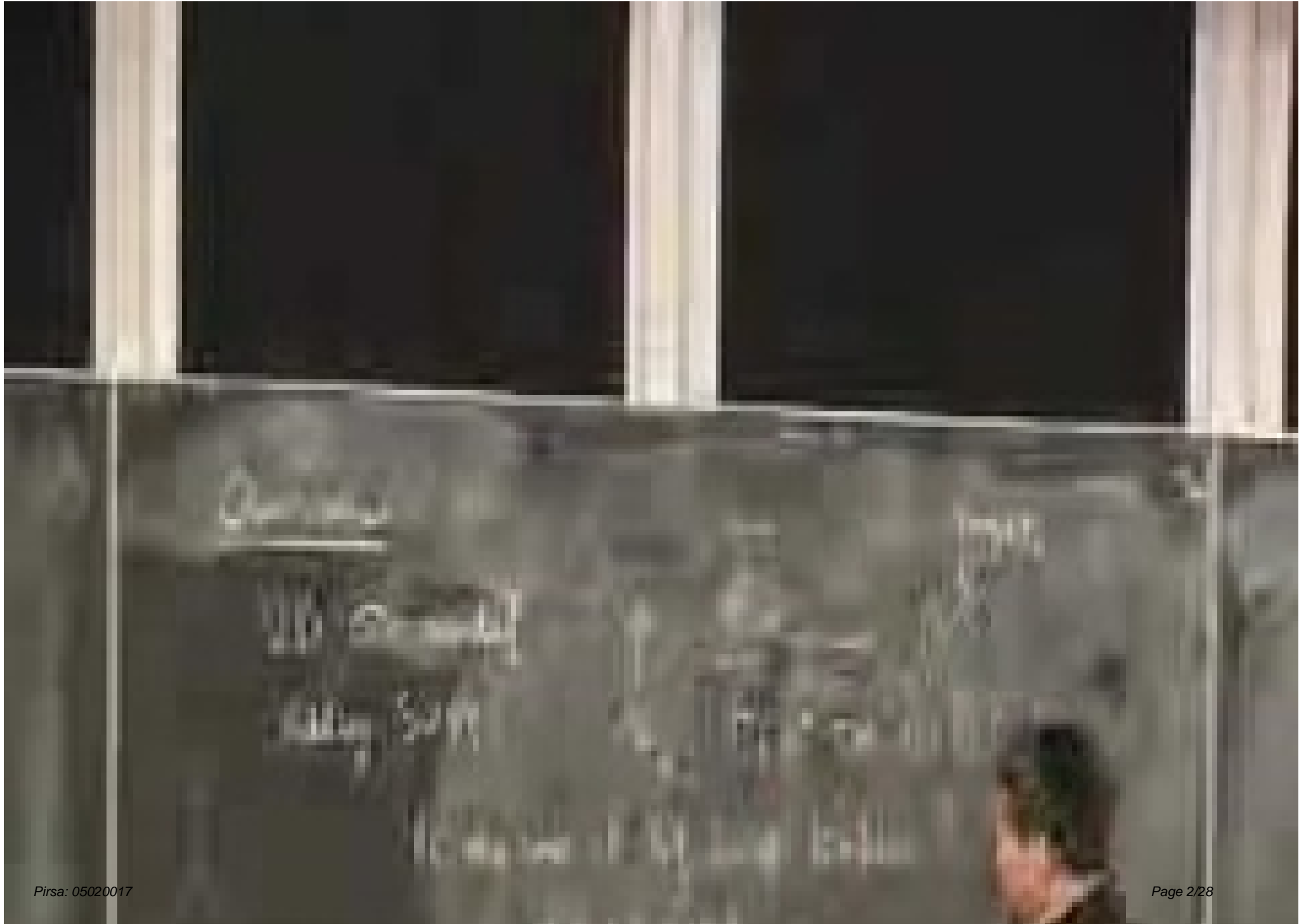


Title: Twisted Generalized Calabi-Yau Manifolds and Topological Sigma Models with Flux (Part 1)

Date: Feb 15, 2005 02:00 PM

URL: <http://pirsa.org/05020017>

Abstract: In these lectures, we examine how twisted generalized Calabi-Yau (GCY) manifolds arise in the construction of a general class of topological sigma models with non-trivial three-form flux. The topological sigma model defined on a twisted GCY can be regarded as a simultaneous generalization of the more familiar A-model and B-model. Emphasis will be given to the relation between topological observables of the sigma model and a Lie algebroid cohomology intrinsically associated with the twisted GCY. If time permits, we shall also discuss topological D-branes in this more general setting, and explain how the viewpoint from the Lie algebroid helps to elucidate certain subtleties even for the conventional A-branes and B-branes. The lectures will be physically motivated, although I will try to make the presentation self-contained for both mathematicians and physicists.



