

# Computer Simulation of Polymers



Editor: E.A. Colbourn

POLYMER SCIENCE & TECHNOLOGY SERIES

063  
C738

9464433

POLYMER SCIENCE AND TECHNOLOGY SERIES

---

# COMPUTER SIMULATION OF POLYMERS

---

Editor: E. A. COLBOURN



E9464433

 Longman  
Scientific &  
Technical

# POLYMER SCIENCE AND TECHNOLOGY SERIES

SERIES EDITORS: DR D M BREWIS AND PROFESSOR D BRIGGS

---

## *Published*

- D E PACKHAM (ed.), *Handbook of Adhesion*  
I S MILES AND S ROSTAMI (eds), *Multicomponent Polymer Systems*  
F R JONES (ed.), *Handbook of Polymer-fibre Composites*

## *Forthcoming*

- D M BREWIS AND B C COPE (eds.), *Handbook of Polymer Science*  
H R BRODY (ed.), *Synthetic Fibres*  
R N ROTHON (ed.), *Particulate Filled Polymer Composites*

**Longman Scientific and Technical**  
Longman Group UK Limited  
Longman House, Burnt Mill, Harlow,  
Essex, CM20 2JE, England  
and Associated Companies throughout the world.

Copublished in the United States with  
John Wiley & Sons, Inc., 605 Third Avenue, New York, NY 10158

© Longman Group UK Limited 1994

All rights reserved; no part of this publication may be reproduced, stored in a retrieval system, or transmitted in any form or by any means, electronic, mechanical, photocopying, recording, or otherwise without either the prior written permission of the Publishers, or a licence permitting restricted copying in the United Kingdom issued by the Copyright Licensing Agency Ltd, 90 Tottenham Court Road, London, W1P 9HE

First edition 1994

ISBN 0 582 08374 5

### **British Library Cataloguing in Publication Data**

A catalogue record for this book is available from the British Library

### **Library of Congress Cataloging-in-Publication Data**

Computer simulation of polymers / editor, E. A. Colbourn.

p. cm.—(Polymer science and technology series)

Includes bibliographical references and index.

ISBN 0-470-23343-5

1. Polymers—Computer simulation. I. Colbourn, E. A. (Elizabeth A.)

II. Series.

QD381.9.E4C68 1994

547.7'01'13—dc20 93-36416 CIP

Set by 16JJ in 10/12½pt Times

Printed and bound in Great Britain  
by Bookcraft (Bath) Ltd

## *List of contributors*

---

A J Hopfinger, Department of Chemistry and Department of Medicinal Chemistry and Pharmacognosy, M/C 781, University of Illinois at Chicago, Box 6998, Chicago, IL 60680, USA

M G Koehler, Department of Chemistry and Department of Medicinal Chemistry and Pharmacognosy, Box 6998, Chicago, IL 60680, USA  
(Permanent address: Allied-Signal Inc., Engineering Materials Research Center, 50 East Algonquin Road, Box 5016, Des Plaines, IL 60017-5016, USA)

J H R Clarke, Chemistry Department, UMIST, Manchester M60 1QD, UK

D Brown, Chemistry Department, UMIST, Manchester M60 1QD, UK

K Binder, Materialwissenschaftliches Forschungszentrum (MWFZ) und Institut für Physik, Johannes Gutenberg Universität Mainz, D-55099 Mainz, Staudinger Weg 7, Germany

E A Colbourn, Materials Division, Oxford Molecular, The Magdalen Centre, Oxford Science Park, Sandford-on-Thames, Oxfordshire OX4 4GA

J Kendrick, ICI Wilton Materials Research Centre, P.O. Box 90, Wilton, Middlesbrough, Cleveland TS6 8JE

G Goldbeck-Wood, H H Wills Physics Laboratory, University of Bristol, Tyndall Avenue, Bristol BS8 1TL, UK

Y Termonia, Central Research and Development, Experimental Station, E I du Pont de Nemours, Wilmington, DE 19880-0356, USA

S Kumar, Department of Materials Science and Engineering, Polymer Science Program, The Pennsylvania State University, University Park, PA, USA

**B E Eichinger**, BIOSYM Technologies, Inc., 9685 Scranton Road, San Diego, CA 92121-3752, USA

**O Akgiray**, BIOSYM Technologies, Inc., 9685 Scranton Road, San Diego, CA 92121-3752, USA

**D J Osguthorpe**, Molecular Graphics Unit and School of Chemistry, University of Bath, Claverton Down, Bath BA2 7AY, UK

