

# Filtering, Retrofiltering and Smoothing:

OPTIMAL QUANTUM STATE ESTIMATION USING  
CONTINUOUS-IN-TIME MEASUREMENT

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**Kiarn T. Laverick**, Areeya Chantasri, Howard M. Wiseman

# Classical State Estimation Techniques

- **Task:** Assign a value to an unknown parameter  $x$  of a system based on observations.
- If we can obtain the distribution  $\varphi(x)$ , we can calculate many useful properties, e.g. the mean  $\langle x \rangle$ .

- **Filtering (F):**

Conditioning on a past measurement records i.e.

$$\varphi_F(x) = \varphi(x | \overleftarrow{Y})$$

- **Retrofiltering (R):**

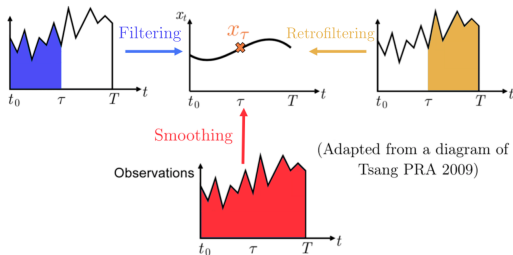
Predicting future measurement records i.e.

$$E_R(x) = \varphi(\overrightarrow{Y} | x) \Leftrightarrow \varphi_R(x)$$

- **Smoothing (S):**

Conditioning on an entire measurement record  $\overleftrightarrow{Y}$  i.e.

$$\varphi_S(x) \propto E_R(x) \varphi_F(x)$$



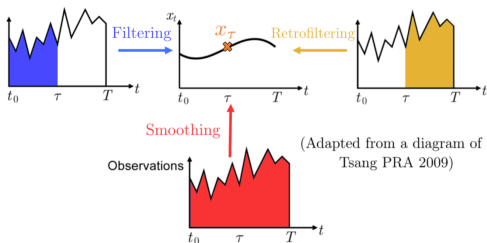
# Quantum State Estimation Techniques

- Transitioning to quantum systems, the problem remains, more or less, the same, where now we are concerned with estimating the quantum state  $\rho$
- **Quantum Filtering:**  
“Quantum Trajectory Theory”  $\Rightarrow \rho_F = \rho_{\vec{O}}$
- **Quantum Retrofiltering:**  
 $\hat{E}_R \Rightarrow \text{Tr}[\hat{E}_R \rho] = \wp(\vec{O}|\rho)$ ,  
POVM describing future results
- **Weak-Value Smoothing:**  
 $\rho_{\text{SWV}} \propto \hat{E}_R \rho_F + \rho_F \hat{E}_R$

$$\Lambda_W = \text{Tr}[\rho_{\text{SWV}} \hat{\Lambda}] \quad (1)$$

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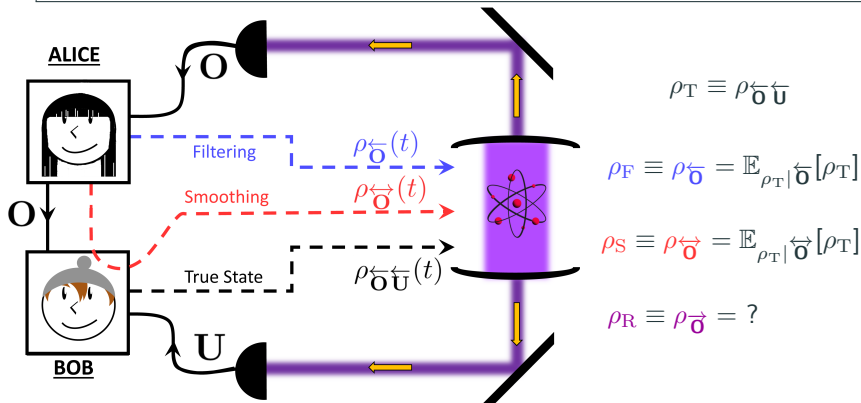
PRL 115, 180407 (2015)

PHYSICAL REVIEW LETTERS

week ending  
30 OCTOBER 2015

## Quantum State Smoothing

Ivonne Guevara and Howard Wiseman



# Optimal Bayesian Estimation of the True State

- Objective Function:

$$\mathcal{B}_c(\check{\rho}) = \mathbb{E}_{\rho_T | \mathbf{O}_c} \{ \mathcal{C}[\rho_T, \check{\rho}] \}$$

- Cost Function: Trace-Square Deviation

$$\mathcal{C}^{\text{TrSD}}[\rho_T, \check{\rho}] = \text{Tr}[(\rho_T - \check{\rho})^2]$$

