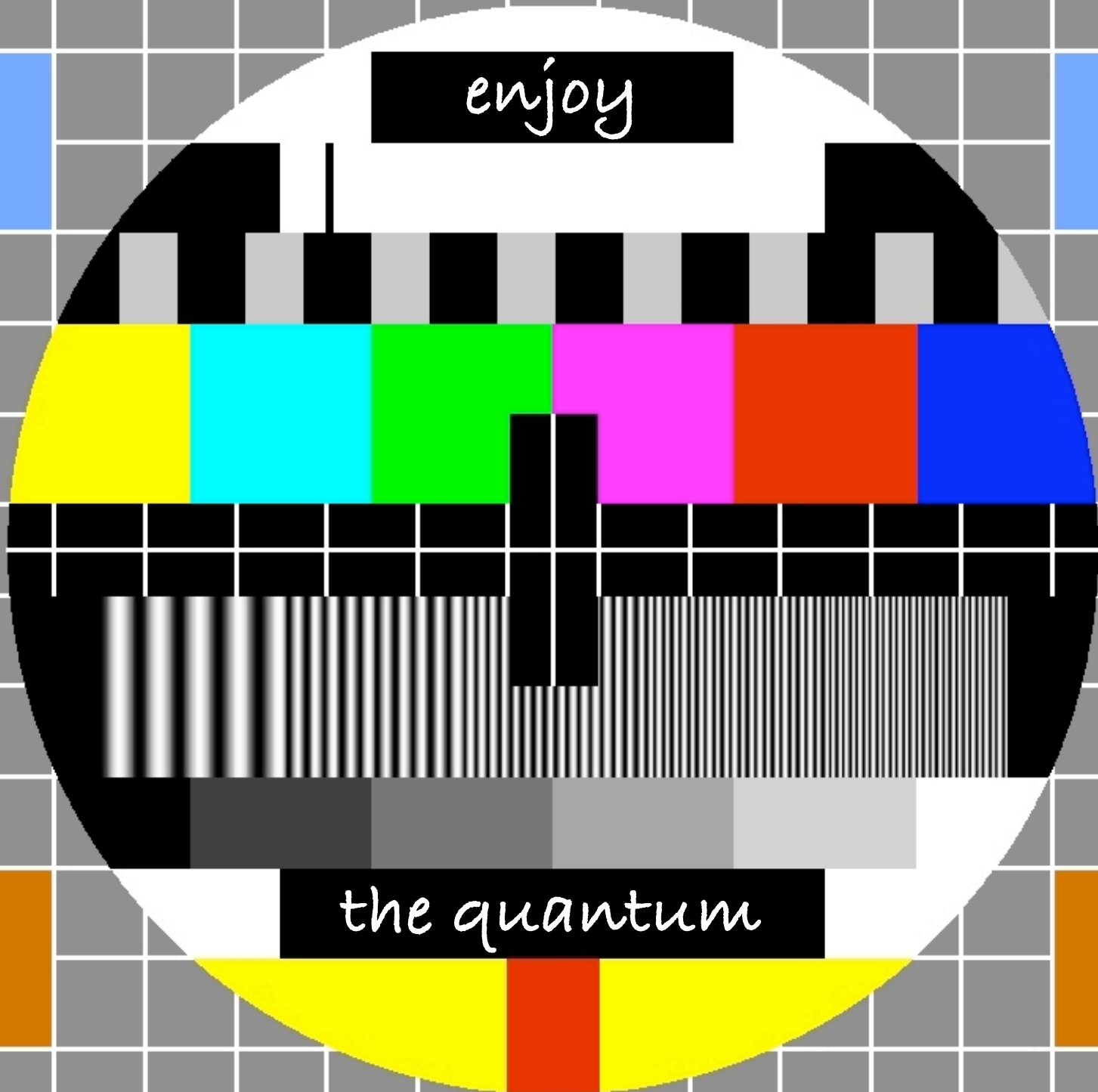


enjoy

the quantum



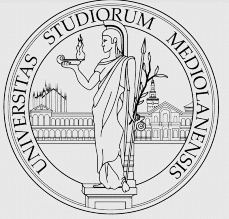
Quantum estimation for quantum technology

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Quantum Future Technologies Conference 1.0
NASA Ames research center, 17-21 January 2012



Quantum technology ("Schroedinger's machines")

- Quantum information
(communication and computing)
- Quantum metrology
(calibration, interferometry, nanopositioning)
- Quantum imaging
(ghost imaging and diffraction, quantum lithography)

Quantum characterization for quantum technology



- It is highly desirable to have theoretical and experimental tools for the precise characterization of signal and devices *at the quantum level*



Quantum estimation

- The "resources" involved in quantum-enhanced metrology/technology are entanglement, nonlocality, entropy, interferometric phase-shift, etc..