**P Dauber-Osguthorpe**, Molecular Graphics Unit and School of Chemistry, University of Bath, Claverton Down, Bath BA2 7AY, UK

## Contents

---

### *List of Contributors*

<i>Chapter 1</i>	Molecular modelling of polymers 9. Description and application of torsion angle unit theory to predict polymer properties	A J HOPFINGER AND M G KOEHLER	1
	1.1	Introduction	1
	1.2	Torsion angle unit (TAU) theory	5
	1.3	Estimation of intramolecular TAU physicochemical properties	9
	1.4	Estimation of intermolecular TAU physicochemical properties	17
	1.5	Results of applying TAU theory	23
	1.6	Discussion of TAU theory and results of QSPR studies	41
		References	44
<i>Chapter 2</i>	Molecular dynamics modelling of amorphous polymers	J H R CLARKE AND D BROWN	46
	2.1	Introduction	46
	2.2	Models and methods	49
	2.3	Preparation of amorphous samples	57
	2.4	Mechanical 'experiments' on model polyethylene	63
	2.5	Conclusions	81
		Acknowledgements	82
		Appendix 1 The loose-coupling algorithm	82
		Appendix 2 Method 1 vs Method 2	86

---

Appendix 3	Unbiased Monte Carlo sampling of chain configurations	88
	References	89
<i>Chapter 3</i>	Monte Carlo studies of collective phenomena in dense polymer systems K BINDER	91
	3.1 Introduction: length and time scales	91
	3.2 Some details on 'coarse-graining' models for polymers	99
	3.3 Crossover in semi-dilute polymer solutions and the dynamics of polymer melts	100
	3.4 Unmixing in polymer blends	111
	3.5 Local ordering and chain stretching in block copolymer melts	121
	3.6 Final remarks	125
	References	127
<i>Chapter 4</i>	Molecular modelling of crystalline polymers E A COLBOURN AND J KENDRICK	130
	4.1 Introduction	130
	4.2 Polymer morphology	131
	4.3 Molecular mechanics force fields	132
	4.4 Determination of force field parameters	136
	4.5 Simulations of crystal structure	145
	4.6 Computer simulation of poly(ethene) crystals	152
	4.7 Simulation of condis crystals	157
	4.8 Modelling studies of polyamide systems	157
	4.9 Conclusions	161
	References	162
<i>Chapter 5</i>	Computer simulation of polymer crystallization G GOLDBECK-WOOD	165
	5.1 Introduction	165
	5.2 The microscopic level	166
	5.3 The macroscopic level	193
	References	197

---

<i>Chapter 6</i>	Molecular models for polymer deformation and failure	Y TERMONIA	200
	6.1	Introduction	200
	6.2	Linear polymers	201
	6.3	Crosslinked polymers	211
	6.4	Conclusions	225
		References	226
<i>Chapter 7</i>	Monte Carlo simulations of the free energies and phase diagrams of macromolecular systems	S KUMAR	228
	7.1	Introduction	228
	7.2	Chain chemical potentials from simulation	234
	7.3	Results and discussion	239
	7.4	Phase equilibrium behaviour	254
	7.5	Summary	258
		Acknowledgements	258
		References	259
<i>Chapter 8</i>	Computer simulation of polymer network formation	B E EICHINGER AND O AKGIRAY	263
	8.1	Introduction	263
	8.2	Analytical theories of gelation	266
	8.3	Simulation techniques	282
	8.4	New results	294
	8.5	Conclusions	299
		References	300
<i>Chapter 9</i>	Computer simulations of biopolymers	D J OSGUTHORPE AND P DAUBER-OSGUTHORPE	303
	9.1	Introduction	303
	9.2	Methods and techniques	307
	9.3	Examples of applications to peptides and proteins	317
	9.4	Conclusions and further directions	332
		References	333
	Index		337



*Molecular modelling of polymers 9. Description and application of torsion angle unit theory to predict polymer properties*

