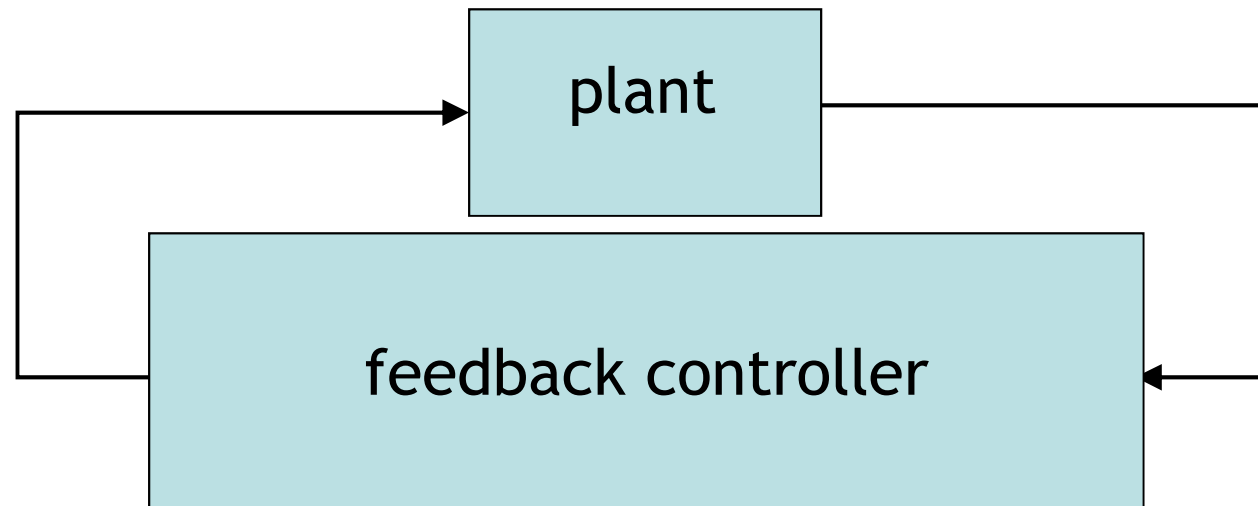


State observers in real-time feedback control



Based on:

- *Model* of plant intrinsic dynamics and exogenous disturbances,
- Knowledge of (noisy, partial) measurement record over some past time interval,

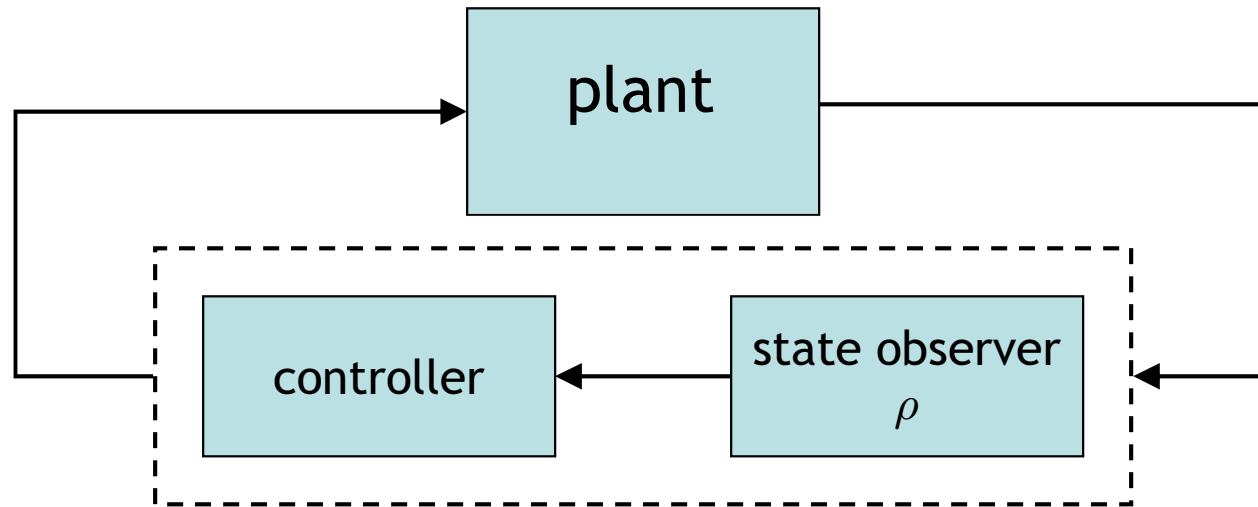
Goals:

- Predict statistics of possible future measurements as accurately* as possible,
- Utilize feedback to alter these statistics in desired ways.

