In small-scale metallic systems, collective dislocation activity has been correlated with size effects in strength and with a step-like plastic response under uniaxial compression and tension. Yielding and plastic flow in these samples is often accompanied by the emergence of multiple dislocation avalanches. Dislocations might be active pre-yield, but their activity typically cannot be discerned because of the inherent instrumental noise in detecting equipment. We apply Alternate Current (AC) load perturbations via Dynamic Mechanical Analysis (DMA) during quasi-static uniaxial compression experiments on single crystalline Cu nano-pillars with diameters of 500 nm, and compute dynamic moduli at frequencies 0.1, 0.3, 1, and 10 Hz under progressively higher static loads until yielding. By tracking the collective aspects of the oscillatory stress-strain-time series in multiple samples, we observe an evolving dissipative component of the dislocation network response that signifies the transition from elastic behavior to dislocation avalanches in the globally pre-yield regime. We postulate that microplasticity, which is associated with the combination of dislocation avalanches and slow viscoplastic relaxations, is the cause of the dependency of dynamic modulus on the driving rate and the quasi-static stress. We construct a continuum mesoscopic dislocation dynamics model to compute the frequency response of stress over strain and obtain a consistent agreement with experimental observations. The results of our experiments and simulations present a pathway to discern and quantify correlated dislocation activity in the pre-yield regime of deforming crystals.

Using basic forms of mechanical loading, such as a force- or displacement-controlled compression, careful examination of the data provided evidence for the presence of short plastic instabilities before the onset of the obvious and apparent strain bursts [2, 3, 18–20] (e.g. 100 - 400 MPa regime of 500 nm pillars, as shown in Fig. 1a). In-situ transmission electron microscope (TEM) nanoindentation experiments revealed the onset of dislocation motion before the first obvious displacement excursion. Creep experiments on single crystals of ice detected acoustic emission events at resolved shear stresses far below the yield stress [23]. These observations have yet to be connected to constitutive relations and a quantifiable stress-strain response. Discrete Dislocation Dynamics (DDD) simulations suggest the existence of intermittent events in the pre-yield regime of crystalline materials [2, 3] and a significant loading rate effect on strain burst response of nano- and micro-crystals due to dislocation jamming and relaxation [20]. Stress-induced probabilistic cross-slip relaxation has also been associated with several non-trivial aspects of crystal plasticity [27]. It’s natural to question whether we can detect and quantify microplasticity in crystals’ pre-yield regime.

Machine noise has been the Achilles’ heel of numerous experimental nano-mechanical investigations. Attempts have been made to characterize the machine noise, with reported values of \(0.2\) nm displacement, \(\sim 30\) nN force- noise floor, and a thermal drift of \(< 0.05\) nm/s for the prevalently used Hysitron TI 950 Triboindenter in quasi-static mode [28]. In uniaxial compression experiments, a flat nanoindenter tip applies compressive load to the top of a commonly cylindrical sample, a so-called
micro- or nano-pillar, and the indenter-sample friction, as well as the electromagnetic assembly responsible for the load control produce substantial and inevitable machine noise. In addition, noise caused by thermal drift sets a limit on the duration of such experiments, which renders long-time mechanical experiments like cyclical or fatigue loading, as well as creep tests, virtually impossible to interpret. Statistical probing is necessary to detect any possible non-linear dislocation activities, which cause axial displacements below the machine noise. We apply Dynamic Mechanical Analysis (DMA) at multiple frequencies that span three orders of magnitude, from 0.1 to 10 Hz, on multiple 500 nm-diameter single crystalline Cu nano-pillars. We statistically characterize the overall DMA behavior and compare it with mean-field dislocation depinning predictions.

Experiment. Cylindrical nano-pillars with diameters of ∼ 500 nm and aspect ratios (height/diameter) of ∼ 3:1 were fabricated following a concentric-circles top-down methodology using a Focused Ion Beam from bulk single-crystalline copper (> 99.9999% purity) with one side polished to a < 30 Å RMS roughness, oriented in ∼ (111) direction. The nano-mechanical experiments were carried out in a nanoindenter equipped with an 8 μm-diameter flat punch diamond tip custom made specifically for these experiments. Fig. 1(a) conveys a representative compressive engineering stress-strain data, with the inset showing the corresponding time series of load and displacement, zoomed into the pre-yield regime. In the experiment, we applied uniaxial quasi-static load that monotonically increased in a step-wise fashion to an individual nano-pillar. Small stress oscillations with the amplitude of 6 μN and a fixed frequency in the range between 0.1 and 10 Hz were superimposed over the static load to each 15 s step interval. Before the initiation of each compression experiment we waited for > 145 s to equilibrate the in-contact displacement drift and used the last 20 s drift data to estimate the thermal drift rate for subsequent correction. Only those experiments where the thermal drift rate was less than 0.05 nm/s were analyzed. The loading time before the occurrence of the first large strain event was usually within the first 200 s for all tests. Fig. 1(b) shows the representative pre- and post-compression SEM images of a representative Cu nano-pillar.

