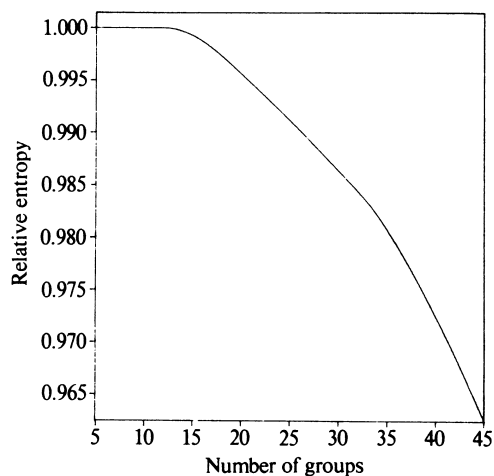
**Figure 2.** Variance vs number of groups (MARC fields).**Table 3. (Letters A-Z) An arrangement into nine quasi-equifrequent groups**

LETTERS	TOTAL FREQUENCY
EZX	874703
AM	874268
NUV	883046
TPWQ	874703
IBK	877821
OH	887511
RC	885749
SYG	871610
LDFJ	879345

increase the run-time of the variance method in comparison with the entropy method decreases from 30.17% in 4 groups to 21.01% in 20 groups. Table 5 contains the results from 5 up to 45 groups for the MARC subfields. Here the variance proves to be, on average, 33% more efficient (in terms of run-time) than the entropy. This can be explained by the fact that the time of fixed overheads (e.g. switching of items) becomes significant in the

Table 4. Time used by relative entropy and variance methods (letters A-Z)

No. of groups	Entropy (s.)	Variance (s.)	Diff. (s.)
4	2.183	1.677	0.506
5	2.500	1.923	0.577
6	2.765	2.135	0.630
7	3.015	2.340	0.675
8	3.327	2.605	0.722
9	3.652	2.887	0.765
10	3.777	3.023	0.754
11	3.700	2.905	0.795
12	3.675	2.888	0.787
13	2.992	2.423	0.569
14	2.745	2.195	0.550
15	2.960	2.375	0.585
16	2.950	2.380	0.570
17	2.715	2.215	0.500
18	2.700	2.220	0.480
19	2.870	2.370	0.500
20	2.684	2.218	0.466

**Figure 5. Time vs number of groups (Letters A-Z).****Figure 3. Entropy vs number of groups (Letters A-Z).****Figure 4. Entropy vs number of groups (MARC fields).****Table 5. Time used by relative entropy and variance methods (MARC fields)**

No. of groups	Entropy (s.)	Variance (s.)	Diff. (s.)
5	148.53	111.95	36.58
6	155.64	116.96	38.68
7	161.59	121.02	40.57
8	165.25	124.21	41.04
9	169.49	127.02	42.47
10	172.30	129.62	42.68
11	176.56	131.79	44.77
12	179.26	133.80	45.46
13	179.12	134.07	45.05
14	181.35	136.00	45.35
15	181.19	136.10	45.09
16	184.23	138.14	46.09
17	186.87	139.95	46.92
18	189.33	141.88	47.45
19	191.16	143.67	47.49
20	193.92	145.48	48.44
21	196.35	146.60	49.75
22	198.51	148.20	50.31
23	198.43	148.42	50.01
24	199.00	150.02	48.98
25	199.90	150.03	49.87
26	202.36	151.52	50.84
27	203.37	152.87	50.50
28	206.04	154.57	51.47
29	206.39	154.41	51.98
30	208.29	155.68	52.61
31	207.15	155.30	51.85
32	203.57	152.45	51.12
33	188.68	142.35	46.33
34	191.35	143.76	47.59
35	193.21	145.71	47.50
36	195.52	146.73	48.79
37	193.30	145.36	47.94
38	195.29	146.93	48.36
39	197.24	148.24	49.00
40	195.34	146.99	48.35
41	190.78	143.88	46.90
42	190.46	143.92	46.54
43	192.75	145.67	47.08
44	195.46	146.79	48.67
45	196.89	148.24	48.65

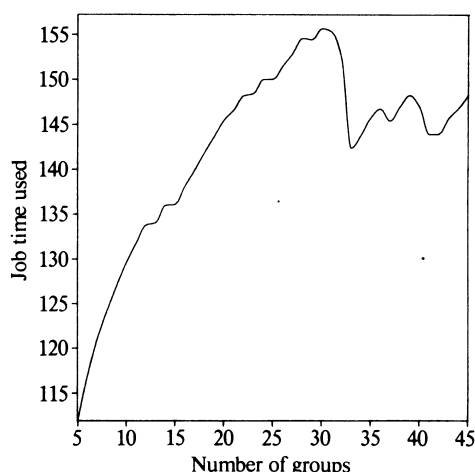


Figure 6. Time vs number of groups (MARC fields).

calculation of the overall time when the size of the collection of items is small. However, when the collection is large, the time of fixed overheads becomes negligible compared with the time taken for the other functions performed.

Figures 5 and 6 show a graphical representation of the time indicated on Tables 4 and 5, respectively, for the variance method, the pattern of which was found to be very similar to the entropy method. In both cases the time increases rapidly then decreases rapidly and finally levels off in a fluctuating pattern. We can explain this as follows: as the number of groups increases, the number of tentative switches among groups increases accordingly. However, this process reaches a turning point (see between 8 to 10 groups in Fig. 5 and between 30 to 32 groups in Fig. 6) where, as the number of groups continues to increase, the number of single item groups increases and this involves less tentative switches between individual groups (i.e. single item groups), the latter being a rule of the algorithm. Therefore the time taken for an arrangement decreases accordingly.

The sensitivity of each method was then studied in terms of the variation of each measure from one tentative switch to the next and from one arrangement to the other. To clarify the concept 'sensitivity', in its present context, let x_n be the measure used (either variance or entropy) for arrangement n and x_{n+1} be the measure of the following arrangement. Then the difference becomes much smaller in the case of the entropy as its value approaches 1 than in the case of the variance. The entropy is thus

characterized as being more insensitive since it successively becomes more and more difficult to choose the best among a number of arrangements produced. The results, therefore, in view of the fact that the final groupings produced by both methods are similar, clearly indicate the superiority of the variance method in terms of speed, flexibility and reliability.

AREAS OF APPLICATION

It is hoped that the results presented herewith will be of value to communication engineers and information scientists working towards efficient transmission and communication. The variety generator seeks to reflect the microstructure of data elements in their description for storage and search, and takes advantage of the consistency of statistical characteristics of data elements in homogeneous data bases.² It is believed that the quasi-equiprecurrent algorithm can serve as a useful tool for analysing these data elements.

Research into coding for optimal record control as presented by Yannakoudakis *et al.* will be able to utilize the present results in order to generate codes for record identification.⁷ This could be achieved by assigning a unique symbol to each of the letter sets of Table 3 which will then be used in the code upon the occurrence of any of the letters in a specific set. For example, given the following assignments:

EZX	0
NUV	2
TPWQ	3
IBK	4
LDFJ	8

The record title EQUIFREQUENT CODING will produce a five digit code 03248.

A fairly recent approach to the optimal file design has been to consider the statistical information of the items concerned and this has in all cases been their frequency of occurrence. If, however, this is supplemented by the frequency of access and particularly co-access it is believed that the use of the quasi-equiprecurrent generation algorithm will partition the items in an optimal arrangement and hence enable optimal placement on storage devices such as magnetic discs and other mass storage devices. Further research on this methodology is at present being carried out at the Computer Centre of this University.

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