

# Performance measure for quantum tomography

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# Outline

Quantum measurement and SVD decomposition

Fisher information matrix and quantum tomographic transfer function (qTTF)

Tomography in phase space

Optics: Image Processing to Wave Front Sensing

# Measurement

Born rule for (normalized) measurement,  
M- channels,  $D^2 - 1$  parameters

$$p_j = \text{tr}(\rho \Pi_j) \quad \rho = \sum_i t_i \Omega_i \quad p'_j = p_j - \text{tr} \Pi_j / D$$

Linear structure of the measurement for  
tomographically (over)complete measurement  $M > D^2 - 1$

$$\mathbf{p}' = \mathbf{C} \mathbf{t}$$

Dimensions:  $M \times 1 = M \times (D^2 - 1) \quad (D^2 - 1) \times 1$

$$\mathbf{C} = \mathbf{O} \mathbf{S} \mathbf{O}'^T$$

SVD and its meaning (S-diagonal,  $\mathbf{O}^T \mathbf{O} = \mathbf{O}'^T \mathbf{O}' = \mathbf{O} \mathbf{O}'^T = \mathbf{1}$ )

$$(\mathbf{O}^T \mathbf{p}') = \mathbf{S} (\mathbf{O}'^T \mathbf{t})$$

## Quest for Tomo Resolution Measure

Attempts to link the resolution with S-eigenvalues:  
conditional numbers of SVD - problems with channel  
duplication, comparison between SIC and MUB ...

Answer: Likelihood and Fisher information matrix

$$\mathcal{L}(\{n_j\}; \rho) = \prod_k p_k^{n_k}$$

$$\mathbf{F}(\rho) = \frac{1}{N} \overline{\left( \sum_k \frac{n_k}{p_k} \frac{\partial p_k}{\partial \mathbf{t}} \right) \left( \sum_l \frac{n_l}{p_l} \frac{\partial p_l}{\partial \mathbf{t}} \right)}$$

$$\mathbf{F}(\rho) = \sum_l \text{tr}(\mathbf{\Omega} \Pi_l) \frac{1}{p_l} \text{tr}(\Pi_l \mathbf{\Omega}) = \mathbf{C}^T \mathbf{P}^{-1} \mathbf{C}$$

## Fisher info as a measure

Quantum Tomographic Transfer Function (qTTF): the universal resolution measure quantifying the overall performance - depends on state, measurement, ...

Consider  $\text{Sp}[\mathbf{F}(\rho)^{-1}]$

Be aware that the inversion does not hold in general!

$$(\mathbf{C}^T \mathbf{P}^{-1} \mathbf{C})^{-1} = \mathbf{C}^{-1} \mathbf{P} (\mathbf{C}^T)^{-1}$$

but using SVD

$$\text{Sp} \mathbf{F}(\rho)^{-1} = \text{Sp}(\mathbf{S} \mathbf{O}^T \mathbf{P}^{-1} \mathbf{O} \mathbf{S})^{-1}$$

Important:  $\text{Sp} \mathbf{F}^{-1}$  does not depend on the state representation ( $O'$ ) but on the measurement through  $S$  and  $O$  matrices!

# qTTF

$\text{SpF}^{-1}$  gives the scaled optimal tomographic accuracy in the Hilbert-Schmidt norm for unbiased state estimators  $\rho$  in the limit of large sampling events, but is very difficult to handle!

**Proposal:**  $q\text{TTF} = \langle \text{SpF}^{-1} \rangle_{\text{all pure states}}$

Fischer Info Matrix is difficult to deal with in general!

Hint: expand  $\text{SpF}^{-1}$  around maximally mixed state and use the Matrix Inversion Lemma

$$(A + BCD)^{-1} = A^{-1} - A^{-1}BCDA^{-1}(1 + BCDA^{-1})^{-1}$$

$$\Delta P = P - \bar{P} \quad \bar{p}_j = \text{tr}(\Pi_j/D)$$

$$\text{Sp}F(\rho)^{-1} = \text{Sp}\bar{F}^{-1} + \text{Sp}\frac{\mathcal{X}\Delta P}{\mathbf{1} - \mathcal{Y}\Delta P}$$

$$\mathcal{X} = \bar{P}^{-1}C\bar{F}^{-2}C^T\bar{P}^{-1} \quad \mathcal{Y} = \bar{P}^{-1}C\bar{F}^{-1}C^T\bar{P}^{-1} - \bar{P}^{-1}$$

Mutual orthogonality

$$\mathcal{X}\bar{P}\mathcal{Y} = 0 = \mathcal{Y}\bar{P}\mathcal{X}$$

$-\bar{P}$  is generalized inverse of  $\mathcal{Y}$

$$\mathcal{Y}\bar{P}\mathcal{Y} = -\mathcal{Y}$$

$$C^T\mathcal{Y} = 0 = \mathcal{Y}C$$

All these subtle relations are important for evaluation of qTTF in special cases!