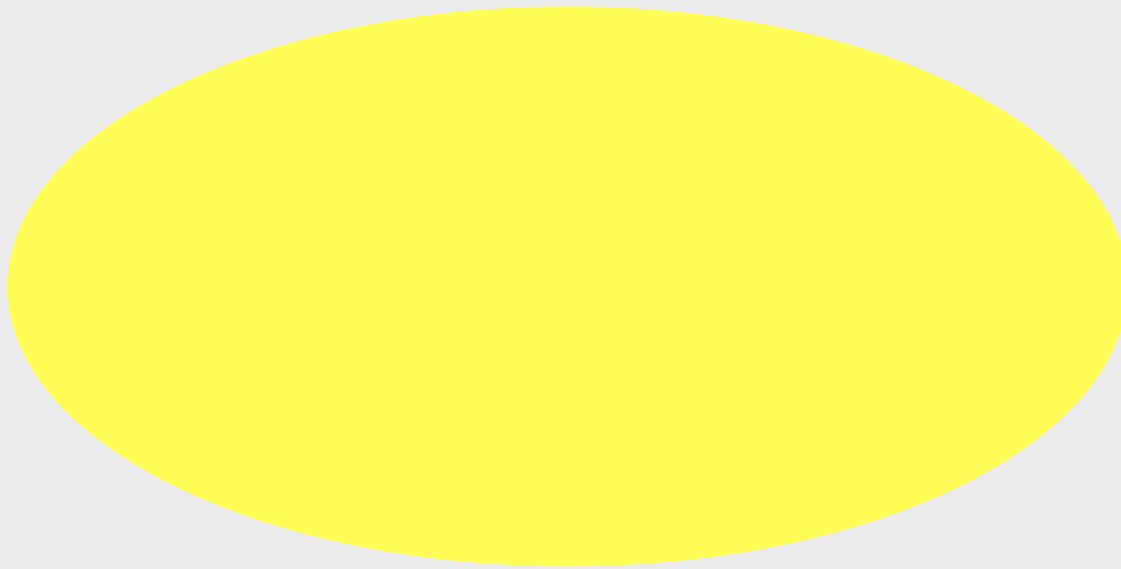
In general they are not observable quantities in strict sense (do not correspond to a selfadjoint operator)

- No correspondence principle
- No uncertainty relations

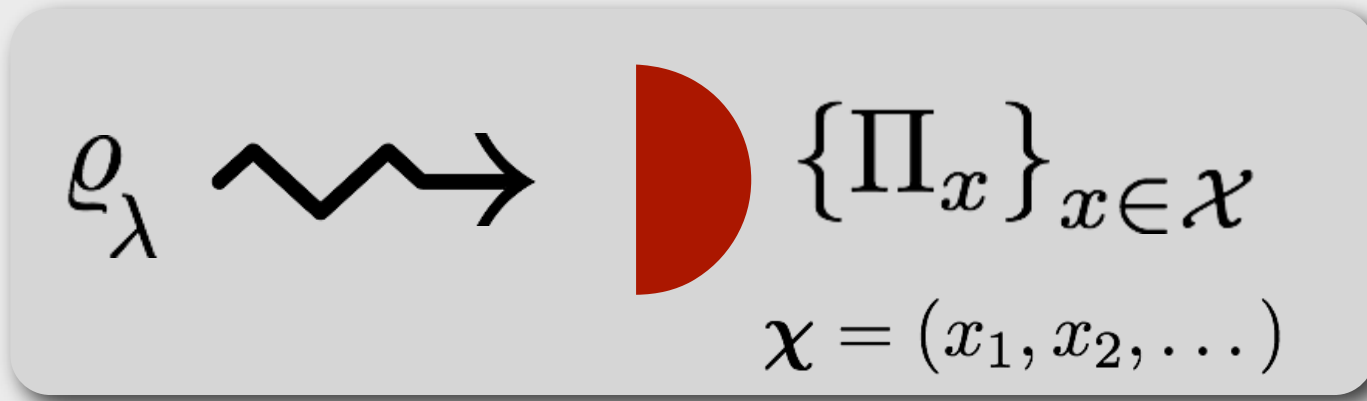


Quantum estimation

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■ Quantum estimation

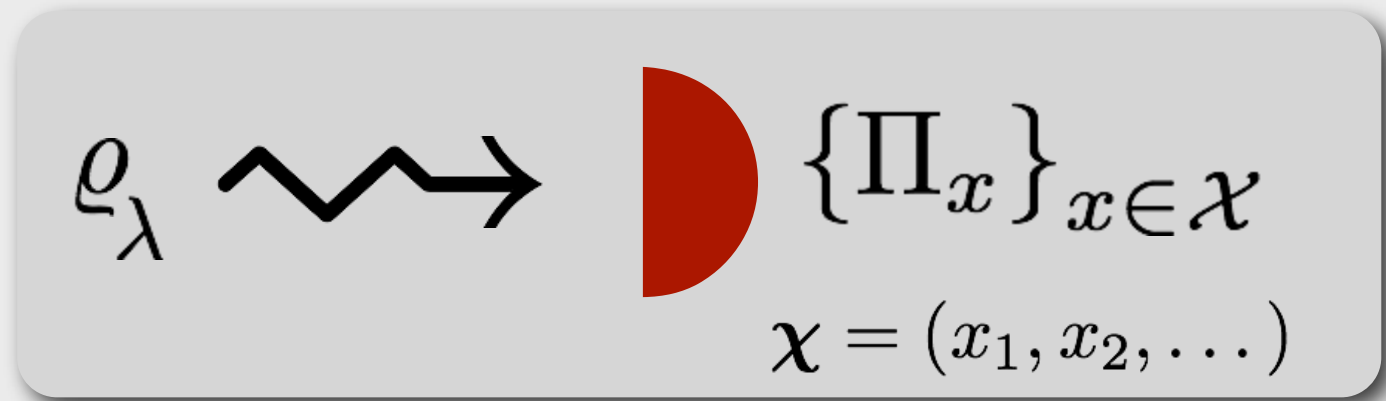


- Optimal measurement
- Ultimate bound to precision

■ Quantum estimation



$$\left(\varrho_0 \quad \Gamma_\lambda \right)$$



- Optimal measurement
- ultimate bound to precision

■ Measurement and estimation

- ~~direct measurements~~
- indirect measurements

 influence on a different quantity



- choice of the measurement
- choice of the estimator



■ Cramer - Rao bound (unbiased estimators)