The dynamic modulus is defined as the frequency response of stress over strain $E(\omega, \sigma_0) = \sigma(\omega)/\epsilon(\omega)$, where $\sigma_0$ is the applied quasi-static stress and $\omega = 2\pi f$ is the driving frequency. Using this definition, we can extract the dynamic modulus from the oscillations that are imposed at each quasi-static stress $\sigma_0$ using a frequency domain analysis. We fit the time series of stress $\sigma(t)$ and strain $\varepsilon(t)$ using the following form which also includes a linear drift term:

$$\sigma_f(t) = x_r \cos(\omega t) + x_i \sin(\omega t) + \sigma_d t + \sigma_0,$$

$$\varepsilon_f(t) = u_r \cos(\omega t) + u_i \sin(\omega t) + \varepsilon_d t + \varepsilon_0,$$

where $x_r, x_i, \sigma_0, \sigma_d, u_r, u_i, \varepsilon_d, \varepsilon_0$ are fitting parameters for the stress and strain. The complex dynamic modulus $E$ can then be calculated as a function of the $\omega$ and $\sigma_0$:

$$E(\omega, \sigma_0) = \frac{x_r - ix_i}{u_r - iu_i} = A(\omega, \sigma_0)e^{i\phi(\omega, \sigma_0)},$$

where $A$ and $\phi$ are the amplitude and phase components of the dynamic modulus.

We applied this type of DMA using different driving frequencies: 0.1, 0.3, 1, and 10 Hz, for each quasi-static load hold and took measurements from 6 samples for each frequency driving test. We solved for the dynamic modulus at each quasi-static loading step at the single driving frequency using the fitting procedure described above. The quasi-static stress at each step is normalized by the yield stress of the system $\sigma_{ys}$. We binned amplitude and phase lag for sample statistics. The binning mean and standard error for amplitude and phase lag were calculated as a function of the stress bin centers and are shown for each driving frequency in Fig. 2(a). This
DMA data reveals a maximum of ∼70% decrease in the average amplitude and a maximum of ∼60° increase in the average phase lag as the applied quasi-static stress approaches yielding at 1 (∼400 MPa). This plot also shows that these deviation from elastic behavior are more pronounced for slower driving frequencies. These results are in stark contrast to the DMA data collected from the same type of uniaxial compression on a ∼500nm-diameter fused silica nano-pillars, which exhibits a constant amplitude of ∼65 GPa and a no-delay response for the driving frequencies of 1 Hz and 10 Hz.

**Simulation.** To reveal the underlying mechanisms that drive the observed non-trivial loss behavior in Cu as the applied stress approaches yielding, we constructed a continuum crystal plasticity model that aims at capturing the salient aspects of the observed mechanical behavior. This model considers the energetics of two competing processes: the dislocation-driven abrupt strain jumps and the slow stress-controlled relaxations towards minimum system energy state. To capture both the fast avalanches and the slow viscoplastic relaxations, we utilized a cellular automaton constitutive microplasticity model enhanced with an additional continuous-in-time strain field that follows a viscoplastic constitutive law [27, 32]. We model the shear strain to consist of the elastic and plastic components $\gamma = \gamma_e + \gamma_p$. The elastic term is calculated using Hooke’s law. The plasticity model that captures the plastic term can be realized using detailed continuum plasticity modeling approaches [1]. It is reasonable to assume that in a single representative volume element for single-crystalline FCC crystals, the following criteria hold: i) uniaxial loading activates one dominant crystallographic slip system, A, with another system, B, assisting dislocation glide along A [40]; and ii) dislocations carry plastic distortion via two distinct mechanisms: (a) fast dislocation avalanche-like glide and (b) slow, stress-relaxation-driven secondary glide on A caused by the coupled A-B dislocation mechanisms (e.g., double cross-slip) [27]. With contributions from both mechanisms, the total plastic strain can be expressed as:

$$\gamma_p = \gamma_p^{(a)} + \gamma_p^{(b)}. \quad (3)$$

In the fast dislocation avalanche-driven mechanism, a volume element at location $\mathbf{r}$ yields a random plastic strain $\delta_p^{(a)}$ if the local stress $\tau(\mathbf{r})$ is larger than a local depinning threshold $\chi(\mathbf{r})$ [27, 33, 34], where $\chi(\mathbf{r})$ follows a uniform distribution [35]. After each avalanche, the threshold value is re-drawn from the same distribution. On the other hand, the slow relaxation mechanism follows a typical constitutive viscoplastic law:

$$\dot{\gamma}_p^{(b)} = \frac{D}{G} (\tau(\mathbf{r}))^n, \quad (4)$$

where $D$ is the relaxation constant, $G$ is the shear modulus, and $n \in [1, 3] < 10$ is the critical quantity to define another timescale which is slow compared to the fast avalanche process [19].