## State averaging for qTTF ...

$$qTTF\{\Pi_j\} = \underbrace{\text{Sp}\bar{F}^{-1}}_{\text{zeroth order}} + \underbrace{\frac{\alpha}{D(D+1)} \sum_{j_1, j_2=1}^M \mathbf{x}_{j_2 j_1} \mathbf{y}_{j_1 j_2} \mathbf{g}_{j_1 j_2}^{(2)}}_{\text{second order}} + \dots$$

Gramm matrix of higher order  $\mathbf{g}_{j_1 j_2 \dots j_n}^{(n)} = \text{tr}(\Pi_{j_1} \Pi_{j_2} \dots \Pi_{j_n})$

For some Minimally Complete Tomo only two lowest terms exist!!!

## Special cases SIC and MUB...

Minimally complete tomo (SIC,...)

$$qTTF\{\Pi_j\}_{\text{MIN}} = \text{Sp}\overline{\mathbf{F}}_{\text{MIN}}^{-1} - 1 + \frac{1}{D}$$

$$qTTF\{\Pi_j\}_{\text{SIC}} = D^2 + D - 2$$

Minimally basis (over-complete) tomo (MUB,...)

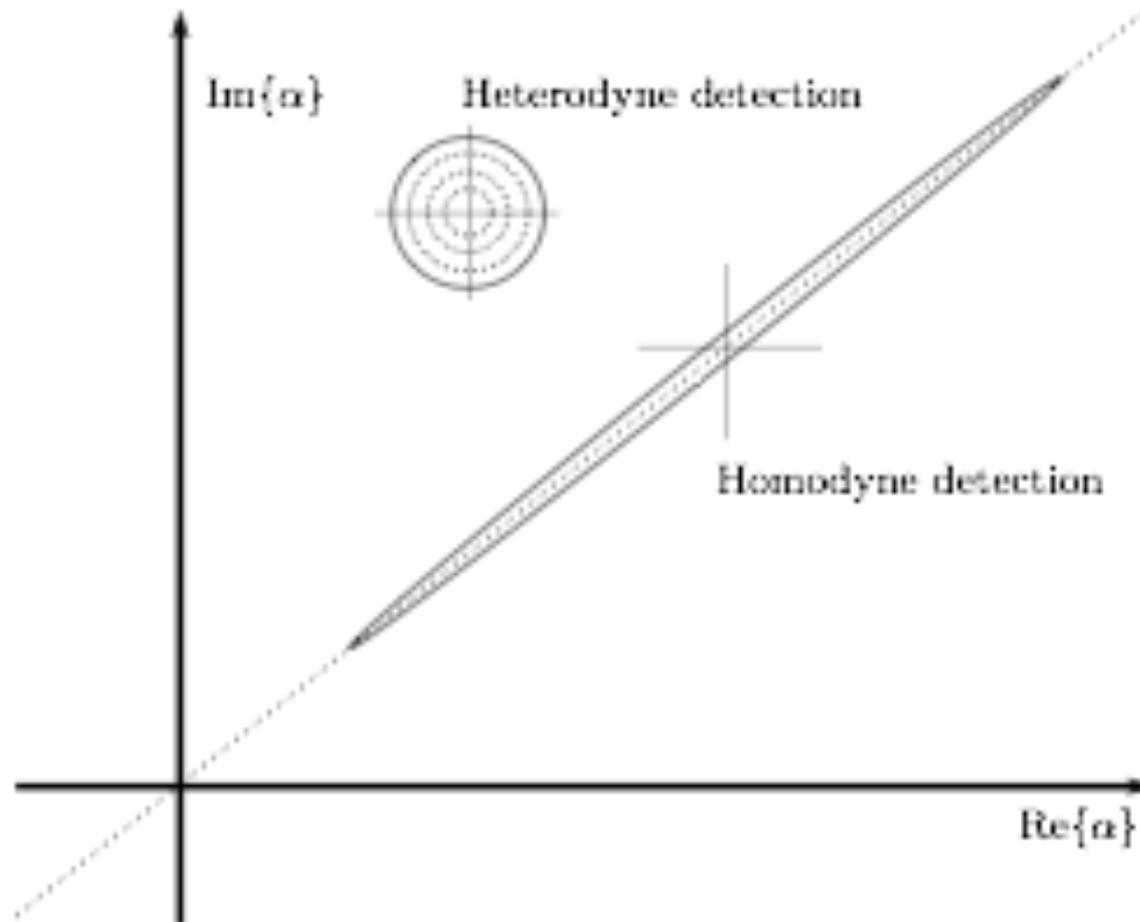
$$qTTF\{\Pi_j\}_{\text{MIN BASES}} = \frac{\text{Sp}(\mathbf{C}^T \mathbf{C})^{-1}}{(D+1)^2}$$