■ variance of unbiased estimators

$$\text{Var}_\lambda[\hat{\lambda}] \geq \frac{1}{MF(\lambda)}$$

■ M -> number of measurements

■ F -> Fisher Information

$$F(\lambda) = \int dx p(x|\lambda) \left[\partial_\lambda \log p(x|\lambda) \right]^2$$

■ Optimal measurement -> maximum Fisher

■ Optimal estimator -> saturation of CR inequality



■ (Asymptotically) optimal estimators

- Bayes estimator from a posteriori distribution

$$p(x|\lambda)p(\lambda) = p(\lambda|x)p(x)$$

Laplace von Mises Th. $p(\lambda|\{x\}) \xrightarrow{M \gg 1} G(\lambda^*, \sigma^2)$

- MaxLik estimator(s) from the measurement likelihood

$$\mathcal{L}(x_1, x_2, \dots, x_M | \lambda) = \prod_{k=1}^M p(x_k | \lambda)$$

■ Let's go quantum (1)

$$\left(\varrho_0 \quad \Gamma_\lambda \right) \varrho_\lambda \rightsquigarrow \left\{ \Pi_x \right\}_{x \in \mathcal{X}}$$

$$\mathcal{X} = (x_1, x_2, \dots)$$

■ probability density $p(x|\lambda) = \text{Tr} [\varrho_\lambda \Pi_x]$

■ symm. log. derivative (SLD) $\frac{L_\lambda \varrho_\lambda + \varrho_\lambda L_\lambda}{2} = \frac{\partial \varrho_\lambda}{\partial \lambda}$

selfadjoint, zero mean $\text{Tr} [\varrho_\lambda L_\lambda] = 0$

■ Fisher information $F(\lambda) = \int dx \frac{\text{Re} (\text{Tr} [\varrho_\lambda \Pi_x L_\lambda])^2}{\text{Tr} [\varrho_\lambda \Pi_x]}$



Let's go quantum (2)

$$\begin{aligned} F(\lambda) &\leq \int dx \left| \frac{\text{Tr}[\varrho_\lambda \Pi_x L_\lambda]}{\sqrt{\text{Tr}[\varrho_\lambda \Pi_x]}} \right|^2 \\ &= \int dx \left| \text{Tr} \left[\frac{\sqrt{\varrho_\lambda} \sqrt{\Pi_x}}{\sqrt{\text{Tr}[\varrho_\lambda \Pi_x]}} \sqrt{\Pi_x} L_\lambda \sqrt{\varrho_\lambda} \right] \right|^2 \\ &\leq \int dx \text{Tr}[\Pi_x L_\lambda \varrho_\lambda L_\lambda] \\ &= \text{Tr}[L_\lambda \varrho_\lambda L_\lambda] = \text{Tr}[\varrho_\lambda L_\lambda^2] \end{aligned}$$

(Braunstein and Caves 1994)

● Fisher vs Quantum Fisher

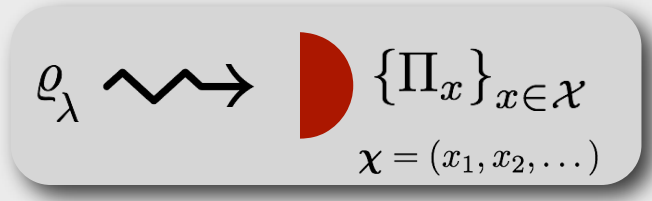
$$F(\lambda) \leq H(\lambda) \equiv \text{Tr}[\varrho_\lambda L_\lambda^2] = \text{Tr}[\partial_\lambda \varrho_\lambda L_\lambda]$$

● Quantum Cramer-Rao bound $\text{Var}(\lambda) \geq \frac{1}{MH(\lambda)}$



■ Optimal quantum measurement (1)

● the optimal measurement is a projective one, the spectral measure is built with the eigenstates of the SLD



optimal quantum measurement:
SLD + classical postprocessing (Bayesian, ML)



General formulas (basis independent)

$$\frac{L_\lambda \rho_\lambda + \rho_\lambda L_\lambda}{2} = \frac{\partial \rho_\lambda}{\partial \lambda} \quad \text{Lyapunov equation}$$

- Symmetric logarithmic derivative

$$L_\lambda = 2 \int_0^\infty dt \exp\{-\rho_\lambda t\} \partial_\lambda \rho_\lambda \exp\{-\rho_\lambda t\}$$

- Quantum Fisher Information

$$H(\lambda) = 2 \int_0^\infty dt \text{Tr} [\partial_\lambda \rho_\lambda \exp\{-\rho_\lambda t\} \partial_\lambda \rho_\lambda \exp\{-\rho_\lambda t\}]$$



General formulas

- Family of quantum states

$$\rho_\lambda = \sum_n \rho_n |\psi_n\rangle \langle \psi_n|$$

- Symmetric logarithmic derivative

$$L_\lambda = \sum_p \frac{\partial_\lambda \rho_p}{\rho_p} |\psi_p\rangle \langle \psi_p| + 2 \sum_{n \neq m} \frac{\rho_n - \rho_m}{\rho_n + \rho_m} \langle \psi_m | \partial_\lambda \psi_n \rangle |\psi_m\rangle \langle \psi_n|$$

- Quantum Fisher Information

$$H(\lambda) = \sum_p \frac{(\partial_\lambda \rho_p)^2}{\rho_p} + 2 \sum_{n \neq m} \frac{(\rho_n - \rho_m)^2}{\rho_n + \rho_m} |\langle \psi_m | \partial_\lambda \psi_n \rangle|^2$$



■ Optimal quantum measurement (2)

ultimate bound on precision $\text{Var}(\lambda) \geq \frac{1}{MH(\lambda)}$

(local quantum estimation theory)

