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Molecular dynamics simulation of the viscoelasticity of polymer nanocomposites under oscillatory shear: effect of interfacial chemical coupling

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In this work by adopting coarse-grained molecular dynamics simulation, we focus our attention on investigating the effect of the chemical coupling between polymer and nanoparticles (NPs) on the viscoelastic properties of polymer nanocomposites (PNCs). Firstly we examine the effect of the interfacial chemical coupling on the non-linear behavior, such as the change of the storage moduli, the loss moduli and the loss factor as a function of the strain amplitude. Besides the reinforcing effect contributed by the interfacial chemical interaction, a much smaller loss factor is also observed attributed to less molecular friction and dissipation. Meanwhile, the effects of temperature, frequency, and the interfacial physical interaction between NPs and polymers on the viscoelastic properties are also probed. To uncover the structural and dynamic effect of the interfacial chemical coupling, we calculate the radial distribution function of polymer chains around NPs, the content of the polymer beads in the first layer of the interfacial region under quiescent and dynamic conditions, the incoherent intermediate dynamic structure factor of the polymer beads, which are chemically or physically tethered to the NPs, and all the polymer beads of the system, the quantitative comparison of the mean relaxation time for different interfacial chemical coupling, and the mean-square displacement of the polymer chains. Lastly we analyze the change of the interfacial energy such as the physical and chemical energies during oscillatory shear. Through these analyses, we conclude that with the increase of the interfacial chemical coupling, the change extent of the interfacial physical interaction *versus* the periodic strain decreases, attributed to a much smaller adsorption–desorption reversible process. This can rationalize the much weaker non-linear behavior or the “Payne effect”. Based on these results, we anticipate that a better molecular-level understanding is provided on the effect of the interfacial coupling on the viscoelastic properties of PNCs.

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1. Introduction

Filled elastomers are of great significance and have been extensively used in industrial applications. When subjected to a certain degree of periodic strain, filled elastomers exhibit

typical nonlinear behavior, such as the reduction of the modulus as a function of the dynamic strain amplitude, which is commonly called the “Payne effect”.^{1,2} Payne explained that this effect was attributed to the destruction of the physical interaction between fillers. The fact that the modulus decreases sharply with the increase of the strain is detrimental for filled polymers used under dynamic loading. It is important to understand the nature of the “Payne effect” and attenuate it in some conditions. Since the concept of “Payne effect” is proposed, several models have been presented to illustrate this phenomenon. There is a widely held view that the non-linear strain-dependent behavior of filled rubbers is attributed to the break-down of filler networks. Kraus³ suggested the first phenomenological quantitative model to describe the “Payne effect” relying on the agglomeration–deagglomeration of carbon black agglomerates. Furthermore, Kraus considered that it is the excess forces between particles or between particles

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and polymers that leads to the non-linear decrease behavior. The van de Walle–Tricot–Gerspacher model⁴ introduced a weighting function to establish the relationship between G' and G'' . The NJ-model⁵ tended to examine the energy dissipation process in junction points between carbon black aggregates and provided a filler network interpretation for modulus increase with increasing filler content. The Links–Nodes–Blobs model^{6,7} is introduced by Lin and Lee to describe the strain-dependent dynamic properties of filled rubber network caused by the destroying of chain-like structures and aggregates interconnected by weak bonds. Alternatively, Cassagnau⁸ proposed that the “Payne effect” is associated with the chain disentanglements as well as the filler network breakdown, which are directly related to the silica content and the dynamic strain amplitude. In contrast to the above mechanisms, Maier *et al.*⁹ proposed that the “Payne effect” can be produced by the stress-induced debonding of polymer chains from the filler surface. Moreover, Sternstein and Zhu *et al.*^{10,11} found that the “Payne effect” still occur at low filler concentrations, in which case the filler network is difficult to form, which results from the release of the trapped polymer entanglement structures at polymer–filler interfaces that determine the viscoelastic behavior of the composite, rather than the breakage of the filler agglomeration or network. Alireza¹² declared in his research that the amplitude-dependent dynamic moduli is attributed to disentanglement of bulk polymer chains from those strongly adsorbed to the surface of nanoparticles. This is essentially due to the mismatch between the relaxation times of the free and adsorbed chains and flow-induced disentanglement of adsorbed chains in fast flows. Moreover, Yang *et al.*¹³ consider that the microscopic strain of the matrix is the major reason for reinforcement and onset of the “Payne effect” instead of the variations of filler agglomeration. The aforementioned theories reasonably illustrate the “Payne effect” to a certain extent and can be summarized to the following reasons from two aspects: (i) the particle–particle interaction and (ii) the polymer–particle interaction.

Up to now, several simulation studies have been carried out to investigate the viscoelasticity of polymer nanocomposites,^{14–24} because of its fine analysis and characterization at the molecular level compared to experimental tools. Our group has carried out the relevant research before. To be specific, Chen *et al.*²⁵ provided an insight on the role of polymer–particle interactions in the strain-dependent viscoelasticity, by studying the cross-linked elastomeric systems containing various concentrations of monodispersed NPs with different interfacial interactions *via* coarse-grained molecular dynamics simulation. Chen concluded the two important factors contributing to the “Payne effect” of filled elastomers, such as the breakage of the rigid polymer shell around NP and the rearrangement of polymer-shell-bridged NP network under oscillatory shear. While in the case of low loading the debonding of polymer monomers from the NP surfaces as well comes into play. Wang *et al.*²⁶ found that, with the increase of the formed semi-interpenetrating NP network by molecular simulation, the “Payne effect” decreases at small strain amplitude, while the dynamic storage modulus was strengthened, and the

energy loss was lowered at large strain amplitude. The result is attributed to the break-up of the aggregation of the NPs and an enhanced interfacial contact area between the polymer chains and NPs which leads to the much friction. Raos *et al.*²⁷ used the coarse-grained DPD simulation to study the change of the storage modulus of filled elastomers under the oscillatory shear flow. They observed a strain stiffening of the material (anti-Payne behavior), different from those observed experimentally. And in their next study,²⁸ they claimed that the particle–particle interaction does not directly influence the “Payne effect”. Moreover, a dependence of the small-strain nonlinear behavior of filled elastomer was also seen in FEM simulations by Gusev *et al.*²⁹