$$qTTF\{\Pi_j\}_{\text{MUB}} = D^2 - 1 < qTTF\{\Pi_j\}_{\text{SIC}}$$

## Conceptual vs practical issues...

- $q$ TTF has good meaning as statistical measure of accuracy
- Dependence on state, (transformation), measurement, data processing and other aspect should be considered
- Tomography in phase space is good example: optical implementations available

# Phase space

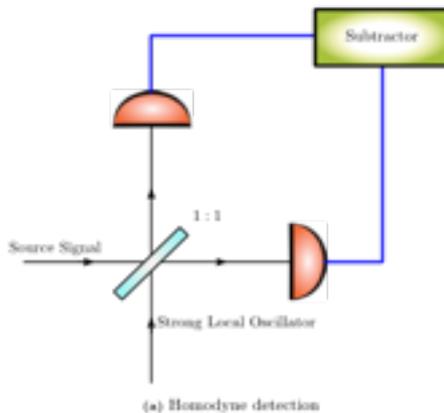


# Homodyne vs heterodyne

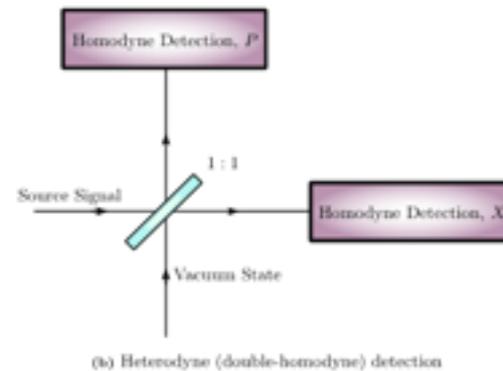
## Rotated quadrature operators

$$x_\theta = \frac{1}{2} [ae^{-i\theta} + a^\dagger e^{i\theta}] \quad p_\theta = \frac{1}{2i} [ae^{-i\theta} - a^\dagger e^{i\theta}]$$