- feedback assisted/adaptive measurements
- one-step adaptive procedure: rough estimate of the parameter on a small fraction of copies + measurement of SLD on the rest of the copies



■ unitary families of quantum states

$$\rho_\lambda = U_\lambda \rho_0 U_\lambda^\dagger \quad \rho_0 = \sum \rho_n |\varphi_n\rangle \langle \varphi_n|$$

$$U_\lambda = \exp\{-i\lambda G\} \quad \partial_\lambda \rho_\lambda = U_\lambda [G, \rho_0] U_\lambda^\dagger$$

● covariance of SLD $L_\lambda = U_\lambda L_0 U_\lambda^\dagger$

$$L_0 = 2 \sum_{n,m} \frac{\langle \varphi_m | [G, \rho_0] | \varphi_n \rangle}{\rho_n + \rho_m} |\varphi_n\rangle \langle \varphi_m|$$

● QFI is independent on the value of the parameter

$$H = \text{Tr} [\rho_0 L_0^2] = 2 \sum_{n \neq m} \frac{(\rho_n - \rho_m)^2}{\rho_n + \rho_m} \langle \varphi_m | G | \varphi_n \rangle^2$$



■ parameter-based uncertainty relations

● pure states $H = 4\langle\psi_0|\Delta G^2|\psi_0\rangle$

$$\text{Var}(\lambda)\langle\Delta G^2\rangle \geq \frac{1}{4M}$$



parameter-based uncertainty relations

- pure states $H = 4\langle\psi_0|\Delta G^2|\psi_0\rangle$

$$\text{Var}(\lambda)\langle\Delta G^2\rangle \geq \frac{1}{4M}$$

- mixed states

$$H = 4\text{Tr}[\Delta G^2 \varrho_0] + 4 \sum_n \varrho_n \langle\varphi_n|\langle G\rangle^2 - 2GK^{(n)}G|\varphi_n\rangle$$

$$K^{(n)} = \sum_m \frac{\varrho_m}{\varrho_n + \varrho_m} |\varphi_m\rangle\langle\varphi_m| \xrightarrow{\varrho_0 \rightarrow |\varphi_0\rangle\langle\varphi_0|} \frac{1}{2} |\varphi_0\rangle\langle\varphi_0|$$

$$\text{Var}(\lambda)\langle\Delta G^2\rangle \geq \frac{1}{4M} \left[1 + \sum_n \varrho_n \langle\varphi_n|\langle G\rangle^2 - 2GK^{(n)}G|\varphi_n\rangle \right]^{-1}$$



estimability of a parameter

- signal-to-noise ratio (single measurement)

$$R_\lambda = \frac{\lambda^2}{\text{Var}(\lambda)} \leq Q_\lambda \equiv \lambda^2 H(\lambda)$$

- relative error for a 3σ confidence interval (after M measurements)

$$\delta^2 = \frac{9\text{Var}(\lambda)}{M\lambda^2} = \frac{9}{M} \frac{1}{Q_\lambda} = \frac{9}{M\lambda^2 H(\lambda)}$$

- # of meas to achieve a given relative error

$$M_\delta = \frac{9}{\delta^2} \frac{1}{Q_\lambda}$$



■ estimability of a parameter: the unitary case

$$\rho_\lambda = U_\lambda \rho_0 U_\lambda^\dagger$$

- QFI is independent on the value of the parameter
- (Any) estimation procedure cannot be efficient for small value of the parameter

$$Q_\lambda \propto \lambda^2 \qquad M_\delta \propto \frac{1}{\delta^2 \lambda^2}$$



■ A nonunitary example: estimation of loss



■ Master equation $\frac{d\rho_\phi}{d\phi} = \tan \phi L[a]\rho_\phi \quad \exp\{-\gamma t\} = \cos^2 \phi$

$$L[a]\rho = 2a^\dagger \rho a - a^\dagger a \rho - \rho a^\dagger a$$

- absorption
- propagation in a noisy channel ($T=0$)



■ A nonunitary example: estimation of loss



■ Master equation $\frac{d\rho_\phi}{d\phi} = \tan \phi L[a]\rho_\phi \exp\{-\gamma t\} = \cos^2 \phi$

$$L[a]\rho = 2a^\dagger \rho a - a^\dagger a \rho - \rho a^\dagger a$$

- absorption
- propagation in a noisy channel ($T=0$)
- optimal measurement: Gaussian operations + photon count.
- ultimate precision $\text{Var}_\gamma[\hat{\gamma}] \rightarrow \frac{\gamma}{\bar{n}Mt} + O(\gamma^2)$

proportional to the loss parameter itself!

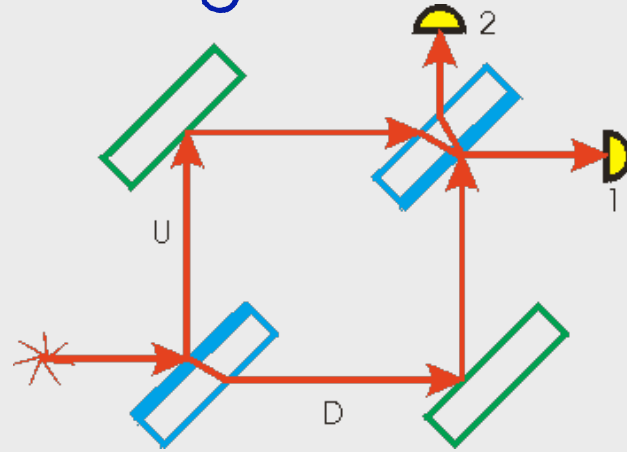
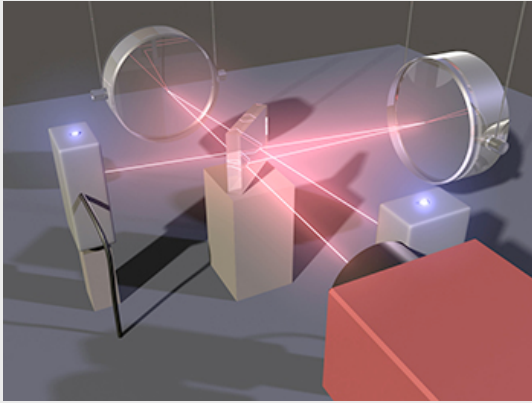
PRL 98, 160401 (2007) PHYSICAL REVIEW LETTERS week ending 20 APRIL 2007