Although some simulation work have been carried out to investigate the mechanism of the “Payne effect” and most of them focused on the role of the physical interactions of the polymer–filler and filler–filler. To promote and realize a better dispersion of NPs and corresponding properties in the polymer matrix, the chemical modification technique is widely used, which is favorable to enhance the compatibility of the NPs with the polymer matrix, such as the coupling agent modification³⁰ and the polymer grafting modification.³¹ Furthermore, some studies^{32,33} indicated that the coupling agent modification of fillers could weaken the “Payne effect” due to the reduction of filler–filler interaction or the filler network. Qiao *et al.*³⁴ used epoxy group-functionalized styrene-butadiene rubbers to form a covalent bonding with the hydroxyl groups on the silica surfaces resulting from a ring-opening reaction. The inclusion of just 1.6 wt% of glycidyl methacrylate (GMA) reduces the initial G' by 75% with the “Payne effect” decreasing significantly. The authors explained this in two factors: (1) the improved dispersion of silica which reduced the filler–filler network; (2) the improved interfacial interaction between rubber and filler which strengthened the filler–rubber network due to the restriction of the mobility of rubber. Nevertheless, other results^{35,36} showed that if graphene oxide and carbon nanotubes are surface grafted with small molecules or polymer chains, the “Payne effect” may become more prominent. It is because these grafted ligands have much weaker interfacial interaction with the polymer matrix, leading to the formation of the additional filler–filler networks compared with the pristine case. Meanwhile, chain modification can also result in a much weaker “Payne effect” compared with the untreated at the same loading of silica fillers.³⁷ Zhong *et al.*³⁸ also found that the non-linear strain-dependent lightly increases *via* vulcanization accelerator grafted the silane modified silica compared to the silane modified, which is induced by the slippage of adsorbed polymer chains on the filler surface during tension.

On the other hand, Gregory *et al.*³⁹ used molecular dynamics simulation to investigate the viscoelastic properties of polymer nanocomposites composed of polymer-grafted NPs. Results illustrated that the moduli decreased with increasing NP size, and increased with the increase of the NP loading, graft length, grafting density and graft/matrix affinity, which results from the additional distortion of the shear field in the polymer matrix caused by the addition of NP grafts into the polymer matrix. Liu *et al.*⁴⁰ found that polymer nanocomposites filled with end-



grafted nanoparticles showed a weaker strain-induced nonlinear behavior compared to NPs without any surface modification *via* molecular dynamics simulation. The phenomenon is more obvious with the increase of the grafted chain length. They further conjectured that the polymer end-grafting of the NPs can enhance the brush/matrix interface thickness, and correspondingly improve the dispersion state and the physical interfacial interaction.

In this work, we focus our attention on the effect of the chemical coupling between NPs and polymers on the viscoelasticity through a non-equilibrium molecular dynamics (NEMD) simulation method. By fixing the filler volume fraction equal to 26.5%, we systematically change the physical parameters such as the interfacial chemical coupling extent, the temperature, and the dynamic periodic shear frequency. Note that the physical interactions between nanoparticles–nanoparticles (ε_{nn}) and between polymer–polymer (ε_{pp}) are set to be weakly attractively, while we vary the physical interaction strength between polymer and NPs (ε_{np}). The viscoelastic properties of the simulated PNCs under oscillatory shear are obtained with the “Payne effect” decreasing as the chemical coupling increases. The mechanism is illustrated by analyzing the structural parameters such as the distribution function of polymers around NPs and dynamic structural evolution. The reasons of the reduction of the “Payne effect” in this work can be summarized to two aspects: (1) with the introduction of the chemical coupling, the polymers become more tightly to NPs which leads to the reduction of filler–filler network; (2) the variation of interfacial physical interaction as periodic shear proceeds which is corresponding to the adsorption–desorption reversible process.

2. Simulation model and method

A standard Kremer and Grest⁴¹ bead-spring model was chosen to represent the polymer chain. Each bead in the polymer chain has the size of σ and the mass of m . Each polymer chain consists of 30 beads, which makes sure that the size of polymer chains is comparable with that of the NPs. And it has been shown that the chains with this length exhibit the static behavior characteristic of long polymers in the melt.⁴⁰ The adjacent beads on a polymer chain are connected with each other *via* a strong bonded interaction, finite extensible nonlinear elastic (FENE) potential,

$$V_{\text{FENE}} = -0.5kR_0^2 \ln \left[1 - \left(\frac{r}{R_0} \right)^2 \right], \quad (1)$$

where $k = 30\varepsilon/\sigma^2$ is the spring constant (ε is the energy unit), r is the distance between bonded beads and $R_0 = 1.5\sigma$ is the maximum extent of the bond length. All the non-bonded interactions between all the polymers and nanoparticles (NPs) were described using a modified Lennard-Jones (LJ) potential,⁴²

$$U_{ij}(r) = \begin{cases} 4\varepsilon_{ij} \left[\left(\frac{\sigma}{r_{ij} - r_{\text{EV}}} \right)^{12} - \left(\frac{\sigma}{r_{ij} - r_{\text{EV}}} \right)^6 \right] - U(r_{\text{cutoff}}) & 0 < r_{ij} - r_{\text{EV}} < r_{\text{cutoff}}, \\ 0 & r_{ij} - r_{\text{EV}} < r_{\text{cutoff}}, \end{cases} \quad (2)$$

where ε_{ij} is the energy parameter, r_{ij} is the distance between interacting sites i and j , r_{EV} represents that the volume effect are excluded for the two interacting sites, and r_{cutoff} is the distance ($r_{ij} - r_{\text{EV}}$) at which the potential is truncated and shifted. All the LJ energy parameter between polymers were set to $\varepsilon_{pp} = 1.0\varepsilon$, which is corresponding to a weak attraction. The cut-off distance of polymer–polymer was taken as $r_{\text{cutoff}} = 2 \times 2^{1/6}\sigma$ to include the attractive part, since it has been reported that a negative thermal expansion coefficient would occur without including such an attraction.⁴³ The filler NPs were also modeled as LJ spheres with diameters equal to 4σ and mass equal to $64m$. While the polymer–NP and the NP–NP interactions were truncated and shifted at $r_{\text{cutoff}} = 2.5\sigma$. The r_{EV} is zero for polymer–polymer interaction, while it is set to 1.5σ for polymer–NP and 3.0σ for NP–NP interactions, respectively. Reduced LJ units where σ , ε , and m were all set to unity, which means that all the calculated quantities are dimensionless. The number of the introduced NPs is 150 with the filler volume fraction equal to approximately 0.265.