### Homodyne



### Heterodyne

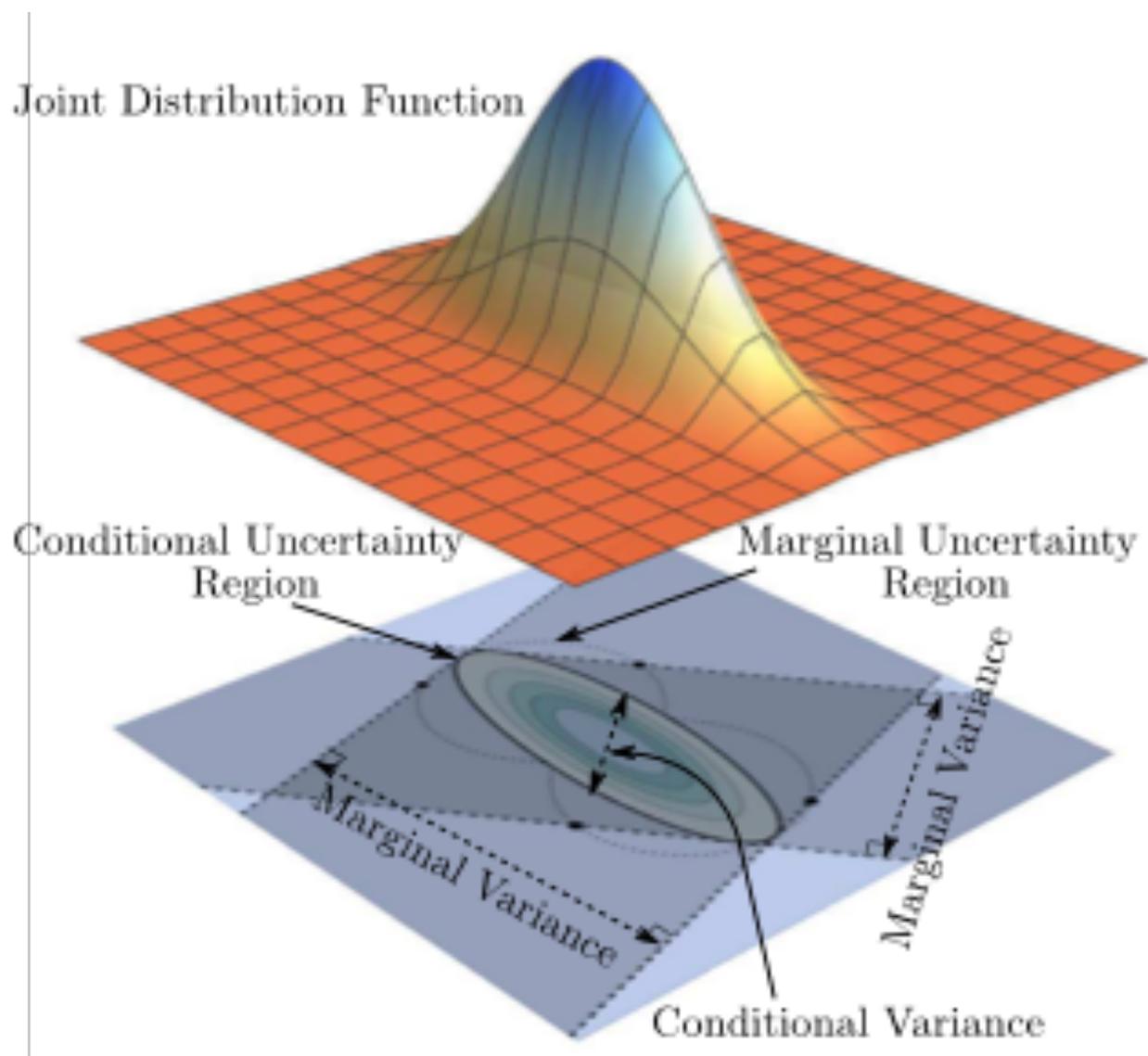


# Tomography in phase space

**Homodyne detection:** Marginal distribution of Wigner function is sampled, the covariance must be reconstructed (more involved reconstruction)

**Heterodyne detection:** The Q function is sampled directly (direct reconstruction)

**Tomography = Detected noise + error from inversion**



# Mathematical tools for reconstruction of covariance matrix

Estimated covariance

Hilbert-Schmidt distance

Cramer- Rao bound

Fisher information

$$\mathbf{G}_w \hat{=} \begin{pmatrix} g_1 & g_3/\sqrt{2} \\ g_3/\sqrt{2} & g_2 \end{pmatrix}$$

$$H = \overline{(\mathbf{G}_w - \hat{\mathbf{G}}_w)^2} = \sum_k \overline{(g_k - \hat{g}_k)^2}.$$

$$H \geq Sp \mathbf{F}^{-1}$$

$$\mathbf{F} = \frac{N}{2} Sp \mathbf{G}^{-1} \frac{\partial \mathbf{G}}{\partial \mathbf{g}} \mathbf{G}^{-1} \frac{\partial \mathbf{G}}{\partial \mathbf{g}}$$

