Model evaluation efforts in the environmental sciences are typically organized around a concept of uncertainty – we try to understand how much error or uncertainty should be associated with model predictions. This approach is incoherent in the sense that any scientific method based on a concept of uncertainty admits at least two distinct logical contradictions (Nearing et al. 2016). I propose that we should instead be asking questions about information, and that by changing our perspective in this particular manner alleviates, or at least attenuates, several longstanding problems that environmental modelers have typically treated as intractable. In particular, I propose that an information-centric perspective to model evaluation leads to: (i) purely logical hypothesis tests for complex system models, which addresses the central problem discussed by Neyman (1957), (ii) direct strategies for process-level diagnostics, which addresses the Duhem-Quine problem, and (iii) asymptotically correct estimates of epistemic uncertainty, which addresses the central problem discussed by Knight (1921). The intuition is that we know from Knuth’s (2005) version of Shannon’s (1948) second theorem that there is a unique quantitative doxastic model of performing a science experiment, and the problem is that this model is missing from all major existing philosophies of science. Adding this model of an experiment to either standard falsificationism or a modern version of inductivism is sufficient to facilitate the resolutions mentioned above. I obviously do not address Hume’s problem, but do apparently reduce its effects to a divergence that reliably approaches epsilon over very large classes of models. Additionally, this philosophy suggests that we should be building models around maximum entropy distributions instead of as solutions to conservation equations. It turns out that if we do this, we get for free the ability for our models to learn directly from data that was posed as a “Grand Challenge” by Beck (2009). And anyway, given recent results like Best (2015), now may be a good time to rethink our general strategy for building models. - References can be emailed upon request.