It should be noted that when mapping the simulation units to experimental ones, the interaction parameter ε is set to be about 2.5–4.0 kJ mol⁻¹ depending on specific polymers.^{44–46} According to the reported mapping process,^{47,48} the modeled NP with the diameter equal to 10σ generally maps to silica particles with the diameter of 12.8–28.8 nm. Therefore, the NP with the diameter equal to 4σ in the present simulation model approximately corresponds to silica particles with the diameter of 5–12 nm. Moreover, it is generally assumed that one polymer bead with a diameter of 1 nm roughly corresponds to 5 repeating units of polyethylene for that the carbon–carbon bond length equal to 0.154 nm in the polymer physics. It indicates that one polymer bead with the diameter of 1σ in the current model corresponds to 3–6 bonds of a realistic polymer chain. Furthermore, when mapping the simulation time unit to experimental scale, τ ranges from 1.8 ps for PTHF to 66 ps for PE.⁴⁰ Thus, our simulation is considered to fall within a realistic range of experiments.^{49,50}

After the fillers were dispersed in the polymer matrix, the chemical coupling between fillers and polymers were introduced. The chemical coupling is composed of the chemical bonds which are randomly created between polymers and NPs, when the distance between them becomes smaller than 2.75σ . The process of the chemical coupling continues until the pre-set interfacial chemical coupling ($N_b = 0$ to 30) is reached. By statistically averaging the number of connected NPs on each polymer chain, we found that the chemical connection points are evenly distributed along the chain backbone. The chemical bonds between polymer and fillers were described as the same FENE potential with the parameters setting to $k = 30\varepsilon/\sigma^2$ and $R_0 = 1.5\sigma$. We controlled the number of the chemical bonds



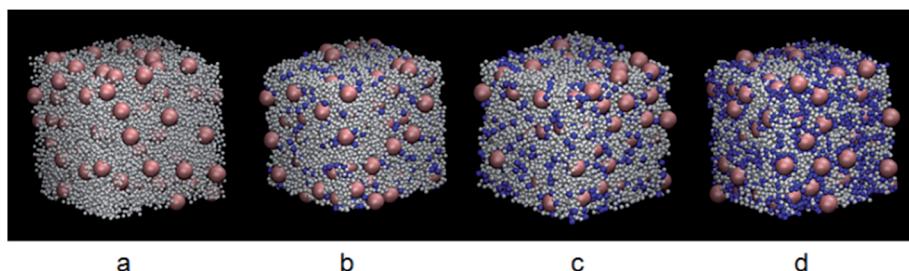


Fig. 1 Morphologies of simulation systems with different chemical coupling. (a), (b), (c) and (d) represent systems in which each filler generates 0, 5, 10, and 30 chemical bonds with the surrounding polymer chains, respectively. White, pink, and blue spheres are shown on behalf of the polymers, the nanoparticles dispersed in the polymers and the polymer beads which are chemically bonded to fillers.

between polymers and each NP equal to 0 to 30 ($N_b = 0$ to 30), respectively, as shown in Fig. 1. Note that the blue beads represent those polymer beads, which are chemically bonded to the NPs.

At the beginning, NPs and polymers were placed in a big simulation box with a periodic boundary condition adopted in all three directions. The temperature was fixed at $T = 1.0$ which is above the glass transition temperature ($T = 0.41$) of polymer chains by using the Nose–Hoover thermostat and barostat. The NPT ensemble of 5×10^6 steps ($5 \times 10^3 \tau$) was adopted here to ensure the number density of the polymer equal to around $\rho = 0.85$, which corresponds to the density of polymer melts. The velocity-Verlet algorithm was applied to integrate the equations of motion with a timestep of $\delta_t = 0.001$, where the time was reduced by the LJ time (τ). Following this, chemical bonds between NPs and polymers were generated if their distances were less than 3.0 in the NVT ensemble. The process could be monitored by collecting the statistics of the total number of chemical bonds. After adequate equilibrium of the above system, NEMD simulations in the canonical (NVT) ensemble were performed at $T = 1.0$ and the algorithm for the oscillatory shear was realized using the SLLOD equations of motion. The shear strain and time-dependent shear rate were respectively expressed as

$$\dot{\gamma}_{xy}(t) = \gamma^0 \sin(2\pi\nu t), \quad (3)$$

$$\dot{\gamma}_{xy}(t) = 2\pi\nu\gamma^0 \cos(2\pi\nu t), \quad (4)$$

where γ^0 and ν are the oscillatory shear strain amplitude and shear frequency, respectively, and t is the time. During the shearing process, the shear stress response at each time step can be expressed by a sinusoidal function:

$$\sigma_{xy}(t) = \gamma^0 [G' \sin(2\pi\nu t) + G'' \cos(2\pi\nu t)], \quad (5)$$

where G' and G'' are the storage (elastic) modulus and the loss (dissipative) modulus, respectively. The loss factor can be calculated from their ratio ($\tan \delta = G''/G'$). In the current simulations, shear frequency in most simulation systems is $0.01\tau^{-1}$, followed from the literature.⁵¹ Furthermore, we adjusted the frequency to $0.001\tau^{-1}$ for comparison. The shear strain amplitude γ^0 was varied to study the viscoelasticity of the systems as a function of the shear strain. NEMD simulation for

each system consisting of 10^7 steps ($10^4\tau$) to guarantee the consistency, followed by 10^6 steps ($10^3\tau$) to sample the properties. Then the structural and dynamic data were collected for ensemble average. All the MD simulations were conducted using the large scale atomic/molecular massively parallel simulator (LAMMPS) molecular dynamics package.⁵²

3. Results and discussion

3.1 Viscoelastic behavior

Following our previous work,²⁵ the “Payne effect” is related to the polymer–filler interaction and the filler loading. Firstly we examine the change of the storage moduli, the loss moduli and the loss factor as a function of the strain amplitude for various extent of interfacial chemical coupling, by fixing the interfacial physical interaction $\varepsilon_{np} = 8.0$, the volume fraction of filler equal to 26.5% and the dynamic shear frequency $\nu = 0.01\tau^{-1}$, as shown in Fig. 2. It can be observed from Fig. 2(a) and (b) that both G' and G'' decrease with the increase of the strain amplitude, and enhance with the increasing interfacial chemical coupling. The decrease of the storage moduli becomes prominent at the strain amplitude greater than 0.02, while at the highest interfacial coupling the storage moduli becomes increased at large strain amplitude. We will interpretate this observation later. Moreover, the beginning strain for the decrease of the loss moduli is around $\gamma^0 = 0.1$, which demonstrates a relative lag response on the shear strain compared to the storage moduli. Fig. 2(c) reveals that the loss factor decreases with the interfacial chemical coupling, indicating less molecular friction and dissipation.