For numerical simplicity, we apply this methodology to edge dislocations only, for which the local resolved shear stress can be explicitly calculated:

$$\tau(\mathbf{r}) = \tau_{ext} + \tau_{int}(\mathbf{r}) + \tau_{hard}(\mathbf{r})$$

$$= \tau_0 + \tau_{A} \sin(\omega t) + \int d^2\mathbf{r}' K(\mathbf{r} - \mathbf{r}') \gamma_p(\mathbf{r}') - h\gamma_p(\mathbf{r}), \quad (5)$$

where $\tau_{ext}$ is the applied external quasi-static stress combined with the oscillation component, $\tau_{int}$ is the stress that accounts for the long-range interactions with other dislocations, and $\tau_{hard}$ is the stress that arises from dislocation hardening. In the expanded form of Eq. 5, $K$ serves as the interaction kernel for single slip straight
edge dislocations, and $h$ represents a mean-field phenomenological hardening parameter $[32]$. For the stress kernels of complete circular dislocation loops or screws, in principle the results would be unchanged, since all these kernels are sufficiently long-ranged $[33, 36]$. The model implementation is such that the system is meshed into $N \times N$ elements, with $N = 32$. We prescribe similar loading conditions to 8 random initial configurations as we did in the experiments, with different driving frequencies of 1, 2, 8, and 64 rad/s. The rate equation associated to Eq. (3) can be numerically solved by Euler integration with a fixed time step $\Delta t = 10^{-2}$ s. Additional simulation details are provided in Supplemental Materials $[35]$.

**Discussion.**—Fig. 2(b) shows the same frequency domain analysis of the dynamic modulus using simulations results as the ones shown in Fig. 2(a) for the experimental data. This qualitative agreement between simulations and experiments motivates a further quantitative comparison. Existing simulations investigated the effect of cyclic loading on the evolved dislocation network and predicted a scaling relation between the normalized strain rate amplitude and the driving frequency, focused on the mean-field depinning theory framework $[24, 37]$, 

$$\frac{|\dot{\epsilon}|}{|\sigma|} \sim \omega^\kappa \quad (6)$$

where $\kappa = 1$ corresponds to a simple harmonic oscillator, i.e. perfectly elastic behavior, and $\kappa = 0.82$ corresponds to a system driven close to the pinning threshold $[24, 37]$. The strain rate amplitude is normalized by the stress amplitude, $|\dot{\epsilon}|/|\sigma|$, which is equivalent to $\omega/A$, where $A$ is the dynamic modulus amplitude measured via DMA. Fig. 3 shows the scaling analysis of the normalized strain rate amplitude vs. driving frequency for the dynamic modulus amplitude calculated from the experiments and simulations at different quasi-static loads. The insets show the scaling parameter $\kappa$ as a function of the normalized stress. These plots convey that at both small and large stress regimes, experiments and simulations produce scaling behaviors that are in agreement with the mean-field depinning predictions, and a smooth, microplastic crossover connects these two extreme regimes. The experiments and simulations reveal enhanced microplastic activities as the system is stressed close to yielding. The actual mechanism that is responsible for the increased ‘susceptibility’ to plasticity can be a thermally activation process like cross-slip, or the collective dislocation bowing out due to long-range interactions, i.e. the Andrade mechanism $[38]$.

**Summary.**—We imposed oscillatory loads in the nominal elastic regime of the uniaxially compressed 500 nm-diameter single crystalline Cu nano-pillars. We applied monotonically increasing stresses above the bulk yield point of 70 MPa $[39]$ to investigate the mechanically correlated material response. Analysis of the cumulative oscillatory response reveals a substantial deviation from the nominally perfectly elastic behavior, as well as an emergent dissipation signature in what has always been considered pre-yield regime. Our experimental observations are corroborated by a mesoscale dislocation plasticity model, which accounts for dislocation avalanches (fast processes) and the viscoplastic response (slow time scales) during oscillatory loading. We formulate a scaling analysis that shows a smooth transition of the system from perfect elasticity to dislocation depinning-driven plasticity that occurs at loads lower than the global yield stress. This approach represents a new pathway to investigate and quantify the abrupt plastic events that emanate from dislocation activities even in the pre-yield regime, that occur ubiquitously during deformation of small-scale single crystals below instrumental noise levels.

The developed methodology can be applied to characterize pre-yield dislocation dynamics in extensive list of FCC, BCC, and HCP materials. The micromechanical study sheds light on detecting cracking noise in macroscopic sample subjected to nominal elastic loading. The observation of such events might lead to better prediction of plastic yielding and even incipient fracture for structural materials. The pre-yield mechanical noise itself can be a hidden problem for instrumentation that requires high strain sensitivity, e.g. advanced LIGO $[15, 16]$.

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* To whom correspondence should be addressed: xnni@caltech.edu

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