Optimal Quantum Estimation of Loss in Bosonic Channels

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 (Received 7 February 2007; published 17 April 2007)

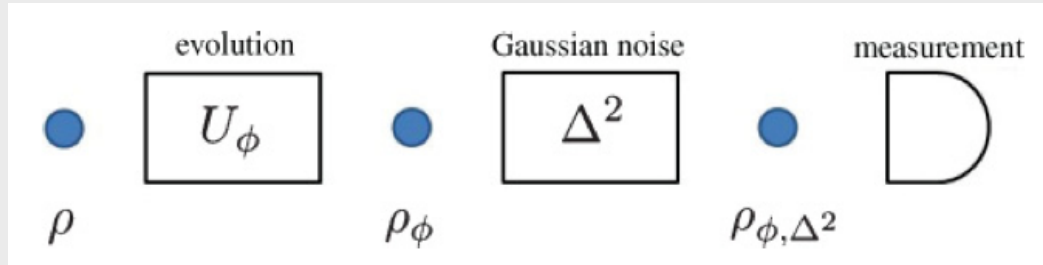
Quantum Interferometry



use of quantum states of light to improve sensitivity

- Optimization over input states (Caves 1981)
- Effects of detection noise
- Effects of losses
- Multiple interference
- Fixed number of particles, atomic interf

Estimation of phase in the presence of phase diffusion



$$U_\phi = \exp\{-i a^\dagger a \phi\}$$

$$\varrho_\phi = \mathcal{N}_\Delta[U_\phi \varrho U_\phi^\dagger] = U_\phi \mathcal{N}_\Delta[\varrho] U_\phi^\dagger$$

$$\mathcal{N}_\Delta[\varrho] = \sum_{nm} e^{-\Delta^2(n-m)^2} \varrho_{nm} |n\rangle\langle m|$$

In the noiseless case the optimal probe is the squeezed vacuum and $H = 8(N^2 + N)$ (Mouras 2006)

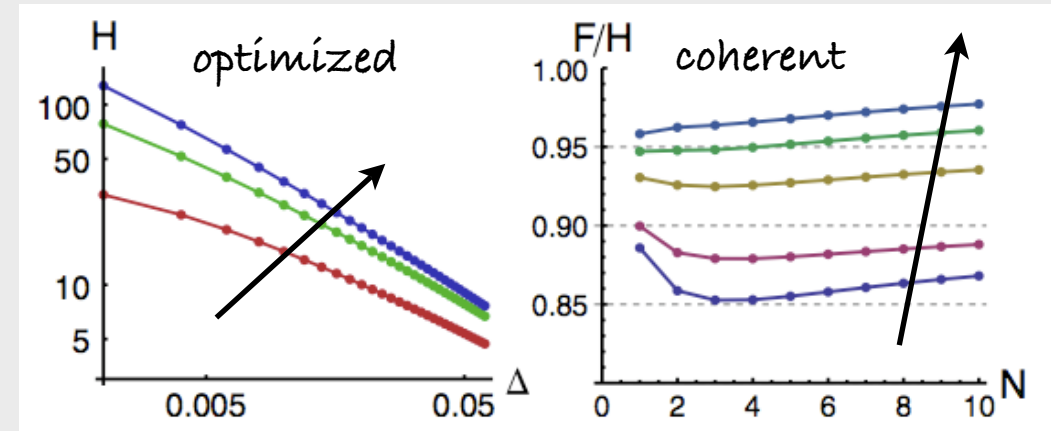


Estimation of phase in the presence of phase diffusion

In the presence of noise we have (approximate) scaling laws

$$H(N, \Delta) \simeq k^2 H(N/k, k\Delta) \quad \beta_{\text{opt}}(N, \Delta) \simeq \beta_{\text{opt}}(N/k, k\Delta)$$

Homodyning is nearly optimal for low and high noise





Estimation of phase in the presence of phase diffusion

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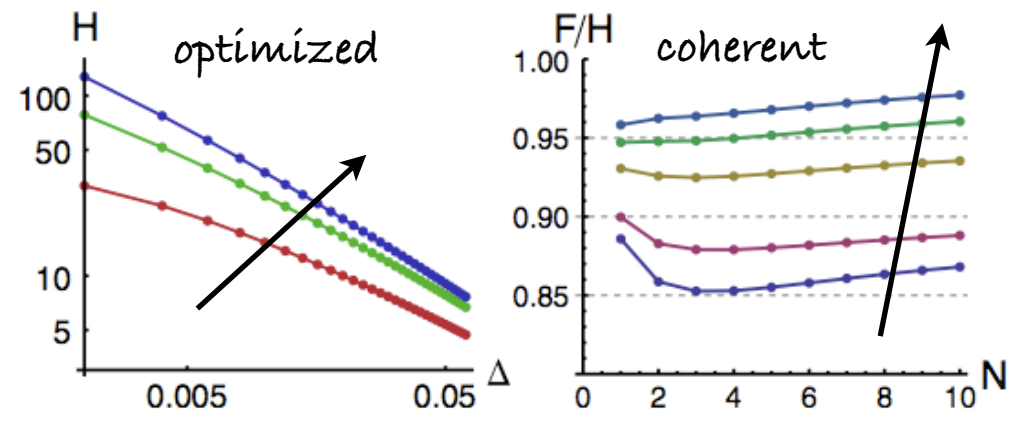
Homodyning is nearly optimal for low and high noise