Controller can maintain jointly sufficient statistics of the past measurement records, updated recursively: the *information state*. In the Kalman filter, e.g., the conditional mean and variance are sufficient statistics for a Gaussian posterior probability distribution on the plant coordinate.

*The most common – but not unique – measure of accuracy is mean-square error.

Quantum filtering: ρ as the *information state*



Quantum measurement-based feedback:

- Information state $\leftrightarrow \rho_c$
- Recursive filter \leftrightarrow Stochastic Master/Schrodinger Equation
- In the canonical design problem, the plant (S,L,H) is given and we are free to choose a specific way of monitoring the output fields (*e.g.*, homodyne or photon counting) and to design a control law with real-time actuation of plant input fields and/or adjustable parameters in the plant Hamiltonian
- The usual, unconditional Master Equation for the plant can be viewed as the “open-loop” Master Equation
- If we implement our measurement-feedback scheme and then re-average over all the noises, we should (formally speaking) obtain a “closed-loop” Master Equation with different properties (although this is not generally feasible in practice)

Quantum filtering & measurement-feedback control

L. Bouten, R. van Handel and M. R. James, SIAM Review **51**, 239 (2009); [math.PR/0606118](#)

L. Bouten and R. van Handel, [math-ph/0511021](#)

Itô amplitude-quadrature homodyne Stochastic Master Equation / Kushner-Stratonovich Equation:

$$d\rho_t = \underbrace{-i[H, \rho_t] dt + \sum_{j=1}^n \{\mathcal{D}[L_j]\rho_t\} dt}_{\text{“predictor” – the unconditional ME}} + \underbrace{\sqrt{\eta} \mathcal{H}[L_1]\rho_t \overline{dW}_t}_{\text{“corrector” – innovation term}}$$

“predictor” – the unconditional ME “corrector” – innovation term

- (S,L,H) model has n input-output channels
- We are monitoring channel 1 with quantum efficiency η

$$\overline{dW}_t = dy_t - \sqrt{\eta} \text{Tr}[(L_1 + L_1^\dagger)\rho_t] dt, \quad dy_t = I_{\text{hom}} dt$$

$$\mathcal{D}[c]\rho \equiv c\rho c^\dagger - \frac{1}{2}(c^\dagger c\rho + \rho c^\dagger c), \quad \mathcal{H}[c]\rho \equiv c\rho + \rho c^\dagger - \text{Tr}[(c + c^\dagger)\rho]\rho$$

For least-squares-optimal recursive filter, dW_t is Gaussian white with variance dt