---

A J HOPFINGER AND M G KOEHLER

## 1.1 Introduction

### 1.1.1 History and background of computer-assisted molecular design

The application of computer-assisted molecular design (CAMD) approaches to the development of new polymeric materials is being actively explored.<sup>1</sup> The considerable interest in this 'hi-tech' approach to the development of polymer products arises jointly from the fundamental insight that can be realized, and the enormous gain in design efficiency that is possible from its use. The extension of CAMD approaches in materials science is a natural progression of the significant impact CAMD has had in the pharmaceutical industry. Virtually every major pharmaceutical company, worldwide, has a group of scientists who are doing computer-assisted drug design.<sup>2-4</sup> In terms of macromolecules, computer-assisted drug design has been directed to expand its focus and effort on modelling the interactions of ligands (potential drugs) with biological macromolecules—proteins and polynucleic acids. The conformations of the biological macromolecules have been determined mainly from X-ray crystallography, and, to a lesser extent, from NMR methods. X-ray crystallography and NMR spectroscopy will likely merge with current computer-assisted drug design methods over this decade to yield integrated experimental and computational approaches for the efficient and reliable design of new biopharmaceutical agents.

Unfortunately, the application of CAMD in polymer science is not as direct as it is in the pharmaceutical sciences. In the case of designing a drug, there is usually an isomorphic relationship between a change in the structure of a drug-candidate molecule and the corresponding biological response. Thus, one can deal with discrete molecular entities – the drug candidate, and even better, also its macromolecular receptor – in the molecular design process.

In the case of the design of polymeric materials the problem becomes

more diffuse, less defined. What must be evaluated (modelled) is a volume element of the polymeric material that is of sufficient size so as to allow the estimation of key molecular properties that, at the least, correlate to, or are indicative of, the bulk properties of the material. Moreover, the polymer is not a distinct molecular entity unto itself. Issues of molecular weight, chemical defects and structural defects must be considered in even the most simple linear, synthetic polymers. In other words, there is not necessarily an isomorphic relationship between the physicochemical properties of a single polymer molecule and the bulk properties of a material derived from the polymer. Overall, the lack of an isomorphic molecule–property relationship makes polymer design more difficult than drug design.

### *1.1.2 Approaches to computer-assisted polymer design*

The most comprehensive approach to the modelling of structural features of polymeric materials is computer simulation–molecular dynamics and/or Monte Carlo techniques.<sup>5,6</sup> Computer simulation of polymer structure permits estimation of average properties of polymer systems containing multiple polymer chains and other types of molecular entities such as solvent, plasticizer and/or crosslinking agents. At this time in the evolution of CAMD there are two significant drawbacks to the use of computer simulation of polymers in a practical design mode.

1. The time, effort and computer resources to do a polymer simulation are often so large as to negate the impact of the calculation on a design programme. It may simply be easier to make and test a targeted polymer system than to do the corresponding simulation.
2. It is often difficult to abstract the relevant information/molecular properties from a simulation that are indicative of the bulk property of interest. For example, what motions in molecular dynamics simulations correspond to the onset of the glass transition,  $T_g$ ?

Molecular simulations on meaningful models of polymer materials have only been possible for a short time. Hence other approaches to predicting bulk properties, obviously of much less computational intensity, have been considered over the past years. The most straightforward of these approaches are models which are based upon the group additive property (GAP) concept.<sup>7–9</sup> The GAP concept assumes that some intrinsic contribution to any bulk property,  $P_B(i)$ , is associated with the  $i$ th structural group of the polymer. Most of the GAP models deal with homopolymers so that consideration of the structural groups composing the monomer unit is sufficient to define all contributions to the bulk property,  $P_B$ .  $P_B$  is

simply taken to be the sum of the  $P_B(i)$  composing the monomer,

$$P_B = \sum_{i=1}^m P_B(i) \quad [1]$$

In Eqn [1] the index  $m$  corresponds to the number of structural groups in the monomer unit. Clearly, group additivity is often far from a reasonable assumption, and most GAP models include non-additive correction factors,  $f_B(i)$ , so that Eqn [1] becomes

$$P_B = \sum_{i=1}^m (P_B(i) + f_B(i)) \quad [2]$$

The majority of polymer GAP models have been developed to predict one particular bulk property with a strong bias favouring the estimation of  $T_g$ .<sup>7,10-13</sup> However, VanKrevelen has undertaken the monumental task, with considerable success, of devising a family of homologous GAP models to estimate a wide variety of bulk properties of homopolymers for both the solution and solid-states (amorphous, semi-crystalline and crystalline).<sup>7</sup> While the mathematical formats of the VanKrevelen GAP models vary somewhat, they can be characterized by