PRL 106, 153603 (2011) PHYSICAL REVIEW LETTERS week ending 15 APRIL 2011

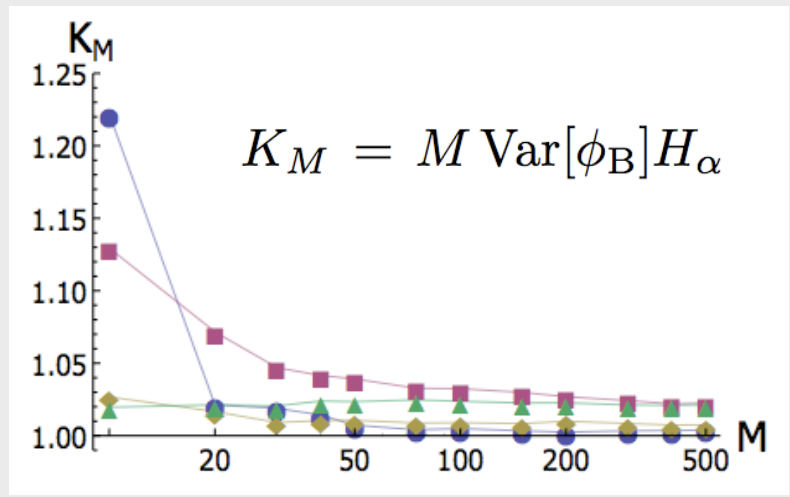
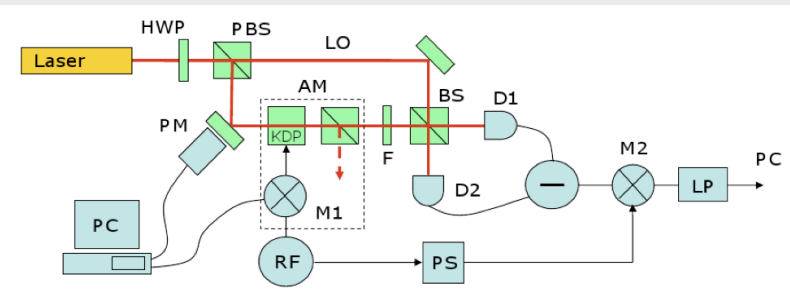
Optical Phase Estimation in the Presence of Phase Diffusion

Marco G. Genoni,^{1,2} Stefano Olivares,³ and Matteo G. A. Paris^{2,*}

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(Received 12 December 2010; published 14 April 2011)



Experiments with coherent states (large noise)



Bayes estim. is optimal for small samples



■ Estimation of entanglement

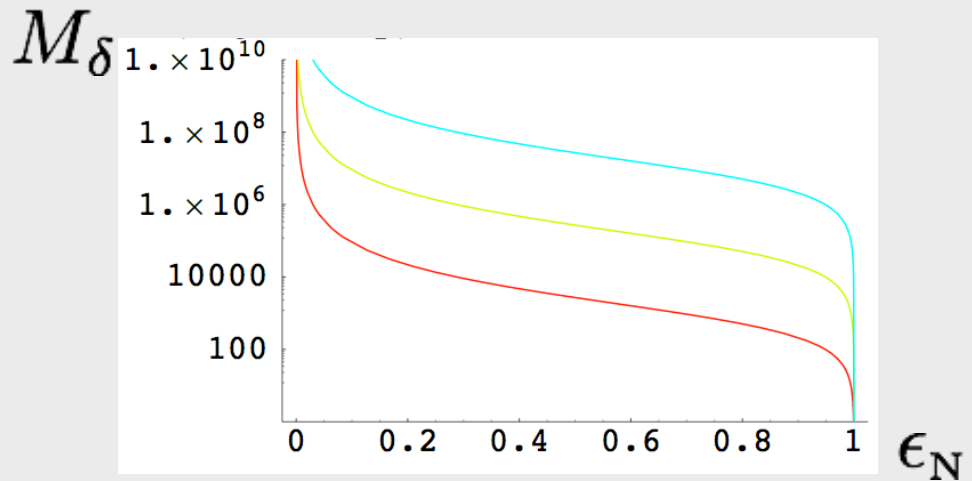
- pure states (Schmidt decomposition)

$$|\Psi_q\rangle = \sqrt{q}|0\rangle_A|0\rangle_B + \sqrt{1-q}|1\rangle_A|1\rangle_B$$

- entanglement measure: function of q (negativity, linear and VN entropy)

- QSNR is vanishing for vanishing entanglement

$$Q(\epsilon_N) = \frac{\epsilon_N^2}{1 - \epsilon_N^2} \stackrel{\epsilon_N \rightarrow 0}{\sim} \epsilon_N^2$$



■ The multiparametric case

● QFI matrix $\mathbf{H}(\boldsymbol{\lambda})_{\mu\nu} = \text{Tr} \left[\rho_{\boldsymbol{\lambda}} \frac{L_{\mu}L_{\nu} + L_{\nu}L_{\mu}}{2} \right]$

