It is well-known that the addition of nanoparticles (NPs) can mechanically reinforce a polymer matrix. The mechanism of reinforcement, however, remains in debate. At one extreme, Long and co-workers used the fact that chain immobilization occurs around NPs; they suggested that mechanical reinforcement results when the NPs with the “bound” glassy layer percolate. A second scenario proposed by Sternstein and elaborated by Goritz is that the NPs form a flexible network, with the polymer chains forming the “bridges” between the NPs. A simple way to resolve the mechanism of reinforcement is to study NPs in a polymer melt far above the glass transition temperature. The linear viscoelastic properties of polymer nanocomposites can be quantitatively predicted by a parameter-free model in which the stress is a simple sum of the NP contribution modeled using critical percolation, while the polymer part is enhanced by the presence of particles, owing to hydrodynamic interactions. The phase diagram at the right shows that small NPs are needed to achieve the stronger reinforcement from glassy bridges at reasonable particle loadings.

lower fractal dimension is reported for fumed silicas, a fractal dimension of $d_f = 2.4\pm0.3$ is reported for aggregated silica-NPs, consistent with the fractal dimension of 2.53 for critical percolation. Compared with experiments, simulation can more easily define the percolated structure, but this result is found to strongly rely on the assumed interparticle potential. According to computer simulation combining mean-field theory with Monte Carlo simulations, critical percolation theory is the relevant theory here in the strong interaction limit. The second question is on dynamics, which are also controlled by the interaction potential between NPs, but in the strong-interaction limit that we focus on here, where the interaction far exceeds the thermal energy $k_B T$, the bridges between particles become effectively permanent on the time scale of our measurements.

Here, an analytical theory is developed to predict the linear viscoelasticity corresponding to the sol–gel transition of NPs based on critical percolation. We assume that the complex modulus of the nanocomposites is the sum of two contributions: one from the surrounding polymer matrix and the other from the particle network bridged by polymer chains, that is, $G(\omega) = G_{\text{polymer}}(\omega) + G_{\text{NP}}(\omega)$.

The simplest assumption, which follows from recent computer simulations, is that the modulus of the polymer with the NPs has the same mode distribution of relaxation times as the bulk polymer: $G_{\text{polymer}}^*(\omega) = k G_{\text{bulk}}(\omega^*)$, where $k(>1)$ is a factor characterizing the increase in the polymer’s modulus due to the presence of the NPs. In general, $k$ can be expressed as a function of filler volume fraction $\phi$, $k = G_{\text{polymer}}^*/G_{\text{bulk}}^* = 1 + c_1 \phi$.

ABSTRACT: Nanometer-sized particles that are well dispersed in a polymer melt, presumably due to strongly favorable particle–polymer interactions, can form fractal structures via polymer bridging, leading ultimately to a nanoparticle (NP) network analogous to a colloidal gel. The linear viscoelastic response of polymer nanocomposites can be quantitatively predicted by a parameter-free model in which the stress is a simple sum of contributions from the polymer matrix and the fractal NP structure linked by bridging polymer chains. The NP contribution is modeled using critical percolation, while the polymer part is enhanced by the presence of particles, owing to hydrodynamic interactions. The phase diagram at the right shows that small NPs are needed to achieve the stronger reinforcement from glassy bridges at reasonable particle loadings.
The number density of clusters of terms having unknown thickness was reported to vary between studies using TGA and dielectric spectroscopy suggest existence of this interfacial bound layer has been suggested by segmental mobility of polymer chains is greatly reduced. The bound layer, the phenomenological equations such as the Guth–Gold equation or Krieger–Dougherty equation are frequently used as substitutes.\(^{24,25}\)