$$P_B = \sum_{i=1}^m \frac{P_B(i) + f_B(i-1, i, i+1)}{M_w} \quad [3]$$

where  $P_B(i)$  is the weighing factor for the  $i$ th group for bulk property  $P_B$ ,  $f_B(i-1, i, i+1)$  is a non-additive correction factor for the  $i$ th group, and  $M_w$  is the monomer molecular weight. The introduction of  $M_w$  in the VanKrevelen GAP models means that the bulk polymer properties are estimated from a parameter set that scales to monomer molecular weight. The individual  $P_B(i)$  and  $f_B(i-1, i, i+1)$  are determined by fitting procedures using a training set of polymers for which the bulk property of interest has been measured.

The VanKrevelen GAP formalism, overall, has been quite successful. In particular, it has been reliable in predicting relative differences in bulk properties of polymers having structural homology. Moreover, GAP methods are quite easy to employ and interpret. The calculations are simple and fast, and the end-point bulk property well defined.

However, GAP methods have three serious limitations:

1. If the group parameters are not available for the polymer of interest, then a bulk property estimation is not possible. Moreover, there do not appear to be any overall guidelines to estimate group parameters for any of the more popular/general GAP formalisms.
2. The real world is not group-additive, and even the introduction of non-additive correction factors still leaves, ultimately, a chemical topological (graph) model, as opposed to a three-dimensional molecular

model. Consequently, those bulk properties that are highly dependent upon spatial behaviour on the molecular level may not be estimated accurately.

3. GAP models do not provide mechanistic information that can be used to provide an understanding of why a particular polymer exhibits a particular bulk property measure. This information can only come from higher level modelling.

There is a level of polymer modelling intermediate to GAP formalisms on the one hand, and molecular simulation on the other. This level (class) of molecular modelling can be described as three-dimensional static (3DS) modelling and encompasses conformational analyses of polymer chains,<sup>14</sup> polymer crystal packing calculations,<sup>15</sup> and statistical mechanics formalisms of which Flory's Rotational Isomeric State (RIS) theory<sup>16</sup> is most prominent. 3DS modelling studies actually represent the majority of polymer structure calculations over the past thirty years. However, 3DS studies have not often been used to predict bulk polymer properties with the exception of crystal structures. 3DS calculations have been used, in the main, to provide 'pictures' of polymer chain conformations under different conditions, usually ranging from the amorphous to crystalline solid-state, and/or dilute to dense solution states.

### *1.1.3 Overview of torsion angle unit theory*

Given the current situation regarding approaches to the molecular modelling of polymers, with the goal of predicting properties, we asked ourselves if some hybrid technique could be devised which maximizes the advantages of each of the current techniques, but minimizes their respective drawbacks. One additional constraint was considered in formulating our answer – that computational power will increasingly grow so that the applications of large molecular simulations will become increasingly practical.

The remainder of this chapter describes our answer to the question posed above. We have developed a polymer modelling formalism we call torsion angle unit, TAU theory.<sup>17,18</sup> The essential features of TAU theory, which are expanded upon in the balance of this chapter, are:

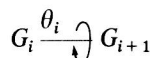
1. The theory, in its current state of development, only treats homopolymers, and uses the structural repeat unit (SRU), which is the monomer in most cases, of a polymer, as the fundamental building block.
2. The SRU of the homopolymer is decomposed into its corresponding set of TAUs, and the molecular physicochemical properties of each TAU are taken from an existing table, or computed directly using

- 3DS modelling and/or molecular simulations. The net physico-chemical property of the polymer is then taken as the sum of TAU contributions, that is, group additivity is assumed.
3. Step 2 is repeated for each member of a set of homopolymers for which the target bulk property ( $T_g$ ,  $T_m$ , modulus, etc.) has been measured.
  4. Step 3 leads to the generation of a structure–property relationship (SPR) table for the ‘training set’ of homopolymers. The data in the SPR table are used to formulate a quantitative structure–property relationship (QSPR). The measured bulk polymer properties are considered as dependent variables to the set of calculated physico-chemical properties of the SPR table which, in turn, are derived from the TAU of the SRU.
  5. The QSPR is generated by performing multi-dimensional linear regression analyses of the molecular physicochemical properties against the bulk polymer property measures. The preferred QSPR is that which maximizes the statistical significance of fit between a specific set of molecular physicochemical properties and the measured bulk polymer property.