● bound on covariance $\text{Cov}[\boldsymbol{\gamma}]_{ij} = \langle \lambda_i \lambda_j \rangle - \langle \lambda_i \rangle \langle \lambda_j \rangle$
(not achievable)

$$\text{Cov}[\boldsymbol{\gamma}] \geq \frac{1}{M} \mathbf{H}(\boldsymbol{\lambda})^{-1}$$

● single parameter (achievable) $\text{Var}(\lambda_{\mu}) \geq \frac{1}{M} (\mathbf{H}^{-1})_{\mu\mu}$

reparametrization $\tilde{\boldsymbol{\lambda}} = \{\tilde{\lambda}_j = \tilde{\lambda}_j(\boldsymbol{\lambda})\}$ $\tilde{\lambda}_1 \equiv g(\boldsymbol{\lambda})$

$$\tilde{L}_{\mu} = \sum_{\nu} B_{\mu\nu} L_{\nu} \quad \tilde{\mathbf{H}} = \mathbf{B} \mathbf{H} \mathbf{B}^T \quad B_{\mu\nu} = \partial \lambda_{\nu} / \partial \tilde{\lambda}_{\mu}$$

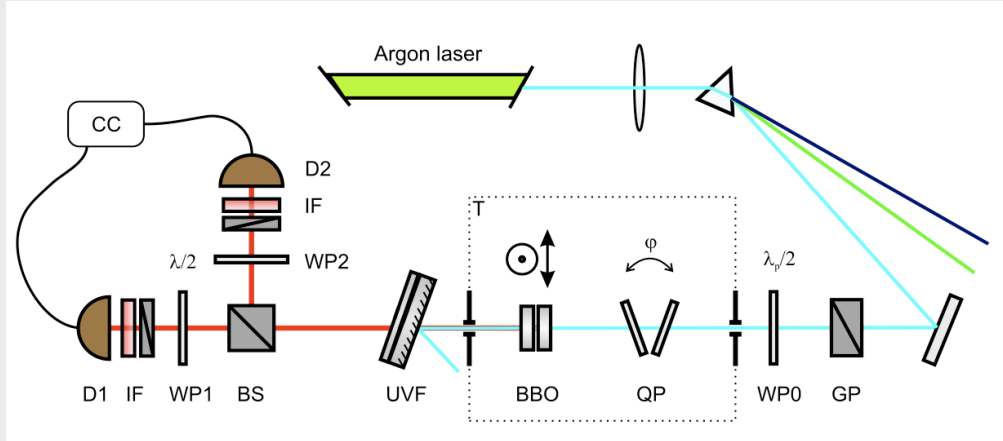


■ Estimation of entanglement

- different measures (negativity, entropy, distance) and families of states (qubit and CV)
- QFI is increasing with entanglement
QSNR diverges for maximal entanglement
- Qubit: QSNR is vanishing for vanishing entanglement
Estimation of (low) entanglement is inherently inefficient
- CV: appropriate entanglement measure may achieve efficient estimation



■ Estimation of entanglement (@INRIM)



$$|\psi_\phi\rangle = \cos \phi |HH\rangle + \sin \phi |VV\rangle$$

$$D_\phi = \cos^2 \phi |HH\rangle\langle HH| + \sin^2 \phi |VV\rangle\langle VV|$$

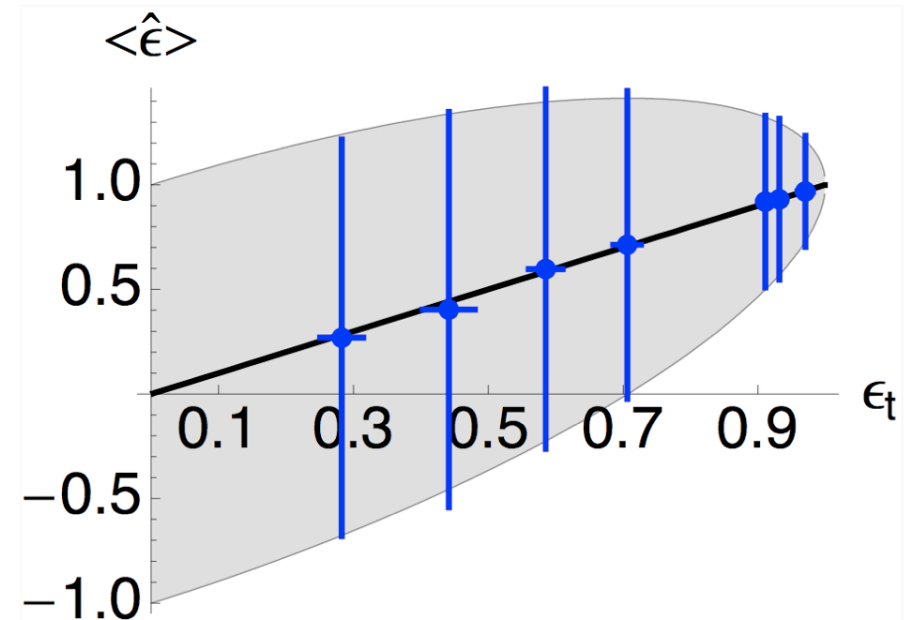
$$\rho_\epsilon = p |\psi_\phi\rangle\langle\psi_\phi| + (1 - p) D_\phi$$

$$\epsilon = p \sin 2\phi$$

optimal estimation by visibility measurements

Fisher information is monotone with entanglement

Estimation of (low) entanglement is inherently inefficient



Experimental Estimation of Entanglement at the Quantum Limit

Giorgio Brida,¹ Ivo Pietro Degiovanni,¹ Angela Florio,^{1,2} Marco Genovese,¹ Paolo Giorda,³ Alice Meda,¹ Matteo G. A. Paris,^{4,5} and Alexander Shurupov^{6,1,7}