We have compared our work with some experimental results in Fig. 2. Jiang *et al.*⁵³ investigated the effect of the non-covalent modification of carboxylated multi-walled carbon nanotubes (MWCNT) by 1-carboxyethyl-3-methylimidazolium bis(trifluoromethyl sulfonyl)imide (CMI), on the polymer–filler interactions, filler dispersion, and mechanical properties of polychloroprene (CR)/MWCNT composites. Fig. 2(d) shows that by with increasing MWCNT/CMI ratio from 1 : 0 to 1 : 2, the “Payne effect” of CR composite becomes more pronounced, indicating CMI can effectively strengthen the interactions between MWCNT and CR. However, when the MWCNT/CMI ratio is as high as 1 : 8, the $\Delta G'$ decreases because CMI has an obvious plasticizing effect on CR/m-MWCNT composites when



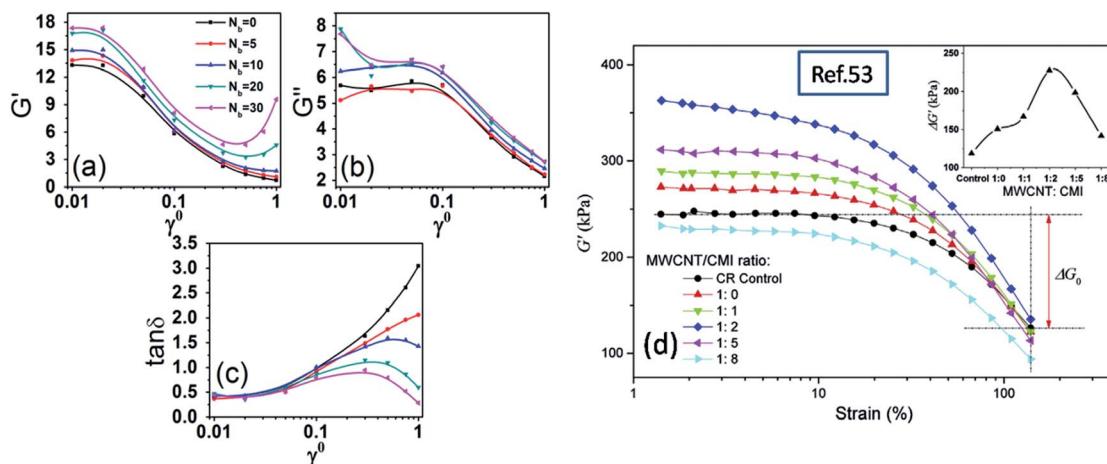


Fig. 2 (a) Storage modulus G' , (b) loss modulus G'' , and (c) loss factor $\tan \delta$, as a function of shear strain amplitude for the systems with different polymer–particle chemical coupling on temperature at 1.0 and frequency at $0.01\tau^{-1}$. (d) Strain dependence of G' for CR/m-MWCNT composites with different MWCNT/CMI ratios (ref. 53).

the CMI content is very high. It can be assumed that the increase of the physical interaction between polymer and filler enhances the “Payne effect”, and the excessive force such as the chemical interaction may weaken the “Payne effect”.

3.1.1 Effect of the temperature. We examined the effect of the temperature on the viscoelastic behavior in Fig. 3, by setting the temperature equal to 0.5, 1.0, 1.5 and 2.0, respectively. With the increase of the temperature, the storage moduli decreases and the non-linear behavior weakens, which results from the increased mobility of polymer chains. Moreover, for systems at high temperatures (1.5 and 2.0), the decrease of the storage

moduli with the strain becomes slight because of the relative low initial moduli at the small strain, which is corresponding to the good mobility of polymers and NPs, which could be accelerated by the imposed shear flow. It should be noted that the storage moduli appears a rise at the large strain ($\geq 30\%$) under all the temperatures, which is attributed to the chemical coupling contributing to the stress at the large deformation. Furthermore, the extent increases with the chemical coupling increasing and then approaches to a plateau for systems with the same chemical coupling, for example, the storage moduli increases to about 9 to 10 for all the systems of $N_b = 30$.

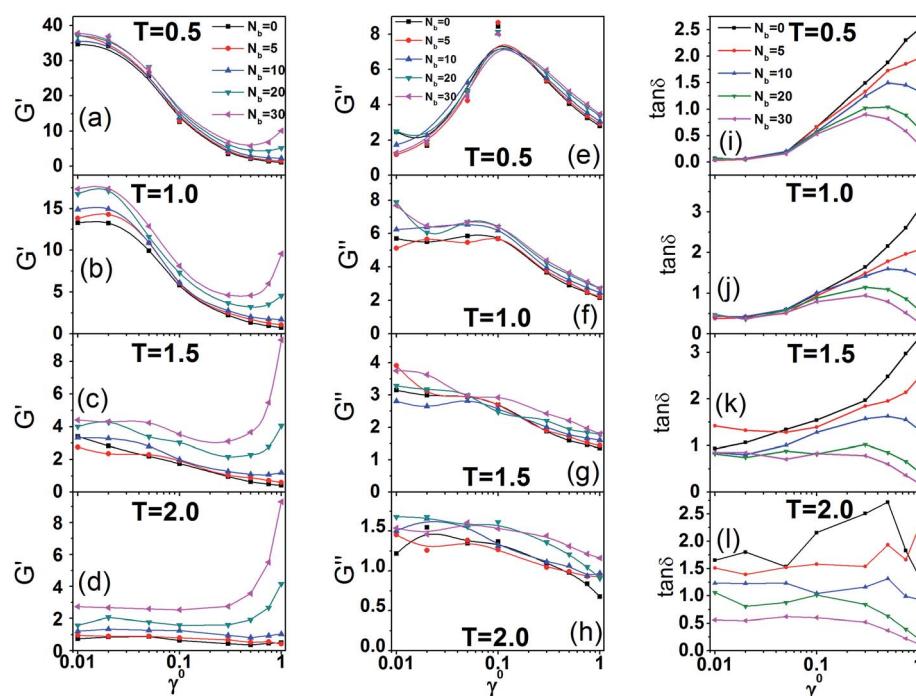


Fig. 3 (a–d) Storage modulus G' , (e–h) loss modulus G'' , (i–l) loss factor, as a function of shear strain amplitude with various chemical coupling at different temperatures.