## 1.2 Torsion angle unit (TAU) theory

### 1.2.1 The torsion angle unit

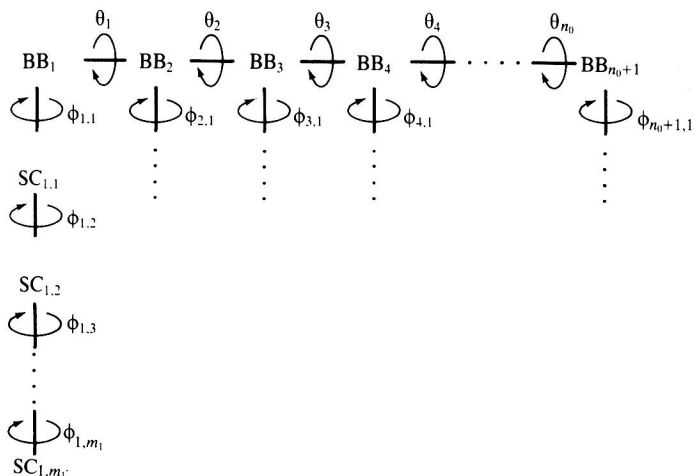
The key to being able to efficiently estimate physicochemical properties of polymers is the representation of the polymer in terms of torsion angle units, TAUs. A TAU is schematically defined as



where  $G_i$  and  $G_{i+1}$  are structural groups connected by a bond about which the torsion angle  $\theta_i$  occurs. The polymer is built up by connecting TAUs together such that the ‘right’ structural group of the  $i$ th torsion angle becomes the ‘left’ structural group of the  $i + 1$  TAU. Figure 1.1 illustrates the structure of a linear polymer in terms of TAUs.

TAUs permit a global molecular property of a polymer system to be computed as a scalar sum of the individual molecular properties of the constituent TAUs. That is, we have formulated the estimation of global molecular properties in terms of a GAP model. The global molecular properties can be correlated to macroscopic properties of the system to hopefully yield a QSPR.

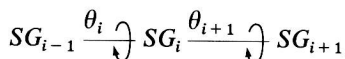
A major limitation of GAP models is that the requisite GAP parameters needed to make an estimation of a bulk property are not always available.



**Figure 1.1** Structure of a linear polymer as formulated to define torsion angle units. BB refers to backbone TAUs and SC to sidechain TAUs.

Thus, while the calculation of the global molecular property is straightforward, it cannot be done because of the missing data. However, the formulation of a GAP model using TAUs permits the application of molecular modelling methods to estimate the GAP parameters associated with any TAU. Thus, one only needs to know, or develop, a scheme to compute the requisite group additive properties using molecular modelling to ensure being able to parameterize any TAU. Moreover, once the GAP properties are computed for a TAU they can be stored in a database and used whenever the TAU turns up in a study. Overall, the marriage of molecular modelling and GAP approaches, and their joint application to estimate TAU molecular properties, provides an open-ended, general approach to estimate bulk polymer properties.

It should be pointed out that the TAU described and used in this chapter is one-dimensional (one degree of conformational freedom) in torsion angle,  $\theta_i$ , when bond lengths and angles are held constant. This representation of the TAU is not a restriction. Higher dimensional TAUs can be defined if needed. For example,



defines a two-dimensional TAU. Obviously, the computational effort needed to compute the molecular properties increases with increasing torsion angle dimensionality. However, the coupling of structural units composing the polymer chain also increases which should enhance the



In the example above, a nine-dimensional problem, in terms of torsion angles, is reduced to nine one-dimensional problems by application of TAU theory. Of course this spatial decoupling process, which is what TAU theory achieves, is paid for at the expense of the accuracy and reliability of predicting the corresponding physicochemical properties of the polymer system. It is reiterated that loss in accuracy and reliability in estimating physicochemical properties can be 'controlled' by selection of TAUs of a higher torsion angle dimension than one.

