

Statistical Methods in Particle Physics

Quiz on chapter 5: Parameter Estimation

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**Heidelberg University
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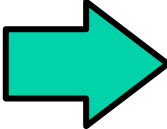
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3. it is a maximum likelihood estimator
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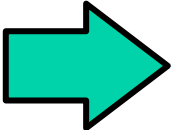
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2. n^2
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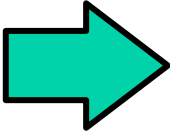
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The variance of a chi-squared estimator for one parameter is related to

1. the first derivative of the chi-squared function
2. the second derivative of the chi-squared function
3. the logarithm of the chi-squared function
4. the integral of the chi-squared from the measured value to infinity

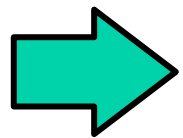
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$$V[\hat{\theta}] = - \left. \frac{\partial \ln L}{\partial \theta} \right|_{\theta=\hat{\theta}}$$

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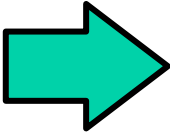
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