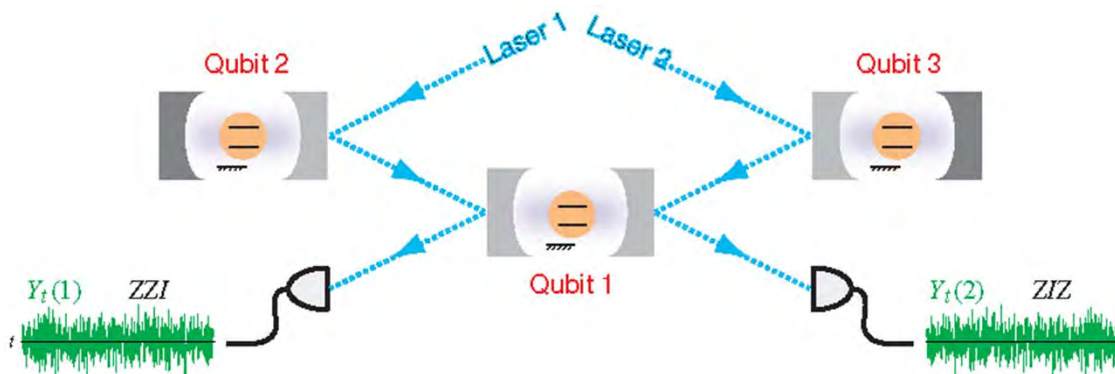
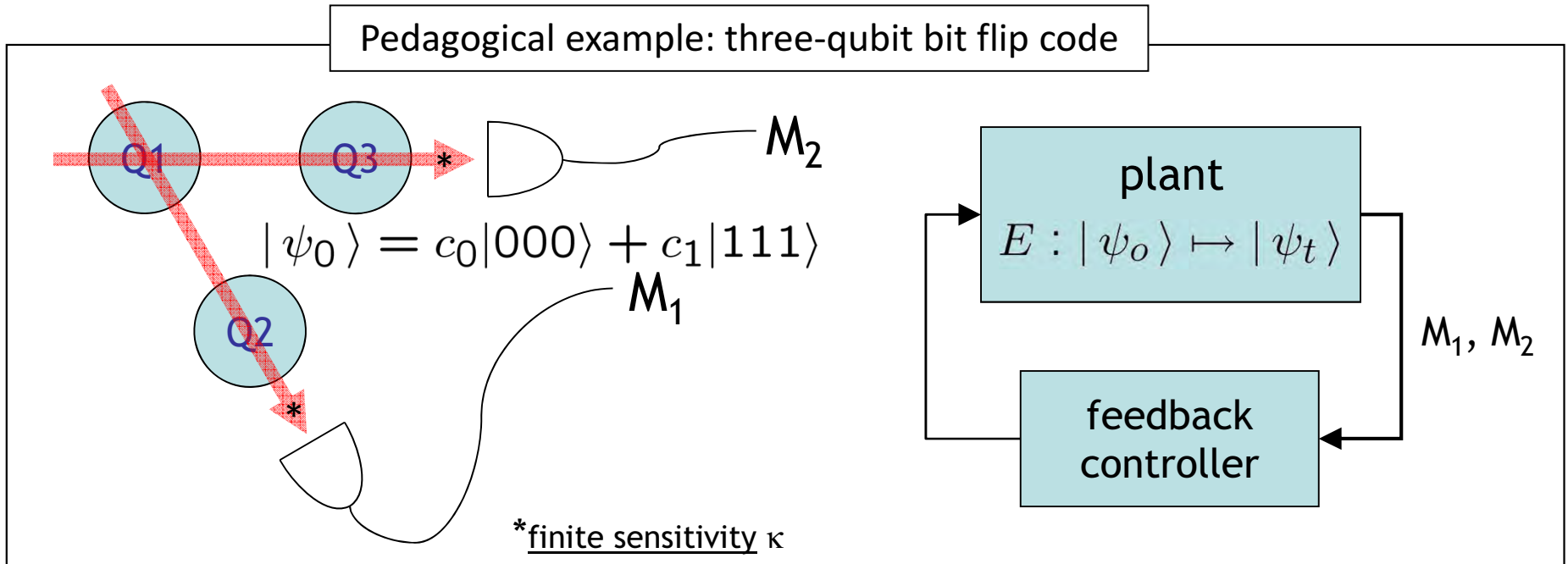
$$\text{“ } I_{\text{hom}} dt \sim \sqrt{\eta} \text{Tr}[(L_1 + L_1^\dagger)\rho_t] dt + dW_t \text{ ”}$$

Coding and continuous syndrome measurement

Continuous-time “relaxations” of QEC (Ahn, Doherty and Landahl, PRA 65, 042301, 2002)

- Encode state in a stabilizer code

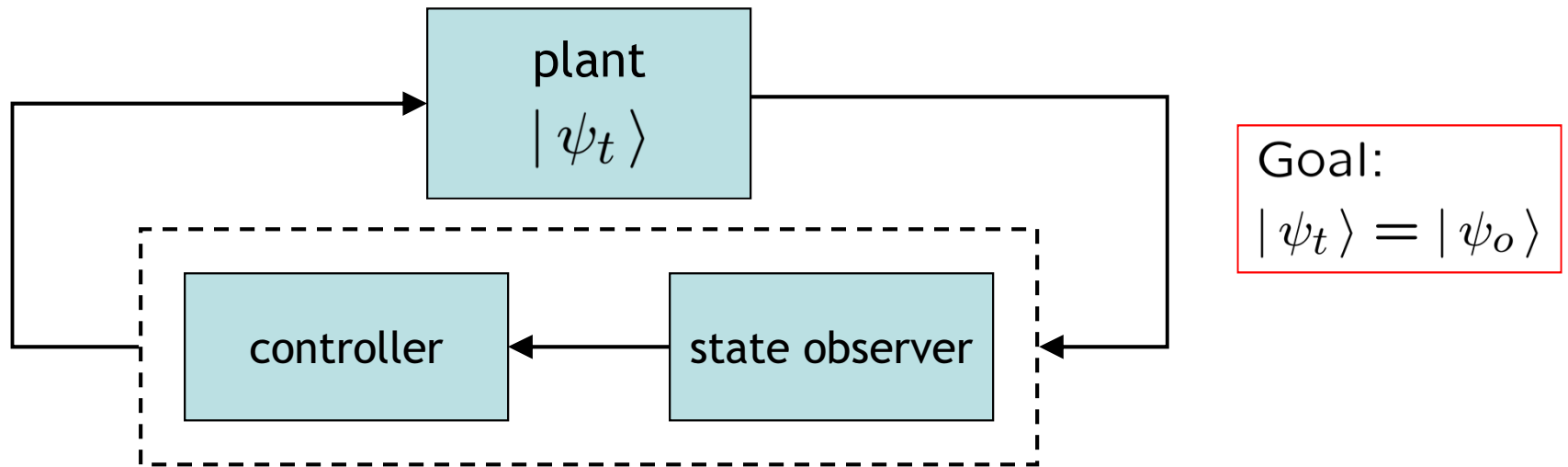
- Continuous QND measurement of syndrome