### *1.2.2 Advantages of TAU theory in predictive polymer modelling*

The one-dimensional TAU units, which will be the only TAUs discussed hereafter, represent relatively simple structures to use in both 3DS modelling and/or molecular simulation studies. In being able to make this statement TAU theory intrinsically overcomes two major drawbacks of GAP models, and one limitation of full-scale molecular simulation, to polymer property predictions. First, physicochemical molecular properties can be determined by prescribed molecular modelling schemes, described below, for any TAU. Thus, there is no parameterization limitation in TAU theory as is the case in GAP models. TAU theory provides an open-ended means to compute requisite physicochemical properties (the parameters).

Since TAU theory involves explicit three-dimensional geometries, that can be as complex as the user selects in terms of numbers of torsion angles, many of the non-additive properties inherent to polymer systems can be taken into account. Thus, TAU theory can overcome some of the shortcomings of GAP formalisms implicit to the additivity assumption.

The overall structural simplicity of TAUs makes the calculation of associated physicochemical properties quite rapid and straightforward. Thus, there are not the computational limitations inherent to molecular simulations of samples of polymer systems.

TAU theory does provide structure–property models that have more insight than GAP formalisms in terms of understanding mechanisms of bulk property behaviour. These TAU structure–property models are generally expressed as QSPRs. The QSPRs are derived from performing multi-dimensional regression analysis on a training set of polymers in which a bulk polymer property of interest is correlated against one, or more, TAU physicochemical properties. This is discussed in the next section. It is not clear at this time in the evaluation of polymer modelling to what extent the mechanistic information of a QSPR has as much insight or meaning compared to that possible from molecular simulations.



### 1.3 Estimation of intramolecular TAU physicochemical properties

#### 1.3.1 Scaling intramolecular TAU calculations

All TAU calculations must be scaled with respect to the structure of the polymer. Only treatment of homopolymers has been developed to date. The SRU of the polymer, which is normally the monomer unit, serves to define the set of TAU units associated with a polymer chain. The TAUs are divided, as noted earlier, into backbone and sidechain TAUs. Backbone torsion angles of the  $i$ th BB-TAU are denoted by  $\theta_i$ , while sidechain torsion angles of the SC-TAU are given as  $\phi_{i,j}$ , where  $j$  stands for the  $j$ th torsion angle of the sidechain attached to the  $i$ th backbone TAU. Using this notation the scaled backbone physicochemical TAU property for the entire SRU,  $F_{\text{BB}}$ , is given by

$$F_{\text{BB}} = \frac{\sum_{i=1}^{N_{\text{BB}}} f_{\text{BB}}(\theta_i)}{N_{\text{BB}}} \quad [4]$$

where  $f_{\text{BB}}(\theta_i)$  is the physicochemical property for the  $i$ th backbone TAU and  $N_{\text{BB}}$  is the number of BB-TAU in the SRU.

In like fashion to Eqn [4] the physicochemical property of the  $i$ th sidechain for the monomer  $F_{\text{SC}}(i)$  is given by

$$F_{\text{SC}}(i) = \frac{\sum_{j=1}^{N_{\text{SC}}(i)} f_{\text{SC}}(\phi_{i,j})}{N_{\text{SC}}(i)} \quad [5]$$

where  $f_{\text{SC}}(\phi_{i,j})$  is analogous to  $f_{\text{BB}}(\theta_i)$  and  $N_{\text{SC}}(i)$  corresponds to  $N_{\text{BB}}$  in Eqn [4]. The total sidechain contribution for the monomer,  $F_{\text{SC}}$ , can be directly computed as

$$F_{\text{SC}} = \frac{\sum_{i=1}^{N_{\text{BB}}} F_{\text{SC}}(i)}{N_{\text{BB}}} \quad [6]$$

Three important components of TAU theory, expressed by Eqns [4]–[6], are (a) the set of TAU is defined by the SRU; (b) each physicochemical property is normalized/scaled against the number of backbone and/or sidechain torsion angles; and (c) physicochemical properties can be partitioned between backbone and sidechain structures as well as between individual sidechains of an SRU. TAU theory does not normalize/scale against SRU molecular weight, as does the VanKrevelen formalism.<sup>7</sup> Molecular weight within TAU theory, is, as discussed below, treated as a physicochemical property.

#### 1.3.2 Conformational entropy, $s_{\text{BB}}(\theta_i)$ and $s_{\text{SC}}(\phi_{ij})$

The principal property used to quantitatively estimate intramolecular flexibility is the conformational entropy associated with TAUs. The