# Rationale behind the noise analysis

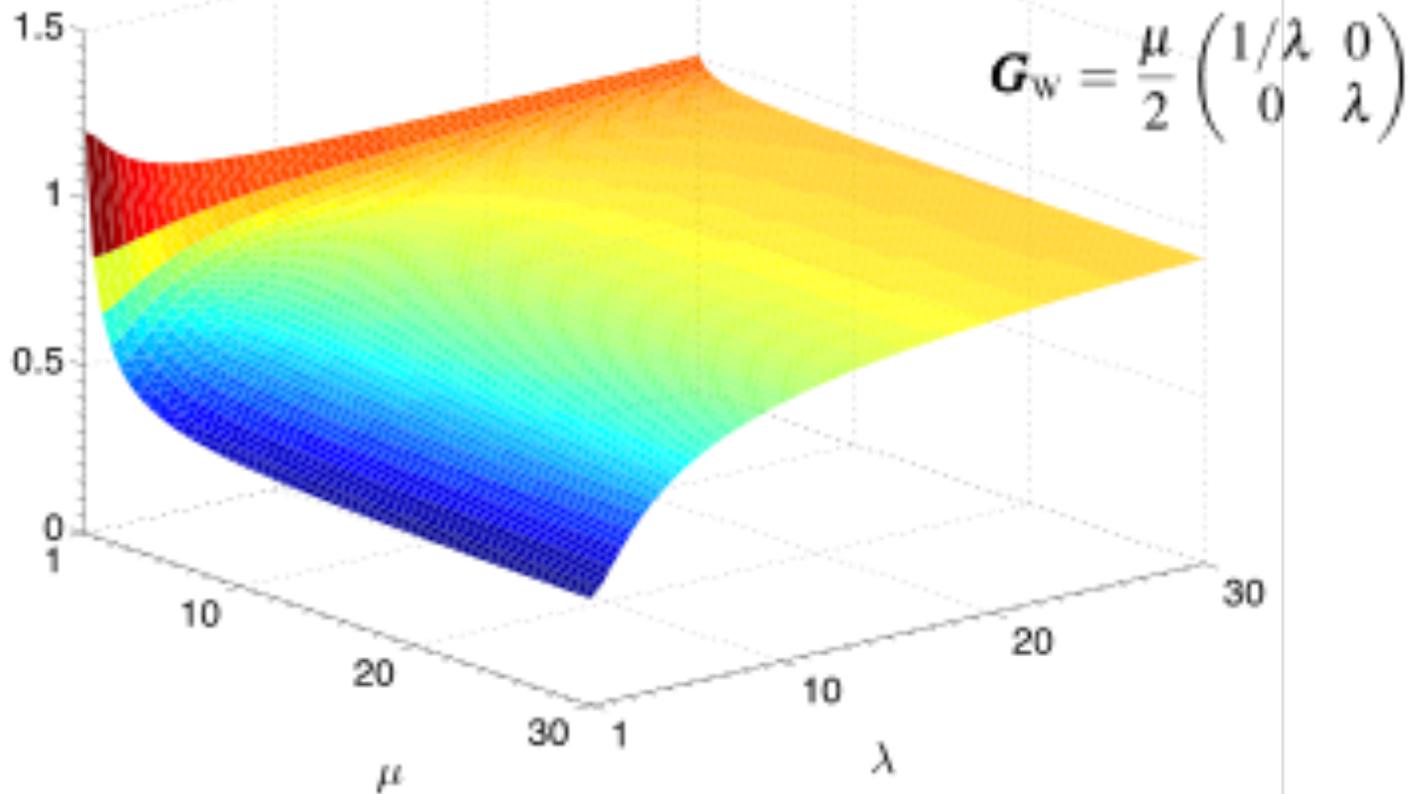
Performance of  
homodyne tomography

$$H_{\text{hom}} = \frac{2 \text{Sp } \mathbf{G}_{\text{hom}} (\text{Sp } \mathbf{G}_{\text{hom}} + 3\sqrt{\text{Det } \mathbf{G}_{\text{hom}}})}{N}$$

Performance of  
heterodyne detection

$$H_{\text{het}} = \frac{2[(\text{Sp } \mathbf{G}_{\text{het}})^2 - \text{Det } \mathbf{G}_{\text{het}}]}{N}$$

## Comparison Het and Homo Tomo strategies



Ratio  $\leq 1$  : Het is doing better !

# Heterodyne vs heterodyne data

Covariance matrices for Gaussian states:

$$\mathbf{G}_Q = \mathbf{G}_W + 1/2$$

Variance of homodyne detection: marginal distribution of Wigner function

$$\sigma_\theta^2 = u_\theta^T \mathbf{G}_W u_\theta + \delta_\eta^2$$

Conditional variance of Q function (along the line in the phase space)

$$\Sigma_\theta^2 = \left( u_\theta^T [\mathbf{G}_Q + \delta_\eta^2]^{-1} u_\theta \right)^{-1}$$

Noise term:

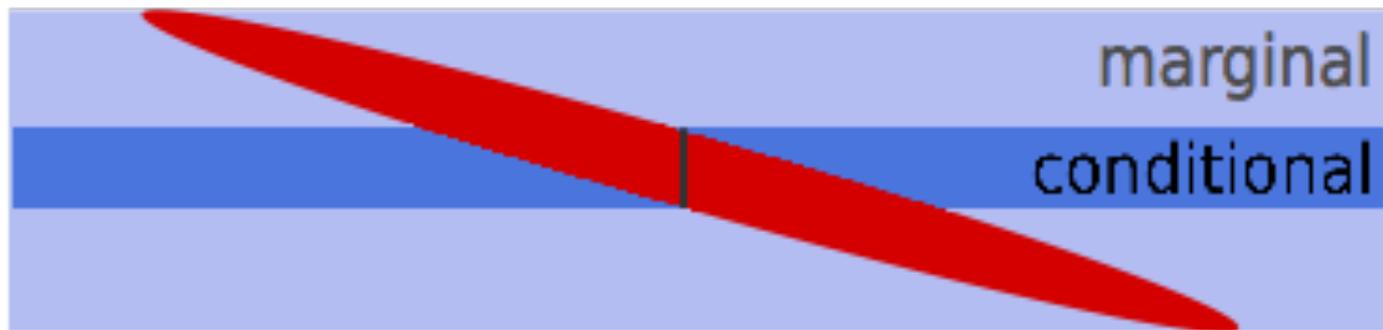
$$\delta_\eta^2 = (1 - \eta)/2\eta$$

# Geometrical relation between marginal and conditional distributions

Relation for generic covariance matrix  $\mathbf{G}$  and orthogonal basis vectors  $u, v$