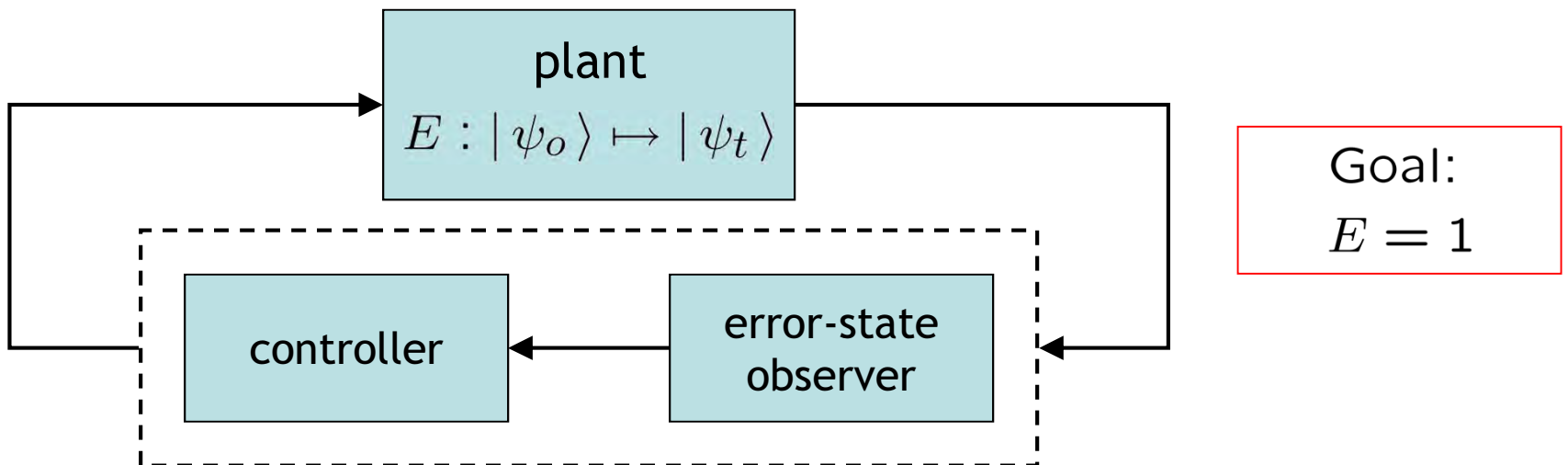
Continuous syndrome measurement:

- constant qubit-cavity couplings
- cw coherent-state laser probes
- homodyne detection