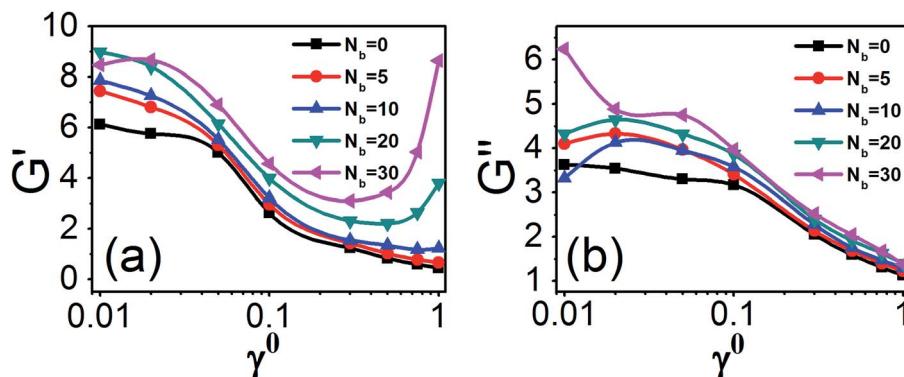


Fig. 4 (a) Storage modulus G' and (b) loss modulus G'' , as a function of shear strain amplitude with various chemical coupling at the shear frequency of $0.001\tau^{-1}$.

Interestingly, the loss moduli at $T = 0.5$ shows a different variation trend compared with others cases, which is due to that the polymer chains become almost immobilized at small strain, because the glass transition temperature of this model is approximately 0.5.⁵⁴ The loss moduli gradually increases as the shear strain increases to 10% at which point a balance is achieved between the restriction on polymer mobility caused by low temperature and the enhancement of the polymer mobility caused by the shear force. The decrease of the loss factor at the large strain results from the increase of the storage moduli as well.

3.1.2 Effect of the shear frequency. Similar results have been obtained for the simulation systems with the shear frequency declining from $0.01\tau^{-1}$ to $0.001\tau^{-1}$ at temperature 1.0 shown in Fig. 4. Storage moduli at low frequency obviously are lower than the values at higher frequency $0.01\tau^{-1}$ shown in Fig. 2, because polymer chains have more time to move. Meanwhile, we find that the storage moduli also increase to an approximated high value for the system with the same chemical coupling at large shearing strain under two frequencies.

3.1.3 Effect of the interfacial physical interaction. To better illustrate this point, a series curves of storage moduli for various interfacial physical interactions between NPs and polymers at different chemical coupling were obtained in Fig. 5. Storage moduli increase distinctly as the ε_{np} increases at any chemical coupling and only slightly increase with the increasing chemical

coupling at any interfacial physical interactions. We can also observe an increase of storage moduli up to the value of 9.0 to 10.0 at large shearing strain of 1 in systems $N_b = 30$ whatever the ε_{np} is. We will interpret this increase later by analyzing the structures.

3.1.4 Explanation of the “Payne effect”. The magnitude of the “Payne effect” was qualified using $(G'_{0.01} - G'_{0.3})/G'_{0.01}$ as shown in Fig. 6, where $G'_{0.01}$ and $G'_{0.3}$ represents the storage modulus at small strain ($\gamma^0 = 0.01$) and moderate strain ($\gamma^0 = 0.3$), respectively. The moduli at moderate strain ($\gamma^0 = 0.3$) was chosen in the formula aforementioned to ensure a uniform comparison between different chemical coupling systems because storage moduli with high chemical coupling increase again at large strain which should be interpreted by other mechanism. $(G'_{0.01} - G'_{0.3})/G'_{0.01}$ represents the relative decline of storage moduli from small to moderate strain, which is corresponding to the magnitude of the “Payne effect”. It is found that the ordinate value decreases with the abscissa value increases, which means the magnitude of the “Payne effect” weakens with chemical coupling increasing in both Fig. 6(a) and (b). The ratios show an overall decline with temperature rising from 0.5 to 2.0, the shear frequency falling from $0.01\tau^{-1}$ to $0.001\tau^{-1}$ and the interfacial physical interaction rising from 12.0 to 2.0, which results from the decreasing of strain-dependent of systems originated from the better mobility of polymers. While there is an inconsistent point that the ordinate

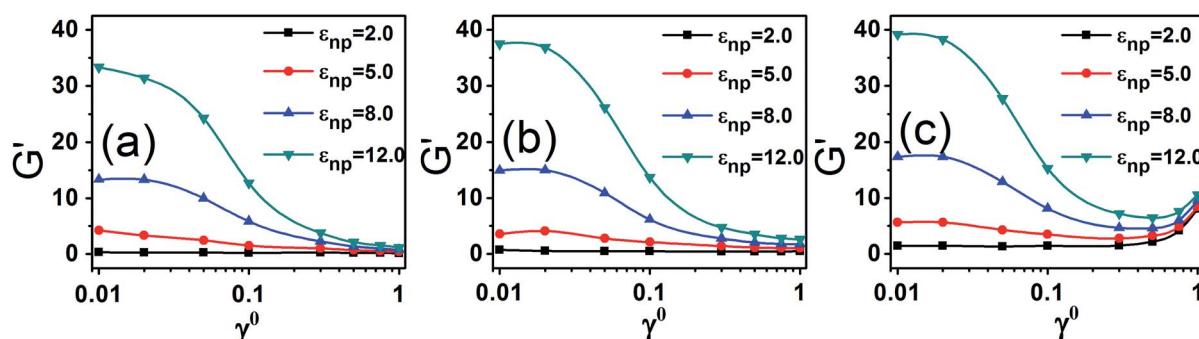


Fig. 5 Storage modulus G' as a function of shear strain amplitude with various interfacial physical interaction between NPs and polymers (ε_{np}) in systems with different chemical coupling, (a) $N_b = 0$, (b) $N_b = 10$, and (c) $N_b = 30$.