1. (a) $\frac{1}{2} \frac{d}{dt} \left(\frac{1}{2} m v^2 \right) = \frac{1}{2} m v \frac{dv}{dt}$



2. (a) $\frac{1}{2} \frac{d}{dt} \left(\frac{1}{2} m v^2 \right) = \frac{1}{2} m v \frac{dv}{dt}$

[11] $\mathbb{C}^2 \rightarrow \mathbb{C}^2$ on \mathbb{R}^2



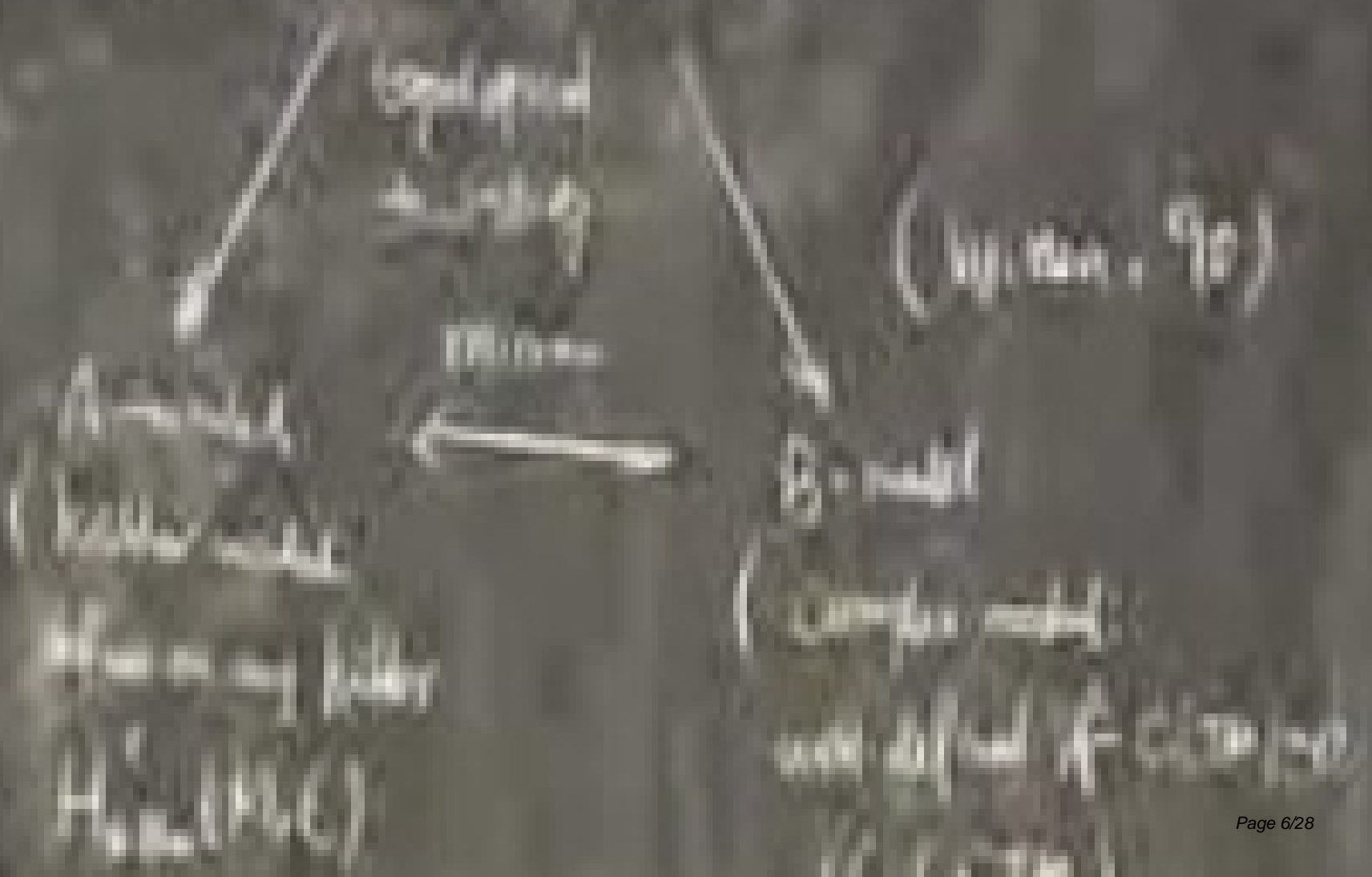
H-Modul

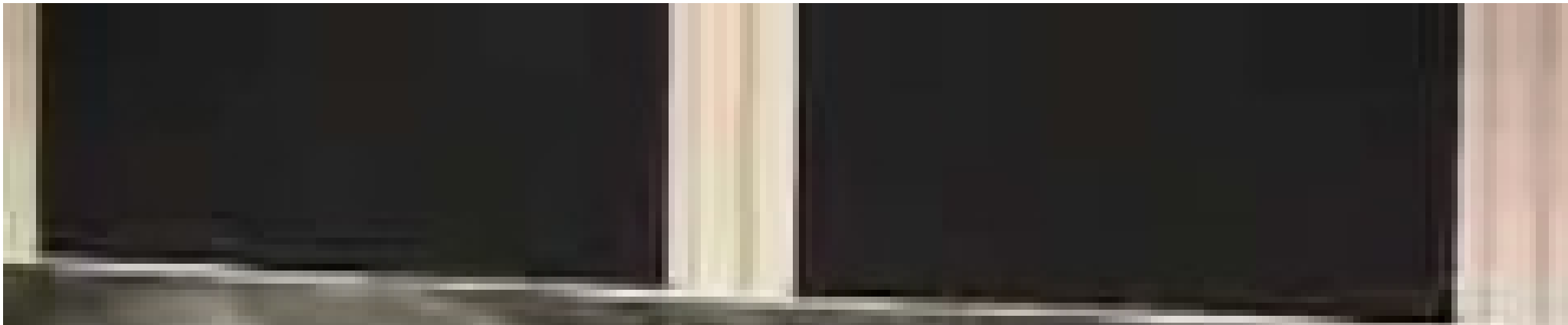
(Complex modul)

well-defined if \mathbb{R}

\mathbb{R}^n (A, T, N)