State observer \rightarrow error-state observer



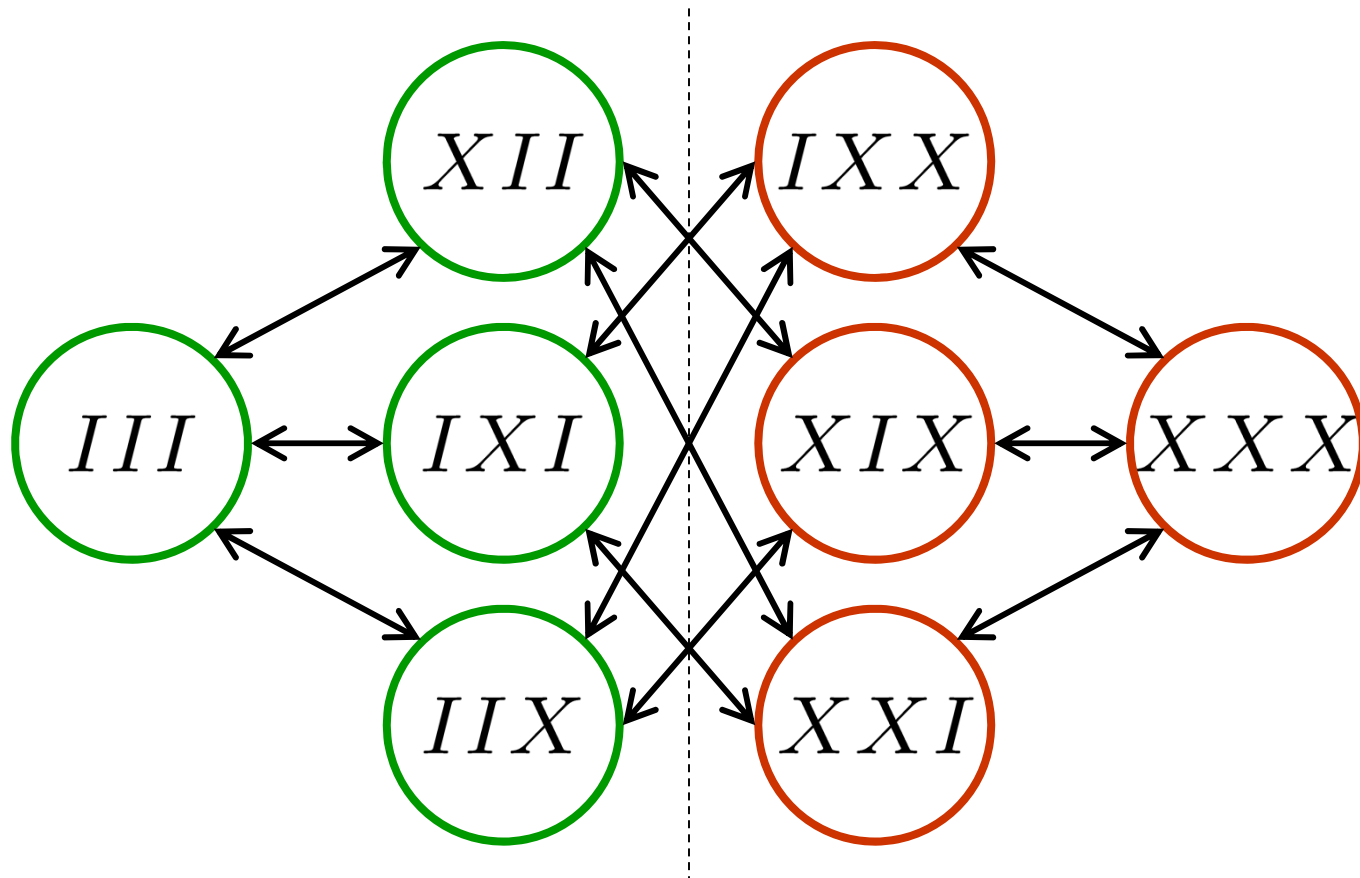
measurement should yield full information on E but none on encoded state