- Optimal Estimator:

$$\rho_c^{\text{opt}}(t) := \arg \min_{\check{\rho} \in \mathcal{T}_t} \mathcal{B}_c(\check{\rho}) = \mathbb{E}_{\rho_T | \mathbf{O}_c} \{ \rho_T \}$$

Filtered

$$\rho_F(t) = \mathbb{E}_{\rho_T | \overleftarrow{\mathbf{O}}} \{ \rho_T \}$$

Retrofiltered

$$\rho_R^{\text{Tru}}(t) = \mathbb{E}_{\rho_T | \overrightarrow{\mathbf{O}}} \{ \rho_T \}$$

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# Estimating the Filtered State and Weak-Values

- One could instead ask the question, what if one wishes to estimate the filtered state, given they only have access to the future measurement record.
- This problem is a slight change to the previous estimation problem ( $\rho_T \rightarrow \rho_F$ ), with the optimal estimator:

Retrofiltered Estimate of the Observed State (REOS)

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## Estimating the Filtered State and Weak-Values cont.

- The final type of estimation we consider is weak-value estimation.
- We have already seen one of these estimators:  $\varrho_{\text{SWV}}$
- One aims to find an estimator that minimizes the distance between its trace with a set of observables  $\hat{\Lambda}_k$  (the generalized Gell-Mann matrices [1]) and their weak-value  $\Lambda_k^w$  [2], i.e.,

$$\varrho_c = \arg \min_{\check{\varrho}} \sum_k \mathbb{E}_{\Lambda_k | \mathbf{O}_c} \{ (\text{Tr}[\check{\varrho} \hat{\Lambda}_k] - \Lambda_k^w)^2 \}$$

- Optimal Weak-valued states:

$$\varrho_c(t) = \frac{1}{d} \hat{\mathbf{1}} + \frac{1}{2} \sum_k \langle \hat{\Lambda}_k \rangle_c \hat{\Lambda}_k, \quad (2)$$

$$\text{with } \langle \hat{\Lambda}_k \rangle_c = \sum_{x_j} \wp(x_j | \mathbf{O}_c) x_j$$

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[1] G. Kimura, Phys. Lett. A, **314**, 339–349, (2003).

[2] Chantasri, A., Guevara, I., Laverick, K. T., Wiseman, H. M., Phys. Rep. **930**, 1–40, (2021).

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## Example: On-Threshold Optical Parametric Oscillator

Let's consider an example, the optical parametric oscillator.

$$\frac{d\rho}{dt} = -i\chi[(\hat{q}\hat{p} + \hat{p}\hat{q})/2, \rho] + \gamma\mathcal{D}[\hat{q} + i\hat{p}]\rho, \quad (3)$$

where  $\mathcal{D}[\hat{c}]\rho = \hat{c}\rho\hat{c}^\dagger - \frac{1}{2}\{\hat{c}^\dagger\hat{c}, \rho\}$ .

- We will consider the system at threshold, i.e.,  $\chi = \gamma$ .
- Alice and Bob both use homodyne measurements with local oscillator phases  $\theta_A$  and  $\theta_B$ . Their respective measurement efficiencies are  $\eta_A = \eta_B = 0.5$ .
- Note: this is a linear Gaussian quantum (LGQ) system, so one only needs to know the mean and covariance matrix to specify the state.

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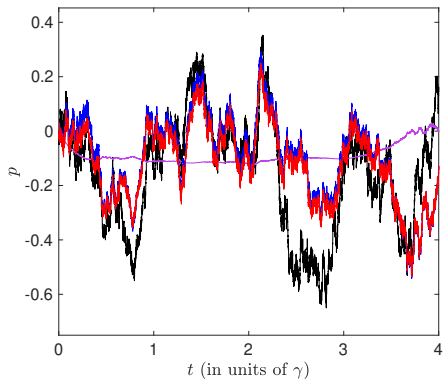
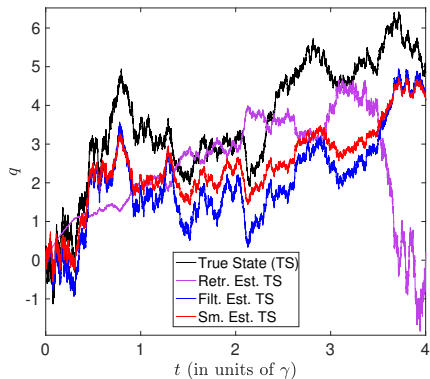
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# Example: OPO Trajectories

