COMPUTER MONTE CARLO MODEL OF THE MONOMER REACTIVITY IN BINARY IRREVERSIBLE COPOLYMERIZATION

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Abstract. The paper describes a method based upon the Monte Carlo tehnique for computing the reactivity ratios in binary irreversible copolymerization with ultimate effect. The method is applied with very good results to the methyl methacry-late / chloroprene copolymerization: $\mathbf{r}_1 = 0.08$, $\mathbf{r}_2 = 5.1$ (experimentally determined values), and $\mathbf{r}_1 = 0.073$, $\mathbf{r}_2 = 5.852$ (here computed values). A brief discussion on the stability of some Monte Carlo algorithms used in Polymer Chemistry is also included.

1. Introduction

The Monte Carlo computation of the monomer reactivity ratios has two important advantages: i) the easy usage of the sequence distribution data, and ii) the method does not change the error structure observed in the original data (this problem has been discussed by Behnken¹, Tidwell and Mortimer², McFarlane, Reilly and O'Driscoll³; for an up-to-date review of Monte Carlo applications in Polymer Science one may consult ref.

Our Monte Carlo method⁵⁻⁷ was developed to compute the reactivity ratios in binary irreversible copolymerization with ultimate effect.

where $\sim \sim M_{\rm I}^{\star}$ denotes the macroradical or macroion, $M_{\rm J}$ stands for the monomer, and $r_{\rm I} = k_{\rm II}/k_{\rm IJ}$, I \neq J, defines the reactivity ratios.

This method is presented and applied to the copolymerization of methyl methacrylate with chloroprene.

We note that it is not difficult to extend the method to copolymerizations which follow other reaction mechanisms, i.e., penultimate effect, reversible copolymerization, etc.

2. Monte Carlo Computation of the Reactivity Ratios

Briefly, our Monte Carlo method for computing the reactivity ratios consists in the following: one generates, using the Monte Carlo model^8 of binary irreversible copolymerization with ultimate effect, all the macromolecules possessing the same specified composition, F_1 , for a given feed composition, f_1 . The resulted r_1 , r_2 values are recorded and used within the strategy described below.

For a given (f_1, F_1) pair one obtains n_1 doublets (r_1, r_2) . Let us denote the set S_1 by :

$$S_{i} = \{ (r_{1}, r_{2})_{ij} \}$$
 , $i = 1, 2, ..., n$ (1) $j = 1, 2, ..., n_{i}$

Because the monomer reactivity depends on its chemical identity, the desired r_1 , r_2 values must exist within every set S_1 , $i=1,2,\ldots,n$, and one obtains them performing the intersections:

$$\bigcap_{i=1}^{n} S_{i} = S \tag{2}$$

Because both the experimental data, i.e., (f,F), and the computing method, i.e., the Monte Carlo model, are mortaged by errors and simplifications, respectively, the intersections (2) are performed according to the rule that two pairs of reactivity ratios are equal if they coincide within chosen tolerances as:

where a and b are the chosen tolerances, and $(r_1, r_2)_{1p} \in S_p$, $(r_1, r_2)_{1q} \in S_q$.

The intersections (2) result in the set S:

$$S = \{ (r_1, r_2)_k \}$$
 , $k = 1, 2, ..., m$

and the desired values r_1 , r_2 of the reactivity ratios are obtained as arithmetic means :

$$\mathbf{r}_1 = \frac{1}{m} \sum_{k=1}^{m} (\mathbf{r}_1)_k$$
 , $\mathbf{r}_2 = \frac{1}{m} \sum_{k=1}^{m} (\mathbf{r}_2)_k$ (3)

3. Application: Copolymerization of Methyl Methacrylate with chloroprene

The experimental data displayed in Table 1 are taken from ref. 9; M_1 refers to methyl methacrylate and M_2 to chloroprene.

Table 1. The input data (the polymerization degree N is 1000)

No.	f ₂ /f ₁	F ₁ x 10 ³	F ₂ x 10 ³	
1	2/98	860	140	
2	3/97	770	230	
3	4/96	710	290	
4	5/95	640	36 o	
5	10/90	480	52o	

26.03 26 125 2 5 17 0.585 3.1915 1.195 Table 2. Reactivity ratios $(\mathbf{r_l},\ \mathbf{r_2}\ \mathrm{values})$ computed by means 4 12 7, 2000 12 3 77 6001 0001 00114 6001 0001 00114 6 N N 0.000 r, 12.174 12.174 16.246 19.917 25.130 23 0.094 0.100 0.117 0.126 0.138 Ľ

1 of MEMORY

The elements of the sets S_1 , i=1,2,...,5, are collected in Table 2 (copolymer number corresponding to Table 1). The elements of the S set are in the area marked by dashed lines in Table 2. By means of relations (3), one obtains :

$$r_1 = 0.587/8 = 0.073$$
, $r_2 = 46.819/8 = 5.852$ (4)

The values (4) compare very well with those obtained by Braun et al. (i.e., $r_1 = 0.08$, $r_2 = 5.1$) using usual methods.