$$\sigma_{\theta}^2 = u_{\theta}^T \mathbf{G} u_{\theta} \equiv G_{uu,\theta}$$

$$\Sigma_{\theta}^2 = (u_{\theta}^T \mathbf{G}^{-1} u_{\theta})^{-1} = \frac{G_{uu,\theta} G_{vv,\theta} - G_{uv,\theta}^2}{G_{vv,\theta}} \leq G_{uu,\theta} = \sigma_{\theta}^2.$$



# Noise analysis for minimum uncertainty states

Wigner covariance matrix

$$\mathbf{G}_w \hat{=} \begin{pmatrix} \frac{1}{2\lambda} & 0 \\ 0 & \frac{\lambda}{2} \end{pmatrix}$$

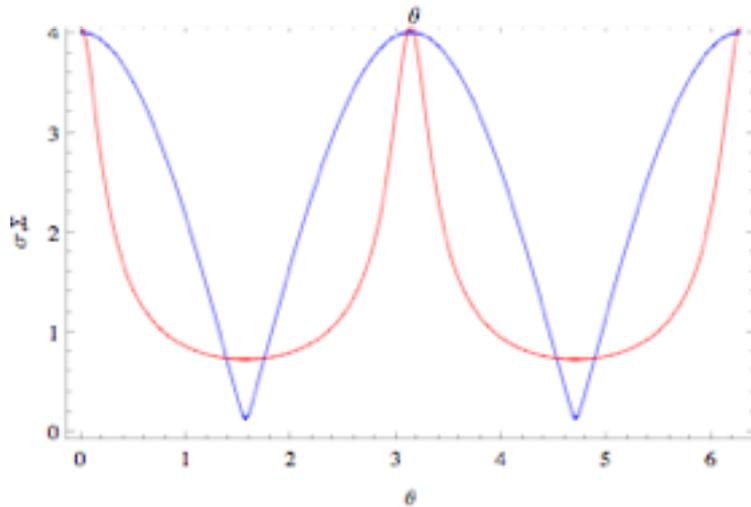
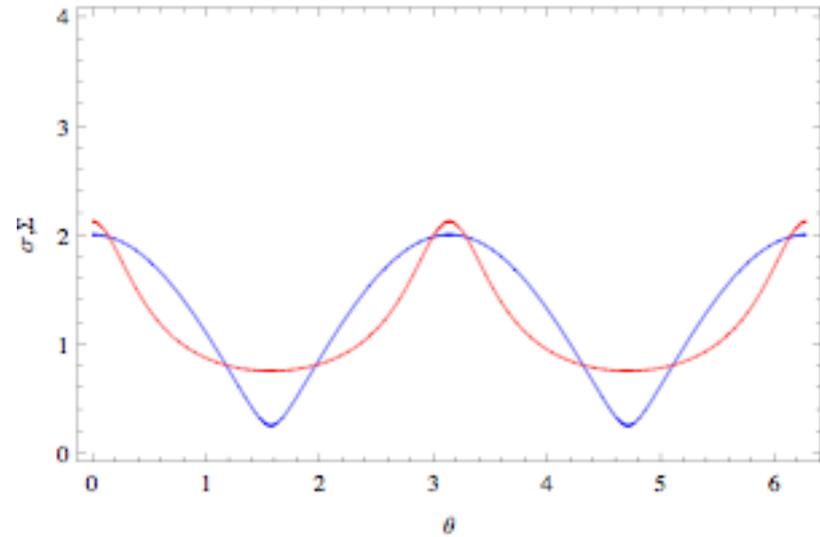
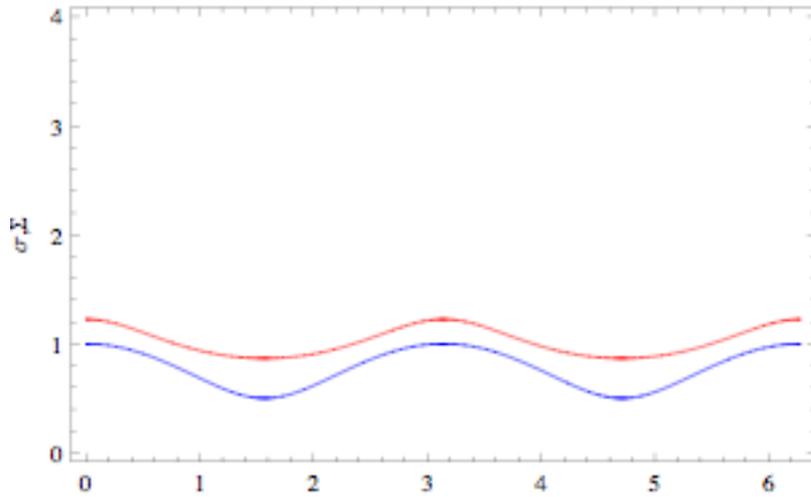
Marginal variance