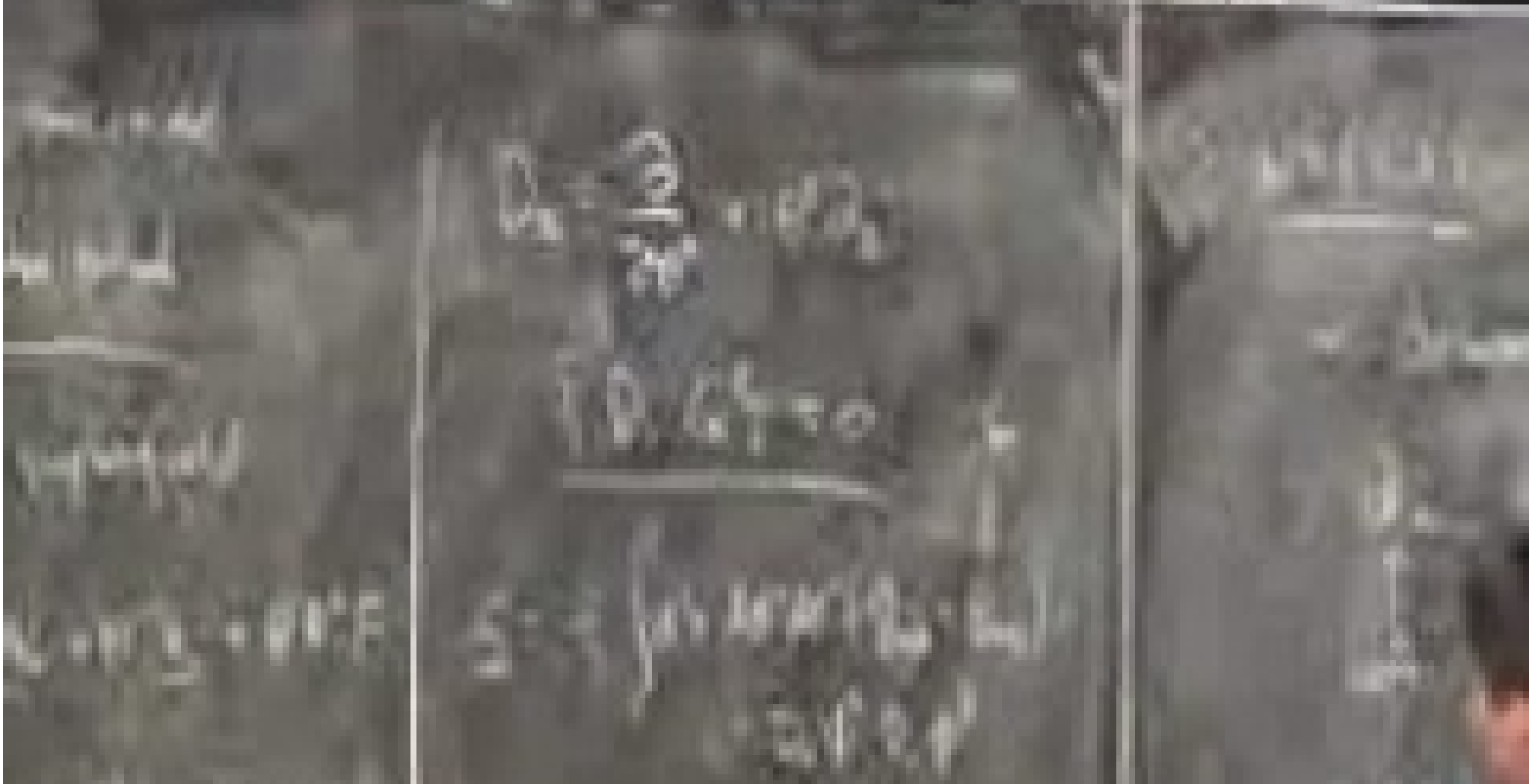
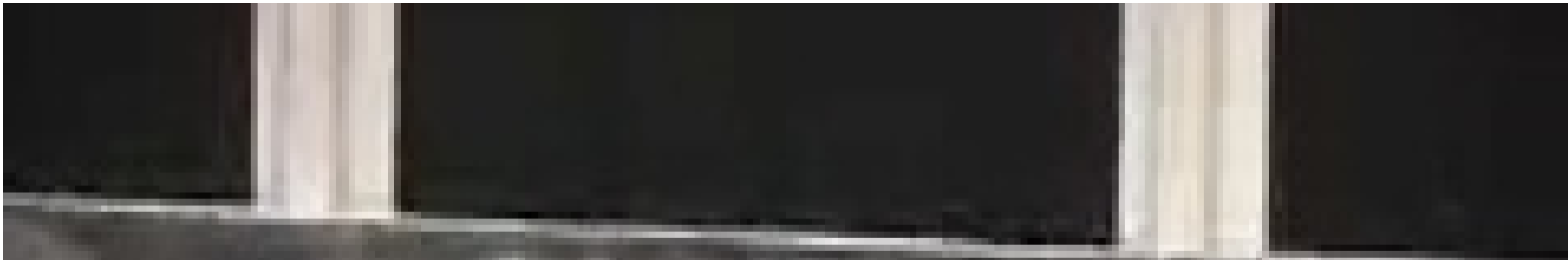
(1.1) Consider $\alpha \in M$





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$$\nabla \cdot \mathbf{G} = 0$$

$$S = \frac{1}{2} \int d^4x \sqrt{-g} (g_{\mu\nu} \dot{\phi}^\mu \dot{\phi}^\nu - D_\mu \phi^\mu D_\nu \phi^\nu)$$

\mathcal{H} field is not well defined

$$\begin{aligned} \mathbb{E}[\hat{\beta}_1] &= \beta_1 + \frac{\sigma^2}{\sum_{i=1}^n x_i^2} \\ \mathbb{E}[\hat{\beta}_0] &= \beta_0 + \frac{\sigma^2}{\sum_{i=1}^n x_i^2} \end{aligned}$$

• Bias = $(\text{true}) - \text{estimate}$

$$\begin{aligned} \text{Bias}(\hat{\beta}_1) &= 0 \\ \text{Bias}(\hat{\beta}_0) &= -\frac{\sigma^2}{\sum_{i=1}^n x_i^2} \end{aligned}$$

$\Rightarrow \hat{\beta}_1$ is unbiased, $\hat{\beta}_0$ is biased