Here, an effective volume fraction \( \phi_{\text{eff}} \) is calculated by accounting for the fact there exists an interfacial region where segmental mobility of polymer chains is greatly reduced. The existence of this interfacial bound layer has been suggested by various experimental techniques.\(^{14,26–32}\) The bound layer thickness was reported to vary between \( \delta = 1–5 \) nm depending on the size of spherical NPs, chemistry of polymer, and strength of NPs/polymer interaction. In good agreement, our recent studies using TGA and dielectric spectroscopy suggest \( \delta = 1–4 \) nm for the silica/poly(2-vinylpyridine) nanocomposites, the model systems studied in this letter.\(^{14,33}\) Due to the immobile particle interactions, the static and dynamic scaling of critical percolation:\(^{34,35}\)

\[
G^*_{\text{polymer}} = \kappa G^*_{\text{bulk}} \quad \text{(1a)}
\]

\[
\kappa = 1 + 2.5 \phi_{\text{eff}} + 14.1 \phi_{\text{eff}}^2, \quad \text{with } \phi_{\text{eff}} = \phi(r + \delta)^3/r^3 \quad \text{(1b)}
\]

The particle network part of the modulus is expressed using the static and dynamic scaling of critical percolation:\(^{34,35}\)

\[
P(n) \sim n^{-\tau} \quad \text{for } 1 \leq n \leq n_{\text{ch}} \quad \text{with } \tau = 2.2 \ \text{for critical percolation; } n_{\text{ch}} \sim \phi_i \phi_i^{-1/\sigma} \quad \text{with } \phi_i \text{ being the volume fraction of NPs at the gel point, and } \sigma = 0.45 \ \text{for critical percolation. (2)}
\]

The module at the relaxation time of a cluster of \( n \) NPs is proportional to the fraction of clusters having more than \( n \) NPs: \( G_{\text{PN}}(n) \sim n^{-\tau} \int_{d_{\text{char}}}^{\infty} \rho(n \phi)^{d_{\text{char}}} \rho(n \phi)^{d_{\text{char}}} \), where \( n_{\text{ch}} \) is the diffusion coefficient of a cluster containing \( n \) NPs.

The modulus of a percolated network can be written as eq. 2a when the NPs have not yet formed a percolated network (i.e., below the gel point), and eq. 2b when the network has been formed (above the gel point):\(^{34}\)

\[
G^*_{\text{PN}}(\omega) = \frac{\rho G_{\text{PN}}}{1 - n_{\text{ch}}^{-1}} \int_{d_{\text{char}}}^{\infty} \left( e^{-\phi_j/(\phi_{\text{eff}}^2 + \phi_j)} \right) \frac{d\phi}{\phi_j} \quad \text{for } \phi < \phi_i \quad \text{(2a)}
\]

\[
G^*_{\text{PN}}(\omega) = \frac{\rho G_{\text{PN}}}{1 - n_{\text{ch}}^{-1}} \int_{d_{\text{char}}}^{\infty} \left( e^{-\phi_j/(\phi_{\text{eff}}^2 + \phi_j)} \right) \frac{d\phi}{\phi_j} + n_{\text{ch}}^{-1} \quad \text{for } \phi \geq \phi_i \quad \text{(2b)}
\]

with (I) \( d_i = 2.5, \tau = 2.2 \) for critical percolation. (II) \( n_{\text{ch}} = 1/\beta - \phi_i^{1-1/\sigma} \), with \( \sigma = 0.45 \) for critical percolation and \( \phi_i \) specifying the volume fraction corresponding to the gel point. (III) \( \varepsilon_0 = 1/t_4 \) and \( \varepsilon_{\text{char}} = \varepsilon_{\text{char}}^{1/2 + f_{\text{bound}}/f_{\text{bound}}} \), with \( t_4 \) being the terminal relaxation time of the surrounding polymer matrix.

The former equation is reasonable when particle size is larger than the entanglement mesh size of the polymer matrix.