'Error-state graph' for the bit-flip code

$$M_1 = Z \otimes Z \otimes I$$

$$M_2 = Z \otimes I \otimes Z$$

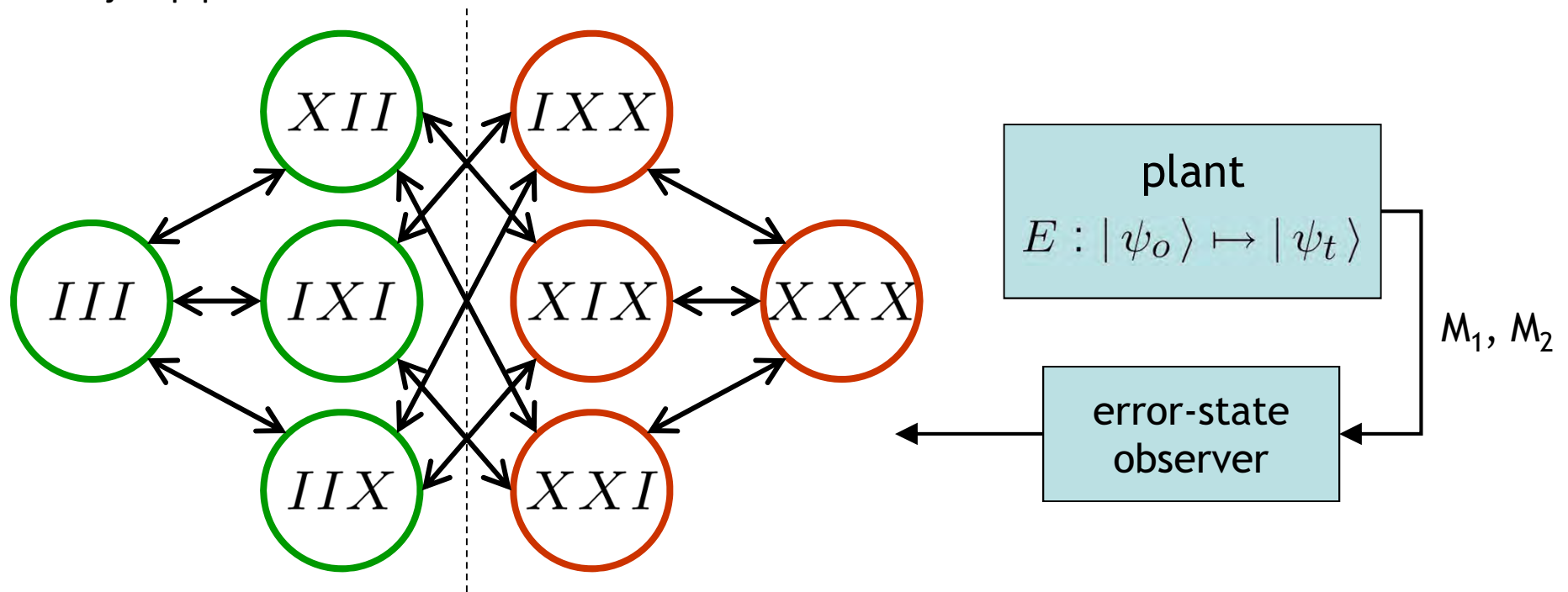


- Continuous QND syndrome measurement \Rightarrow Markov jump dynamics for error state
- Mapping of error state to syndrome is degenerate

Error-state tracking with a Wonham filter

Ramon van Handel and HM, quant-ph/0511221

Assertion (numerically testable via comparison to SME): optimal filter for the error state can be derived as a *Wonham filter* (Wonham, 1965) for the induced Markov jump process of the error state