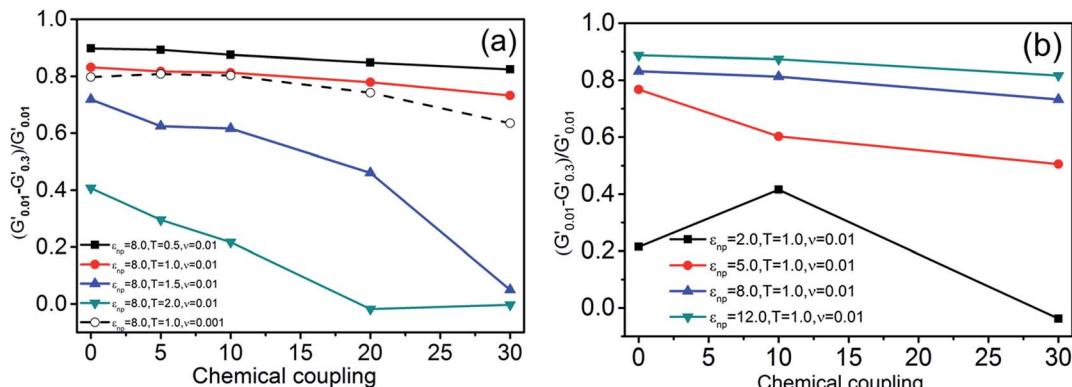


Fig. 6 $(G'_{0.01} - G'_{0.3})/G'_{0.01}$ as a function of chemical coupling represents qualified magnitudes of Payne effect for the systems, (a), with $\varepsilon_{np} = 8.0$ at four temperatures ($T = 0.5, 1.0, 1.5, 2.0$) and two shear frequency ($\nu = 0.01, 0.001$) and (b), with different ε_{np} at certain temperature of 1.0 and the shear frequency of $0.01\tau^{-1}$.

value increases from $N_b = 0$ to $N_b = 10$ at $\varepsilon_{np} = 2.0$, and it is attributed to the low modulus at this interfacial physical interaction. Moreover, the chemical coupling could increase the storage moduli greatly at this low physical interaction, and the $(G'_{0.01} - G'_{0.3})/G'_{0.01}$ decreases more compared with that of the system of $N_b = 0$. The results above demonstrate that chemical coupling between NPs and polymers play an important role on decreasing the magnitude of the “Payne effect”.

3.2 Structures

In order to realize the essence of role of chemical coupling between polymers and NPs, a series of simulated processes were performed to analyze the internal structures of simulated

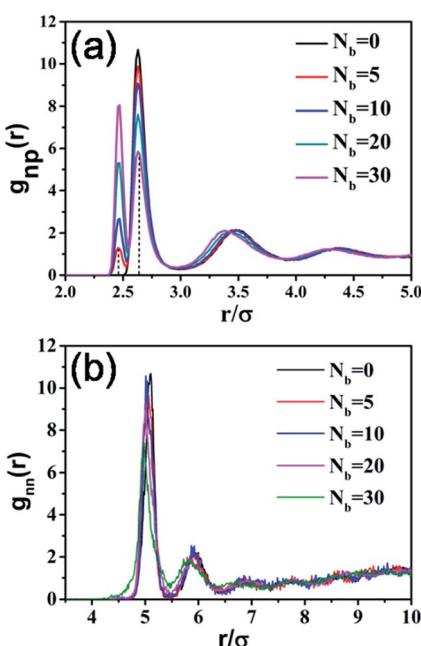


Fig. 7 (a) Radial distribution function of polymers around NPs, $g_{np}(r)$; (b) radial distribution function of NP-NP, $g_{nn}(r)$ in the system with $\varepsilon_{np} = 8.0$.

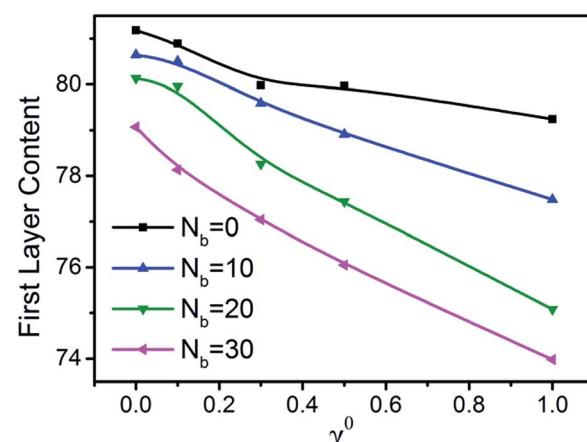


Fig. 8 Plot of the first layer content with respect to the shear amplitude at $\varepsilon_{np} = 8.0$.

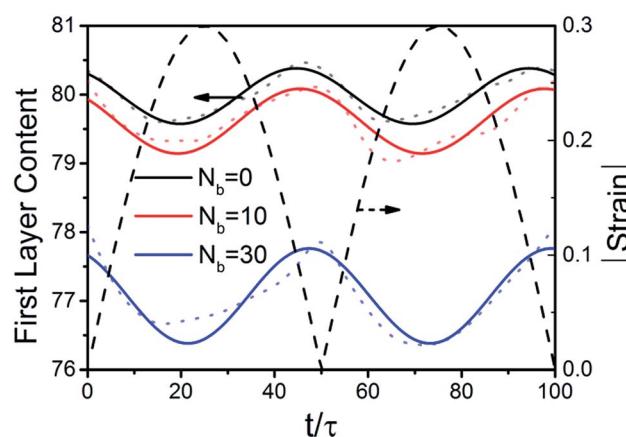


Fig. 9 Plot of the first layer content with respect to the oscillatory shear with the amplitude of 30% at $\varepsilon_{np} = 8.0$. The periodic dash line represents the shear strain with respect to the time. The solid lines are the fitting curves according to the dot lines which represent the original data.