In eq. 2, the amplitude \( G_{\text{PN}} \) is the only parameter to be specified. To this end, focus is placed on the modulus per particle, that is, \( G_{\text{PN}} \) normalized by \( \rho \varepsilon_{\text{char}} k_B T \), where \( \varepsilon_{\text{char}} \) being the number density of particles.\(^{36–38}\) Previous studies revealed a power law dependence \( \varepsilon_{\text{char}} \sim \phi^{m}, \) with \( m \approx 3–5 \) for strong particle–particle interactions. In this study, the particles bridged by polymer chains are assumed to store the stress, with \( G_{\text{PN}} = \rho \varepsilon_{\text{char}} k_B T n_{\text{PN}} \) and \( n_{\text{PN}} \) the number of bridges per particle. de Gennes predicted that the density (number per area) of bridging chains is \( \sim h^{-2} \), where \( h = r f(\phi) \) is the distance between two plates of size \( r \). Then, the number of bridges per particle is \( n_{\text{PN}} = S/h^2 \sim 4\pi r^2/(r f(\phi))^2 \), with \( S = 4\pi r^2 \) the surface area of each particle.\(^{39}\)

Considering a particle should occupy a volume of \( V \sim r^3/\phi \), the average distance between particle surfaces would be \( h_{\text{av}} = \frac{V^1/3}{r} \sim 2r \phi^{-1/3} - 1 \), that is, \( f(\phi) = 2r^{-1/3} - 2 \).

Nevertheless, particles are not uniformly distributed and bridge density should be controlled by the mean nearest-neighbor distance. Torquato, Lu, and Rubinstein analyzed the nearest neighbor probability density function to obtain an expression allowing numerical calculation of the mean nearest-neighbor distance as\(^{40}\)

\[
h_{\text{char}} = 2r_1 \int_0^{\infty} \exp \left(-\phi(8(1 + \phi)(x^3 - 1) - 6\phi(3 + \phi)) (x^3 - 1) + 12\phi^2(x - 1)\right) dx
\]

where \( h_{\text{char}} \) can be analytically approximated\(^{40}\) for \( \phi > 0.2 \) as \( h_{\text{char}} \approx r(1 - \phi)^3/(6\phi(2 - \phi)) \), see equation 6.12 of ref 40. (IV) the amplitude \( G_{\text{PN}} = \rho \varepsilon_{\text{char}} k_B T n_{\text{PN}} \) with \( n_{\text{PN}} \approx 4\pi r^2/h_{\text{char}}^2 \).

This expression is valid only if \( h_{\text{char}} \) is in the range between segment size \( b \) and chain size \( R \) of the matrix. Since the mean size of bridging chains is \( h_{\text{char}} \), the volume per bridge is \( (h_{\text{char}}/b^2) \times b^3 \), and the ratio of volume of bridges per particle and particle volume \( V_{\text{particle}} = 4\pi r^2/3 \sim n_{\text{PN}} \times (h_{\text{char}/b^2} \times b^3) \). \( V_{\text{particle}} = 3b/r \), meaning that the volume fraction of bridging chains is \( 3b^2/r \), which is usually smaller than \( \phi \) since \( b < r/3 \).

Figure 1 schematically illustrates \( G_{\text{polymer}} \) and \( G_{\text{PN}} \) as functions of angular frequency \( \omega \) based on critical percolation. With NPs present, \( G_{\text{polymer}} \) has the same relaxation frequency,
\( \varepsilon_\omega \) but larger amplitude compared with that of the bulk polymer, from hydrodynamic coupling with the less mobile NPs. The relaxation of the NP clusters \( G''_{\text{PN}} \) is activated by that of \( G'_{\text{polymer}} \). \( G''_{\text{PN}} \) then has a power law viscoelasticity with slope \( 2/3 \) that continues until the cutoff frequency \( \varepsilon_{\text{chc}} \) of the largest cluster, followed by either a terminal relaxation below the gel point or a network modulus above the gel point.