4. The Precision and the Stability of the Monte Carlo Models.

The Monte Carlo models are relatively widely used in Physics 11, Chemistry 12 and Polymer Science 4 because they allow to construct the state history (in chronological order) of the real system. The value of the Monte Carlo experiments is conditioned by the precision and the stability of the results one obtains.

It is well known¹³ that the precision of the Monte Carlo computations is proportional to $(D/N)^{1/2}$, where D is a constant and N stands for the number of trials. For the purpose of the Polymer Chemistry (i.e., simulation of polymerization reactions) one assures an acceptable precision considering a polymerization degree N \geqslant 500. The following results, obtained using our MEMORY-3 program⁸, support the above statement:

N	10	50	100	200	300	400	500	600	700	800
_	90	60	57	57	56.67	56.00	56.00	56.17	56.17	56.25
	900	1000								
	56.3	3 5	6.31							

Table 1. The stability of the Monte Carlo results $(\texttt{MEMORY} - 5 \texttt{ type computations}^{15})^{\underline{a},\underline{b}}$

No.	%м ₁	No.	%м ₁	No.	%M ₁	No.	%M ₁
1	66.7	26	66.1	51	65.2	76	65.2
2	67.1	27	66.7	52	66.2	77	66.7
3	66.2	28	66.3	53	66.8	7 8	66.6
4	65.9	29	65.2	54	67.0	7 9	66.9
5	67.3	30	67.6	55	67.8	80	66.8
6	67.6	31	65.7	56	67.6	81	66.9
7	67.7	32	66.9	57	66.3	82	66.2
8	65.7	33	66.6	58	67.0	83	67.3
9	67.3	34	66.2	59	67.1	84	67.6
10	65.3	35	67.0	60	67.7	85	66.4
11	66.6	36	65.9	61	64.6	86	66.5
12	66.7	37	67.0	62	67.3	87	66.8
13	66.8	38	65.9	63	67.7	88	67.3
14	67.3	39	65.4	64	67.3	89	65.8
15	65.9	40	65.5	65	66.4	90	66.3
16	66.6	41	66.3	66	65.3	91	66.6
17	66.5	42	66.7	67	66.7	92	65.4
18	66.1	43	66.1	68	66.6	93	66.7
19	67.5	44	67.1	69	66.9	94	66.4
20	66.0	45	66.2	7.0	67.2	95	66.3
21	65.9	46	67.1	71	66.1	96	66.9
22	66.3	4.7	67.2	72	66.1	97	66.1
23	68.4	48	66.1	7.3	66.9	98	66.5
24	67.7	49	66.2	7.4	67.3	99	67.0
25	67.4	50	65.9	75	66.1	100	67.1

The total computing time is 5 min. 1 sec. on the IRIS 50 computer.

 $[\]frac{b}{c}$ No. 1 means that one has used the sequence 6000 - 7000, No. 2 the sequence 7000 - 8000 etc.

 ${\rm MM}_1$ stands for the molar percent of the ${\rm M}_1$ - mer in the macromolecule with polymerization degree N :

Concerning the stability of Monte Carlo results, we note that they are very senzitive to the statistical quality of the random numbers sequence between 6000 and 7000 is the best sequence of 1000 random numbers furnished by ALEAT (i.e., C.I.I. random numbers generator). Working with the best sequence, one may credit the results obtained in one run. In this way, one gets the important advantage to visualize some molecular features of the macromolecule. In order to study if the results depend upon the random numbers sequence, we "synthetized" 100 macromolecules (N=1000) using 100,000 random numbers (i.e., the sequences 6000 - 7000, 7000 - 8000 etc):

$$\sim M_{I} - M_{J}^{*} + M_{K} \xrightarrow{penultimate} \sim M_{J} - M_{k}^{*}; I,J,K=1,2$$

 $(r_1=0.325, r_1=1.330, r_2=r_2=0.000, M_1$ in feed, molar fraction is 0.6). The results displayed in Table 3 clearly show the dependence upon the statistical quality of the random number sequences which we used.

Because the standard deviation, s, of %M₁ values collected in Table 3 compares well with the experimental accuracy, one may conclude that the variability of the Monte Carlo results is not very important:

$$\mathbf{S} = \left[\sum_{i=1}^{n} (\mathbf{x}_{i} - \overline{\mathbf{x}})^{2} / (\mathbf{n} - 1) \right]^{1/2} = \left[\sum_{i=1}^{100} (\%\mathbf{M}_{1,i} - \%\mathbf{M}_{1})^{2} / 99 \right]^{1/2} = 0.693 \text{ where } \%\mathbf{M}_{1} = \sum_{i=1}^{n} \mathbf{x}_{i} / n = \sum_{i=1}^{100} \%\mathbf{M}_{1,i} / \log = 66.574.$$

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