Statistical Methods in Particle Physics

Quiz on chapter 5: Parameter Estimation

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2. its expectation value differs from the true value
3. it is a maximum likelihood estimator
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