To test the validity of this model of mechanical relaxation for nanocomposites, silica NPs (Nissan Chemicals) with various radii \( r \) (=7, 11.5, 25, and 50 nm) were well-dispersed in a poly(2-vinylpyridine) matrix (P2VP from Polymer Source, \( M_n = 97000; M_w/M_n = 1.08; T_g = 371 \, \text{K} \)), with the weight fractions \( \omega \) of NPs varied from 0 to 50 wt \%, corresponding to volume fractions \( \phi \) up to 34\%. The P2VP is chosen due to its strong hydrogen bonding interaction to the silica surface, which facilitates the uniform dispersion of the NPs in the polymer.\(^{1,14} \)

We believe that a bound layer of polymer with a long lifetime forms at the particle surface, allowing the polymer bridges to be permanent on the time scales of our measurements.

The samples were prepared by solution casting from either 2-butanol (for \( r = 7 \) and 25 nm silica) or isopropanol (for \( r = 11.5 \) and 50 nm silica) with pyridine (Sigma-Aldrich) as dispersant.\(^{14} \) Ultrasonication was applied in solution to improve dispersion of nanoparticles, confirmed by TEM in our previous study. The cast samples were first annealed under vacuum for 7 days at 80 °C and then 10 days at 150 °C prior to being molded into \( \sim 1 \, \text{mm} \) thick and 7.9 mm diameter discs. Linear viscoelastic (LVE) measurements were conducted, under \( N_2 \), for those discs between 7.9 mm diameter parallel plates on an Advanced Rheometric Expansion System ARES-LS1 rheometer (Rheometric Scientific) at 180 °C. The linear strain amplitudes were verified by strain amplitude sweeps.

Figure 2 shows the linear viscoelastic storage and loss moduli, \( G' \) and \( G'' \), as functions of angular frequency \( \omega \) for silica NPs (\( r = 7, 11.5, 25, \) and 50 nm) dispersed in the P2VP matrix at \( T = 180 \, \text{°C} \). At the highest frequencies studied, the modulus of nanocomposites increases with the volume fraction of NPs, the increase becomes weaker with increasing particle size. At \( \omega \) lower than the terminal relaxation of the polymer, an additional stress component manifests: For low NP content (\( \phi = 6\% \)), a transition from power law to terminal relaxation can be observed, corresponding to relaxation of nanoparticle clusters below their gel point. Finally, at high content of nanoparticles, a low \( \omega \) plateau can be observed, reflecting the formation of a percolated NP-network (gel). The thin solid curves in Figure 2 are predictions from the proposed model.

The overall agreement between experimental results and model predictions is remarkable, supporting the molecular view that the NP gelation obeys critical percolation with bridging chains\(^{5-8} \) bearing the stress as long as the mean nearest-neighbor distance between particles is larger than the Kuhn length. Nevertheless, deviations can be clearly seen at high loading of small NPs, which we will understand below.

For the polymer part of the modulus, the bound layer thickness \( \delta \) gives the effective volume fraction \( \phi_{\text{el}} \). Since our recent studies suggest \( \delta = 1-4 \, \text{nm} \) for the same silica/P2VP nanocomposites, a reasonable choice \( \delta = 2 \, \text{nm} \) is made here for all the nanocomposites.\(^{14,33} \)

For the particle network component of the modulus, the gel point must first be determined. In LVE, the gel point is identified from a low frequency power law as \( G' \sim G'' \sim \omega^u \), with \( u = 2/3 \) for critical percolation.\(^{5,41-44} \) This feature allows determination of \( \phi_c \) where \( G' \) and \( G''/\tan(\pi/3) \) cross at low frequency, as explained in Figure S1 of Supporting Information, which yields \( \phi_c = 16, 22, 29, \) and 34 (±5\%) for \( r = 7, 11.5, 25, \) and 50 nm, respectively. \( \phi_c \) is plotted against \( r \) in the inset of Figure 2(a2). The powers \( G' \sim G'' \sim \omega^{2/3} \) at the gel point \( \phi_c \) are shown in thick red solid lines in Figure 2. Meanwhile, the
characteristic time for particle motion (III) \( \tau_d = 1/\dot{\gamma} \) is reasonable when NPs are considerably larger than the entanglement mesh size of the polymer matrix. Here, the diameter of particles \( 2r \geq 14 \text{ nm} \) is larger than entanglement length \( a = 13 \text{ nm} \) for P2VP (calculated from \( a^2 = M_0(R^2)/M \), with \( R^2/M = 6.6 \times 10^{-3} \text{ nm}^2 \) for P2VP in a \( \theta \) solvent and \( M_0 = 27000 \) for P2VP\(^{46} \)).