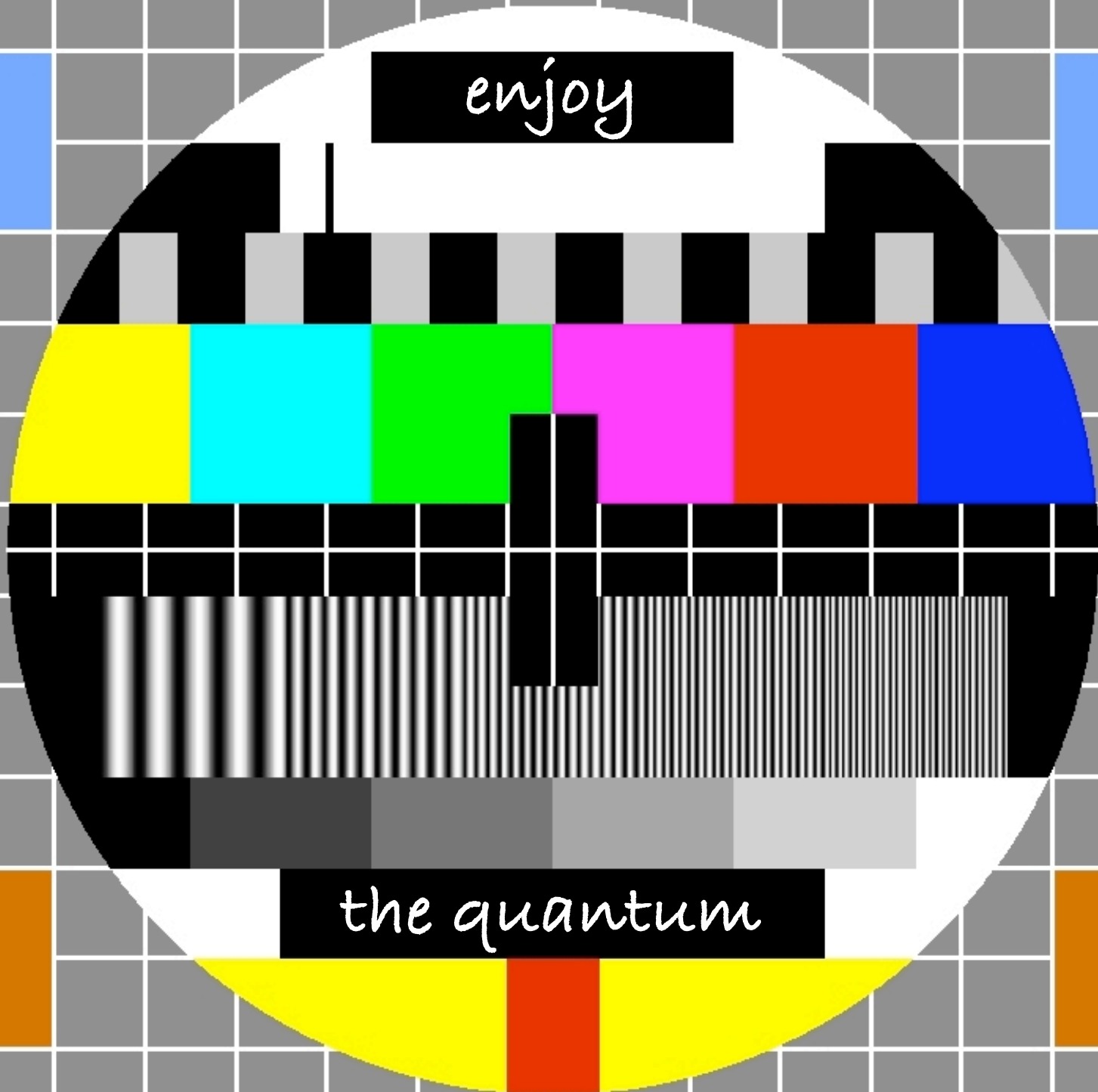


■ Summary

- Quantum estimation for quantum technology:
 - Optimal quantum measurement in terms of SLD and ultimate bounds to the precision of the estimation of any quantity of interest including non-observables
 - intrinsic estimability of a parameter
 - classical and quantum contributions to uncertainty
- Quantum estimation of nonobservable quantities
 - coupling constants
 - interferometry
 - entanglement
 - ...

enjoy

the quantum





■ (classical) Bayesian estimators (1)

● Bayes theorem $p(x|\lambda)p(\lambda) = p(\lambda|x)p(x)$

● M independent events: a posteriori distribution

$$p(\lambda|\{x\}) = \frac{1}{N} \prod_{k=1}^M p(x_k|\lambda) \quad N = \int d\lambda \prod_{k=1}^M p(x_k|\lambda)$$

● Bayesian estimator: $\lambda_B = \int d\lambda \lambda p(\lambda|\{x\})$

mean of the a posteriori distribution



■ (classical) Bayesian estimators (2)

- Laplace - Bernstein - von Mises theorem

$$p(\lambda|\{x\}) \xrightarrow{M \gg 1} G(\lambda^*, \sigma^2)$$

- Bayes estimator is asymptotically efficient

$$\sigma^2 = \frac{1}{MF(\lambda^*)}$$



■ MaxLik estimation

- Probability distribution $p(x|\lambda)$
- Random sample x_1, x_2, \dots, x_M
- Joint probability of the sample

$$\mathcal{L}(x_1, x_2, \dots, x_M | \lambda) = \prod_{k=1}^M p(x_k | \lambda)$$

Maxlik estimation \rightarrow take the value of the parameters which maximize the likelihood of the observed data