$$\sigma_\theta^2 = \frac{1}{2\lambda} (\cos \theta)^2 + \frac{\lambda}{2} (\sin \theta)^2$$

Conditional variance

$$\Sigma_\theta^2 = \left[ 1 + \frac{\lambda - 1}{\lambda + 1} \cos(2\theta) \right]^{-1}$$

# Noise comparison



**BLUE:** marginal  
for homodyne detection

**RED:** conditional  
for heterodyne detection

# Measurement in Phase Space

Homodyne detection - Projection into the Rotated Quadrature Eigenstates

Heterodyne detection - Projection into the coherent state basis with fluctuating position in phase space

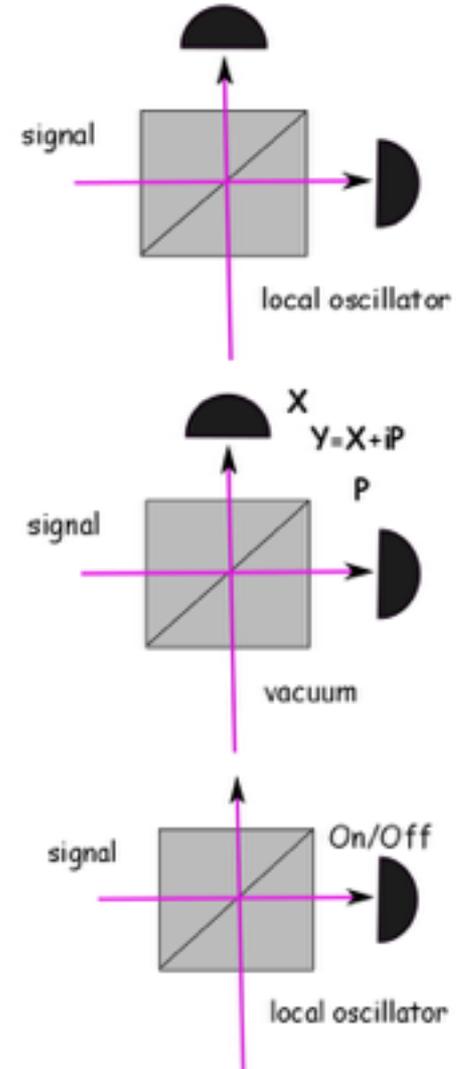
Unbalanced homodyning - Projection into the coherent state basis with prefixed position in phase space