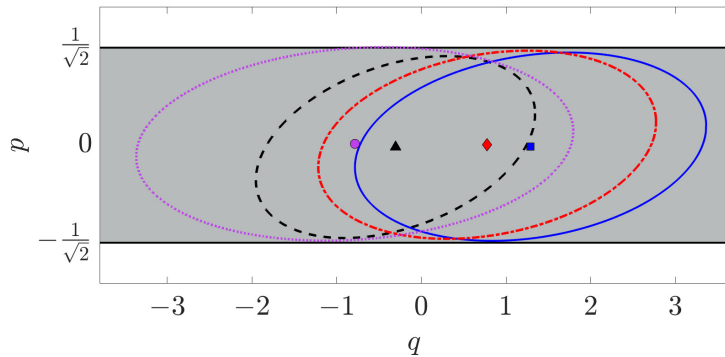


$$\theta_A = 3\pi/8$$

$$\theta_B = 3\pi/8$$

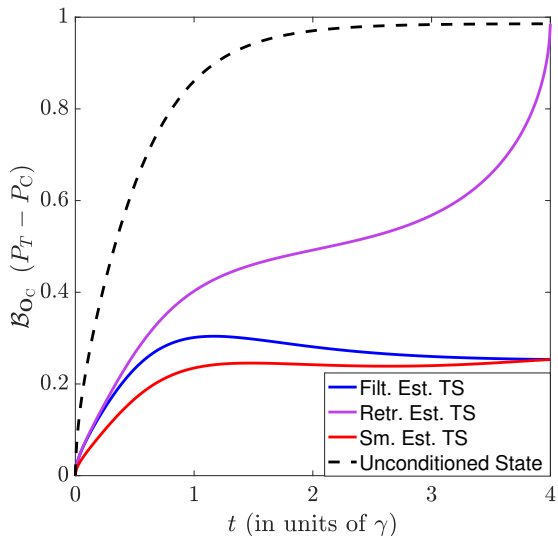
(4)

## Example: OPO Phase-Space Diagram





## Example: Expected Cost Function



# Conclusion

	Filtered ( $\overleftarrow{\mathbf{O}}$ )	Retrofiltered ( $\overrightarrow{\mathbf{O}}$ )	Smoothed ( $\overleftrightarrow{\mathbf{O}}$ )
True ( $\rho_T$ )	$\rho_F$	$\rho_R^{\text{Tru}}$	$\rho_S$
Observed ( $\rho_F$ )	$\rho_F$	$\rho_R^{\text{Obs}}$	$\rho_F$
Weak-Valued ( $\varrho$ )	$\rho_F$	$\varrho_{\text{RWV}}$	$\varrho_{\text{SWV}}$

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Observed ( $\rho_F$ )	$\rho_F$	$\rho_R^{\text{Obs}}$	$\rho_F$
Weak-Valued ( $\rho$ )	$\rho_F$	$\rho_{RWV}$	$\rho_{SWV}$

Thursday, 15 December, 2:45 pm

2:45 PM Conditional quantum states  
of a continuously monitored mec...



3:00 PM

Location: Room R6, Adelaide Convention Centre

Speaker: Soroush Khademi

7 0 0

## Extra Slide: LGQ Equations

### True Retrofiltered State

$$\langle \hat{\mathbf{x}} \rangle_{\text{R}}^{\text{Tru}} = (V_{\text{R}}^{\text{Tru}} - V_{\text{T}}) \left[ (V - V_{\text{T}})^{-1} \langle \mathbf{x} \rangle + (V_{\text{R}} + V_{\text{T}})^{-1} \langle \hat{\mathbf{x}} \rangle_{\text{R}} \right] \quad (5)$$

$$V_{\text{R}}^{\text{Tru}} = \left[ (V - V_{\text{T}})^{-1} + (V_{\text{R}} + V_{\text{T}})^{-1} \right]^{-1} + V_{\text{T}} \quad (6)$$

### Observed Retrofiltered State

$$\langle \hat{\mathbf{x}} \rangle_{\text{R}}^{\text{Obs}} = (V_{\text{R}}^{\text{Obs}} - V_{\text{F}}) \left[ (V - V_{\text{F}})^{-1} \langle \mathbf{x} \rangle + (V_{\text{R}} + V_{\text{F}})^{-1} \langle \hat{\mathbf{x}} \rangle_{\text{R}} \right] \quad (7)$$

$$V_{\text{R}}^{\text{Obs}} = \left[ (V - V_{\text{F}})^{-1} + (V_{\text{R}} + V_{\text{F}})^{-1} \right]^{-1} + V_{\text{F}} \quad (8)$$

### Retrofiltered Weak-Valued State

$$\langle \mathbf{x} \rangle_{\text{RWV}} = V_{\text{RWV}} V_{\text{R}}^{-1} \langle \hat{\mathbf{x}} \rangle_{\text{R}} \quad (9)$$

$$V_{\text{RWV}} = (V_{\text{R}}^{-1} + V^{-1})^{-1} \quad (10)$$