In addition, (IV) \( G_{PN} \approx v_{\text{particle}}f_0R_0n_0 \) with \( n_0 = 4\pi r^2/h_{\text{ent}}^2 \) is based on an assumption of flexible bridges: this assumption is valid only if \( h_{\text{ent}} \) is in between the segment and chain sizes of the matrix polymer, i.e., in between Kuhn length \( b \approx 1 \text{ nm} \) and the end-to-end distance \( R = 25 \text{ nm} \) (see Figure S2 of Supporting Information).\(^{3,4} \) In other words, the particles should neither be too crowded, so that the bridges remain flexible, nor too sparse, so that particle–particle bridges can form. We define two critical concentrations \( \phi^* \) and \( \phi^{**} \), at which \( h_{\text{ent}} = b \) and \( R \), respectively. Using the analytical approximation \( b = h_{\text{ent}} \approx (1 - \phi^*)^{1/2} \) (the asymptotic expression of eq 3 can be used since \( \phi^* > 0.2 \) is expected), there is an analytical solution for \( \phi^* \):

\[
\phi^* = 1 + 2b/r - b^2/(r^2g(b/r)) - g(b/r),
\]

with \( g(x) = (-8x^3 + (9x^2 - 48x + 44)^{1/2})/2 + 3x^{1/3} \) (eq 4).

In Figure 3, \( \phi^* \) is plotted against \( r \) for all nanocomposites in this study to create a phase diagram. \( \phi^* \) and \( \phi^{**} \) from numerical calculation (eq 3) are shown as the thin solid black and dashed curves, respectively, while \( \phi^* \) from eq 4 is shown as a thick solid orange curve for comparison. Both \( \phi^* \) and \( \phi^{**} \) increase with \( r \), where \( \phi^* \) separates regimes of rubbery flexible bridges and glassy bridges, and \( \phi^{**} \) separates flexible bridge and no bridge regimes. It is obvious that all the nanocomposites in this study have sufficient particle concentration for bridges to form. Nevertheless, for small particles at high loading, the bridges are predicted to become nonflexible and glassy, see the unfilled symbols for \( \phi \geq \phi^* \) (\( h_{\text{ent}} \leq b \)), which is likely responsible for the failure of our prediction for the two highest content samples with \( r = 7 \text{ nm} \), and the highest content sample with \( r = 11.5 \text{ nm} \) in Figure 2, considering that rubbery bridges is one vital assumption of our flexible percolation model. Not surprisingly, the data for \( \phi = 26\% \) and 34\% of the \( r = 7 \text{ nm} \) NPs in Figures 2(a1) and 2(a2) are significantly above the flexible bridge percolation model predictions, as these have some mixture of flexible and more rigid glassy bridges between NPs.

In summary, a parameter free analytical model is proposed to predict the LVE of hydrophilic NPs well-dispersed in a polar polymer matrix with strong interactions between the particles and the polymer, using critical percolation of NPs. This model is applicable for particles that are neither too crowded nor too sparse, so that the flexible bridges form between particles. The limit where particles form glassy bridges when they are too close is evident in our LVE data for small NPs at high loading, exactly as anticipated by our phase diagram (Figure 3) and perhaps described by other models,\(^{3,4} \) but that is left for future work.

### ASSOCIATED CONTENT

#### Supporting Information

Figure S1: determination of the gel point; Figure S2: comparison of the mean nearest-neighbor distance \( h_{\text{ent}} \) between particles and segment and chain sizes of the polymer matrix. This material is available free of charge via the Internet at http://pubs.acs.org.

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**Notes**

The authors declare no competing financial interest.

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### REFERENCES


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