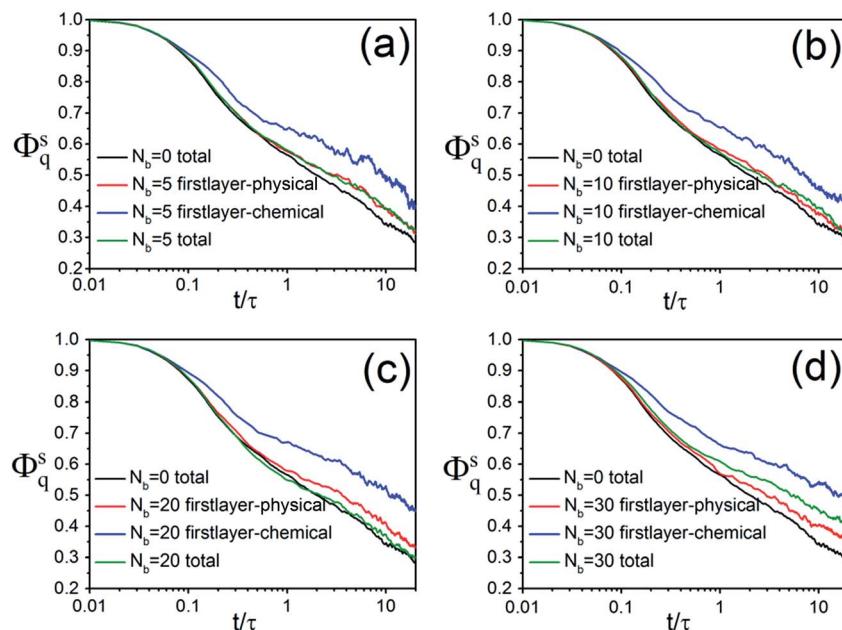


Fig. 10 Incoherent intermediate dynamic structure functions (Φ_q^S) of the polymers with or without chemical coupling located in the first polymer layer and NPs at $\varepsilon_{np} = 8.0$. (a) $N_b = 5$, (b) $N_b = 10$, (c) $N_b = 20$, and (d) $N_b = 30$.

systems. Fig. 7(a) shows distribution states of polymers around NPs in unsheared systems at five chemical coupling degrees using radial distribution functions (RDFs). The value of the peak at $r = 2.65\sigma$ which caused by physical adsorption means that there is a packed polymer layer (the first layer) referred to “bound rubber” or “bound shell” in rubber materials^{55,56} and it declines consistently with the increasing chemical coupling. At the same time, a new peak at $r = 2.45\sigma$ appears and increases with the increasing chemical bonds. It is interpreted that the chemical coupling converse partial interactions between NPs and polymers from physical adsorption to chemical bonding which makes polymers bound to the NPs’ surfaces more tightly. Furthermore, Fig. 7(b) shows the radial distribution functions (RDFs) of NPs around NPs, which reflects the filler dispersion

state. The remarkable peak emerging at about $r = 5\sigma$ indicates that there is a layer of chain molecules between NPs. The remarkable peak emerging at about $r = 5\sigma$ indicates that there’s a layer of polymer molecules between NPs. The value of the peak gradually declines and moves left as the N_b increases, which manifests that the chemical coupling between polymer and NPs makes the NPs closer to each other.

Fig. 8 shows the first layer content of polymers which represents the average number of polymers within a distance of σ from each NP surface with respect to the shear amplitude in quiescent condition. That means if the polymer molecule is bound to any NP within a distance of σ , it will be classified as polymers in the first layer around NP. We average the number of

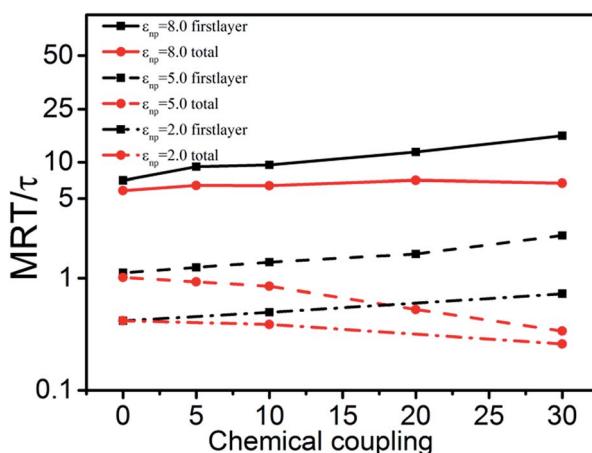


Fig. 11 Mean relaxation time (MRT) of polymers in the first layer and that of total polymers as a function of chemical coupling at varies ε_{np} .

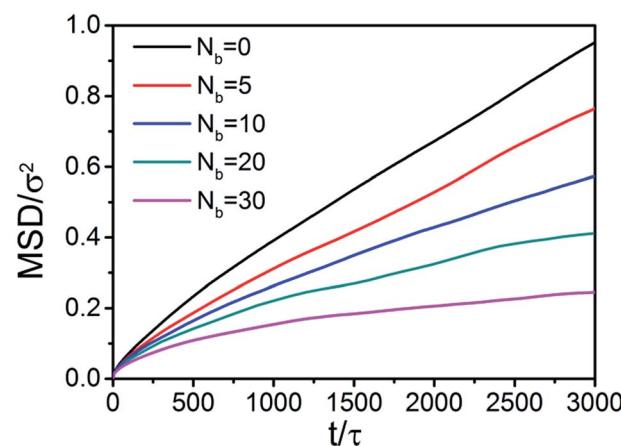


Fig. 12 Mean square displacements (MSD) of polymer chains as a function of equilibrium time for systems with different chemical coupling.

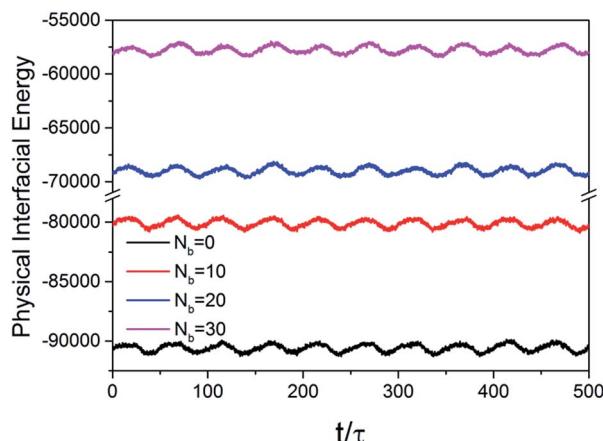


Fig. 13 Plot of the physical interaction energy between polymers and NPs as a function of shearing time at shear strain of 30% and $\epsilon_{np} = 8.0$.

this type of polymer molecules for each NP, which is referred to the value of the first layer content. The value decreases with the shear amplitude increasing uniformly, which means that more polymers separate from the NP surface as the shear amplitude increases. It is also found that the overall value decreases with the increasing chemical coupling. It demonstrates that more polymers without chemical coupling are crowded out of the layer within the distance of σ from the NP surface as the chemical coupling increases, which indicates polymers are more tightly to NPs' surface.