# Heterodyne detection

Frequency mismatch between signal and local oscillator

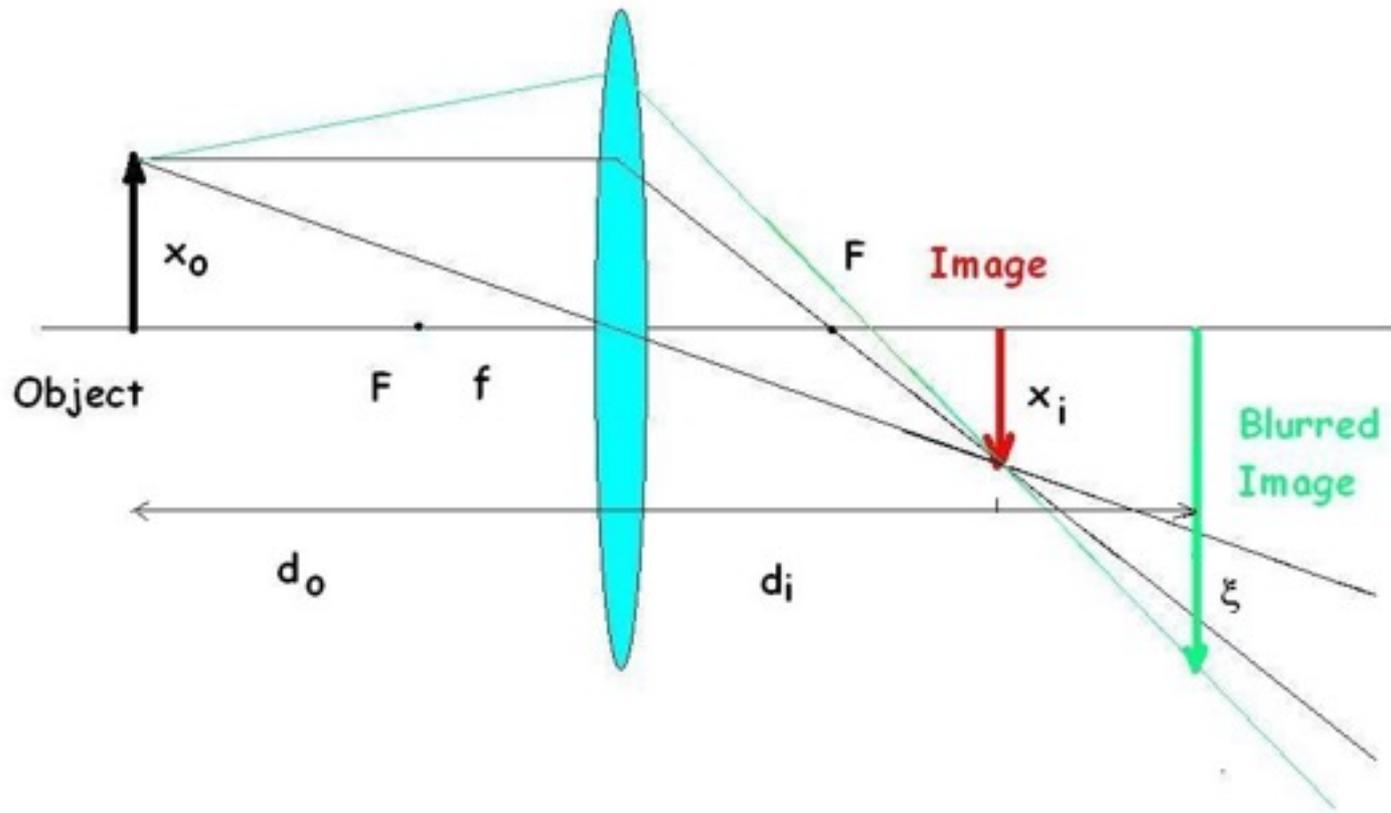
Double homodyne detection

Unbalanced homodyning: low transmittivity for LO



# Phase space in optics

# Optical Imaging: Lens equation in geometrical optics



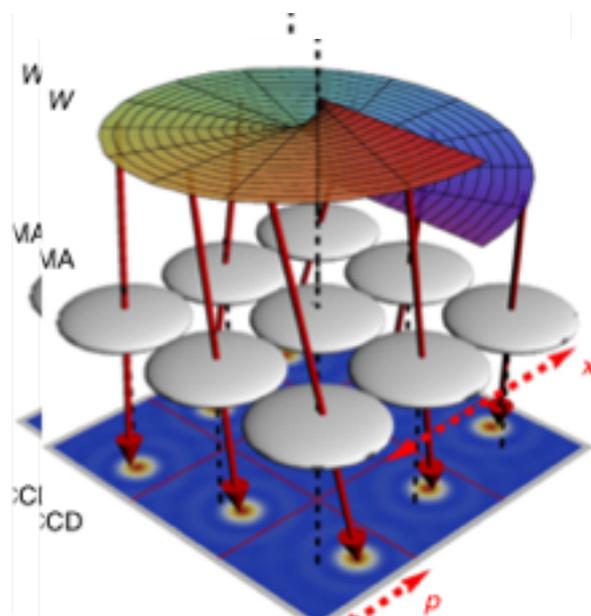
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Received 1 Sep 2013 | Accepted 17 Jan 2014 | Published 7 Feb 2014

DOI: 10.1038/ncomms4275

# Wavefront sensing reveals optical coherence

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# Conclusions

## Fisher Info Matrix provides a useful tool for assessing the performance of reconstruction schemes

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- C. R. Muller, C. Peuntinger, T. Dirmeier, I. Khan, U. Vogl, Ch. Marquardt, G. Leuchs, L.L. Sanchez Soto, Y.S. Teo, Z. Hradil, and J. Rehacek, Evading Vacuum Noise: Wigner Projections or Husimi Samples? (unpublished).
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**Thanks for your attention!**