The first layer content generates a periodic variation that the value decreases with the strain increasing in a shearing period of time equal to 100τ as shown in Fig. 9. The lowest value appears at the maximum strain amplitude, which means that the content of polymer chains desorbing from the surface of the NPs becomes the biggest at that point.

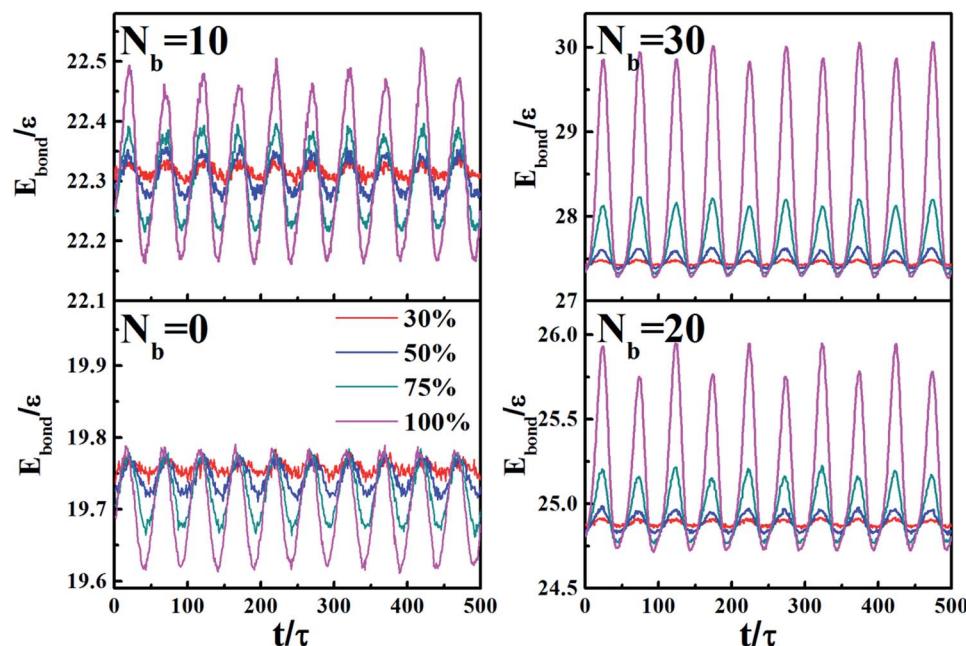


Fig. 14 Bond energy (E_{bond}) of chemical bonds between polymers and NPs during the oscillatory shear, for $N_b = 0, 10, 20$, and 30 .

Fig. 10 shows the incoherent intermediate dynamic structure functions (Φ_q^s) of polymers which reflects the mobility of polymers. The faster the values decline, the higher the mobility of polymers is. It is obvious that polymers with chemical coupling (the blue lines) have weaker mobility compared to polymers without chemical coupling (the red lines) in the first layer around NPs in each chemical coupling system. Moreover, the function of system $N_b = 0$ which has no chemical coupling reduces the fastest corresponding to the highest mobility of polymers. Furthermore, the functions of total polymers always behave the same as the functions of polymers without chemical coupling in each system except the system $N_b = 30$, which is due to that the considerable chemical coupling is enough to affect the total mobility.

Results of MRT curves shown in Fig. 11 demonstrate that the introduction of chemical bonds between NPs and polymers increases the relaxation time of polymers in the first layer but decreases it of total polymers. On one hand, polymers in the first layer become tight to NPs and the relaxation behavior becomes weak as the chemical coupling increases. On the other hand, for total polymers, the mean relaxation time decreases with the number of chemical bonds increasing, which is due to that the tightly limited polymers with chemical coupling around NPs weaken the physical interactions between NPs and all the rest polymers.

Results of mean square displacements which represent the mobility of whole polymer chains are shown in Fig. 12. The mobility of whole polymer chains become weaker with the increasing chemical coupling, which exhibits a similar case to the mobility of polymer monomers results shown in Fig. 7.

3.3 Energy evolution under the oscillatory shear

The interfacial potential energy which reflects the polymers around NPs were monitored under oscillatory shear and

presented in Fig. 13. The energy exhibits periodically changes with time obviously which is synchronous with shear strain, which is attributed to polymer chains debonding to particle surface under shear. Partial polymer monomers were bonded to NP surface through chemical coupling and these monomers could not debond under shear. The absolute value overall energy gradually decreases with the increasing chemical coupling from $N_b = 0$ to $N_b = 30$, which is corresponding to the gradually weakened physical adsorption–desorption process. It suggests that chemical bonds between polymer and NPs hinder the interfacial physical bonding–debonding effect and consequently weaken the downward trend of G' .

Fig. 14 directly shows the corresponding energy changes of total bonds including bonds on polymer chains and chemical coupling bonds as the shear cycles at varying amplitudes. The cyclic shear period is 100τ , which means the simulation box deforms and returns to the initial state every 50τ with the maximum value of bond energy appears at every maximum strain. Meanwhile, by comparing ordinates of the four graphs, it is found that the peak energy has a significant increase as the chemical coupling increases. Moreover, the changing model of above the equilibrium value depends on the chemical coupling. It illustrates that chemical coupling between polymers and NPs obviously increase bond energy during the shear process, and it attributes to the resulting stress to prevent the destruction of the chemically bonded networks.

4. Conclusions

The effect of chemical coupling between polymers and NPs on the viscoelastic properties of polymer nanocomposites (PNCs) has been examined by coarse-grained simulation. The computational results have shown that the formation of chemical coupling weakens the “Payne effect” under various imposed conditions. The decreasing of the “Payne effect” is attributed to the three points: (i) chemical bonds decrease the physical bonding–debonding of polymer chains from the filler surfaces; (ii) the polymer-shell-bridged network becomes more difficult to be destroyed; (iii) the mobility of polymers has been constrained. Moreover, the storage moduli's abrupt rise results from the chemical bond energy increasing sharply at large strain. The above simulation work provides a useful information on how chemical coupling between polymers and nanoparticles influence the “Payne effect” as well as an understanding of structure and viscoelastic properties of filled polymer nanocomposites influenced by grafting NPs that could coupled with polymer chains.

Conflicts of interest

There are no conflicts to declare.

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