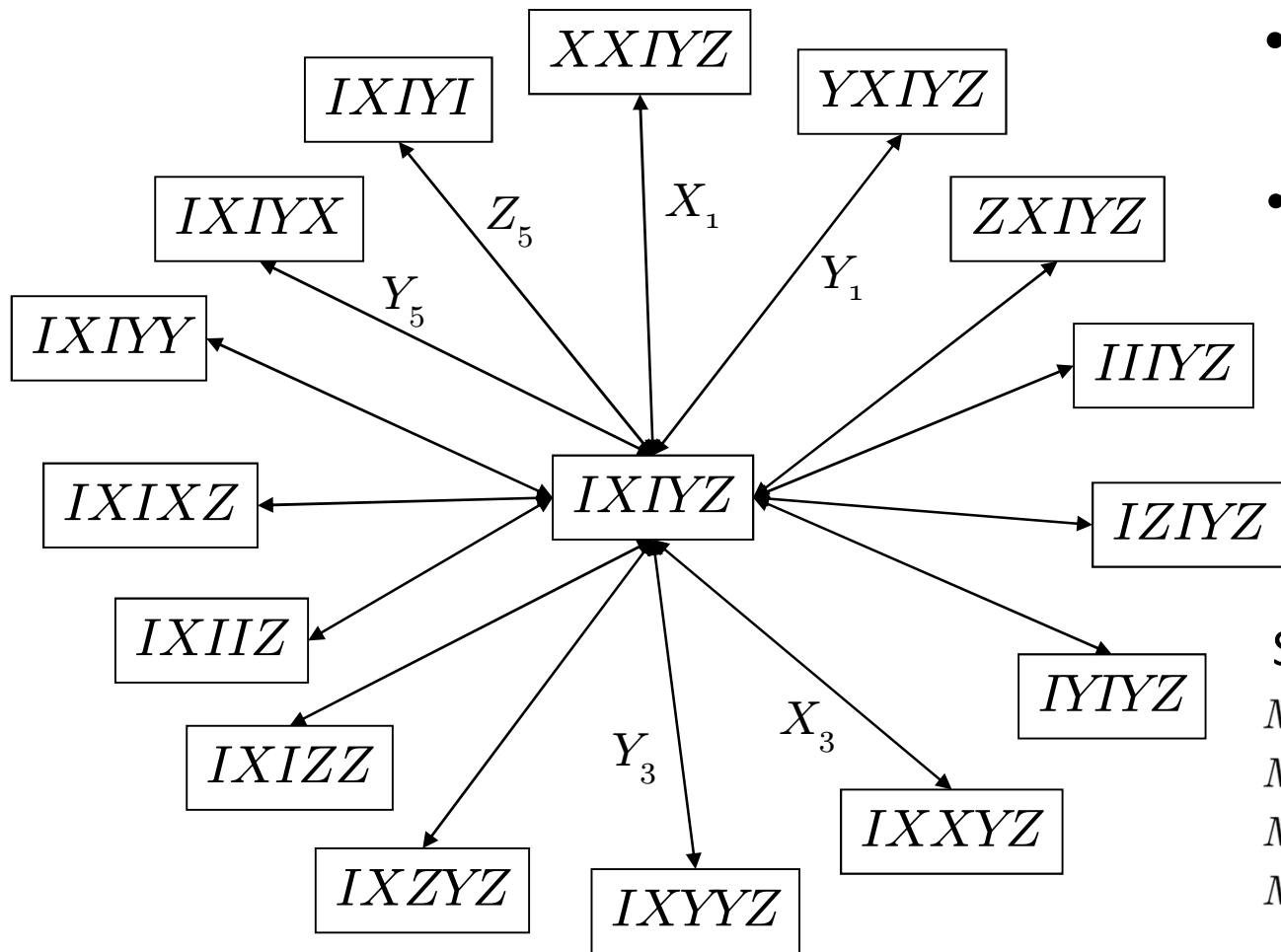


$$dp_j = (\sum_{i \neq j} \nu_{ij} p_i - \nu_j p_j) dt + \sum_k \beta_k^{-2} p_j (a_{jk} - \langle M_k \rangle) (M_k - \langle M_k \rangle)$$

- nonlinear filter, much studied in “hybrid stochastic” control theory

Filter *stability* results: P. Chigansky and R. van Handel, “Model robustness of finite state nonlinear filtering over the infinite time horizon,” Ann. Appl. Probab. 17, 688 (2007).

Error-state graph for the five-qubit code



- $4^5=1024$ error states (I, X, Y, Z for each qubit)
- $3 \times 5=15$ edges/state (X, Y, Z for each qubit)

Stabilizer generators:
 $M_1 = X \otimes Z \otimes Z \otimes X \otimes I$
 $M_2 = I \otimes X \otimes Z \otimes Z \otimes X$
 $M_3 = X \otimes I \otimes X \otimes Z \otimes Z$
 $M_4 = Z \otimes X \otimes I \otimes X \otimes Z$

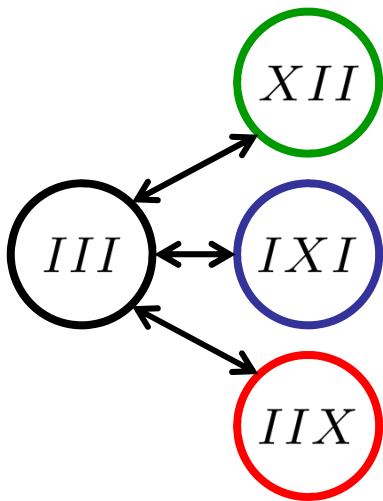
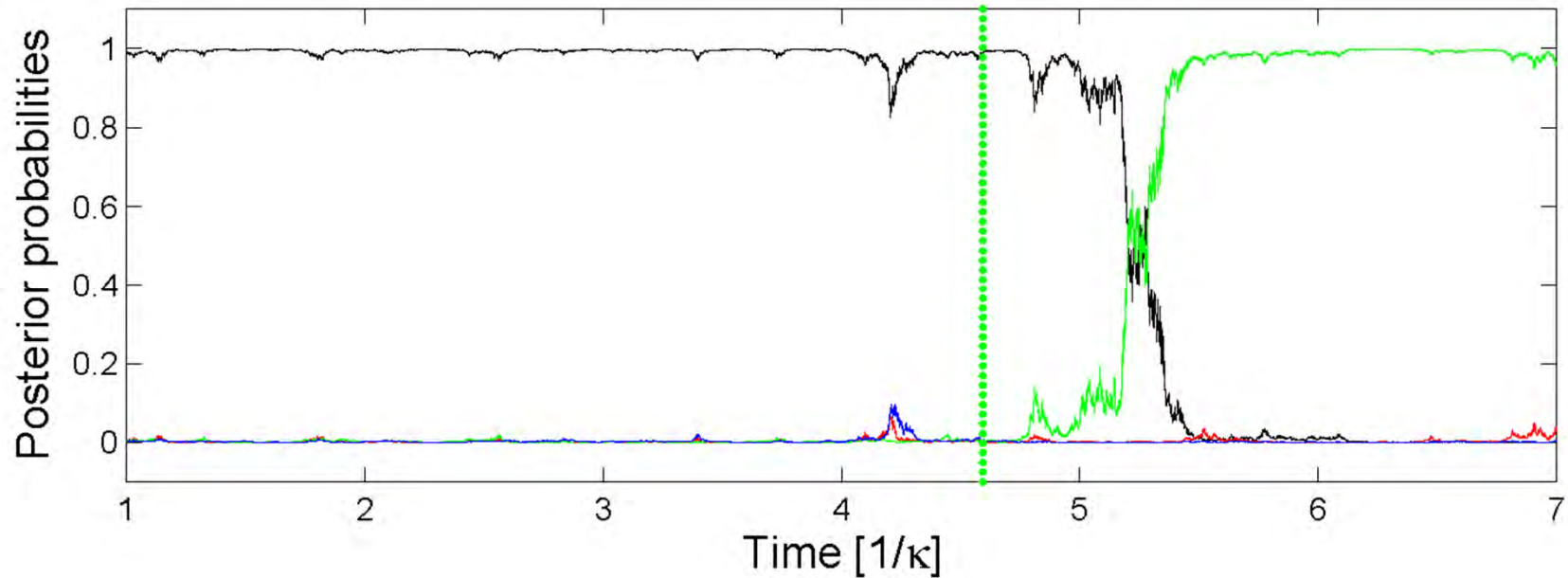
Stochastic Master Equation:

$$\kappa = 1, \quad dY_j = 2\langle M_j \rangle dt + dW_j$$

$$d\rho = \sum_{i=1}^5 \gamma(\sigma_i^\alpha \rho \sigma_i^\alpha - \rho) dt + \sum_{j=1}^4 \left\{ (M_j \rho M_j - \rho) dt + (M_j \rho + \rho M_j - 2\langle M_j \rangle) dW_j \right\}$$

$\alpha \in \{x, y, z\}$

Jump dynamics of the error state



Continuous syndrome measurement localizes the error state; bit-flip decoherence induces jump-like transitions

Finite measurement strength/sensitivity gives rise to detection delay and quiescent fluctuations

Purity of conditional distribution for the error state

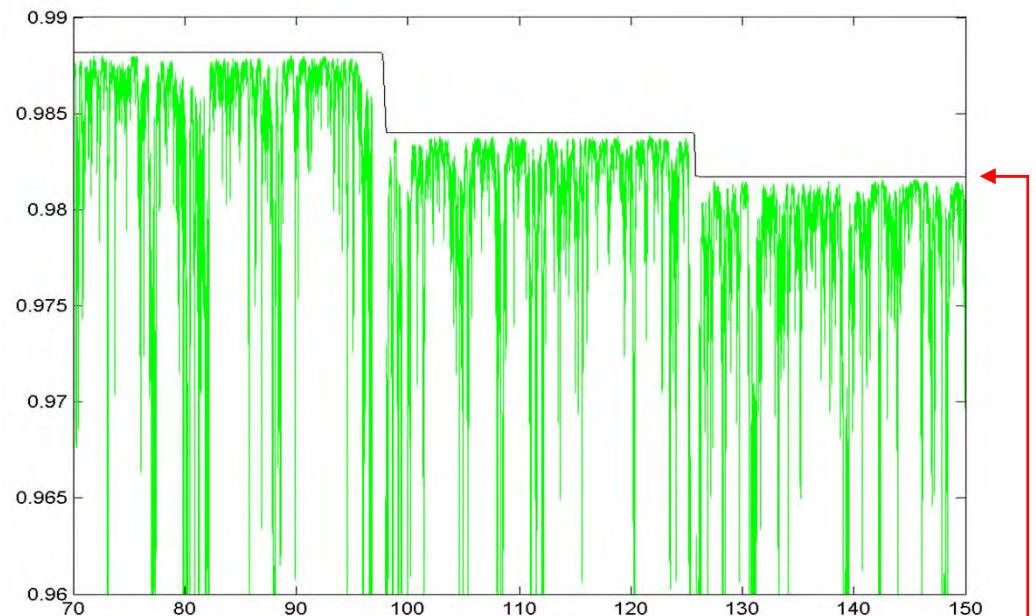
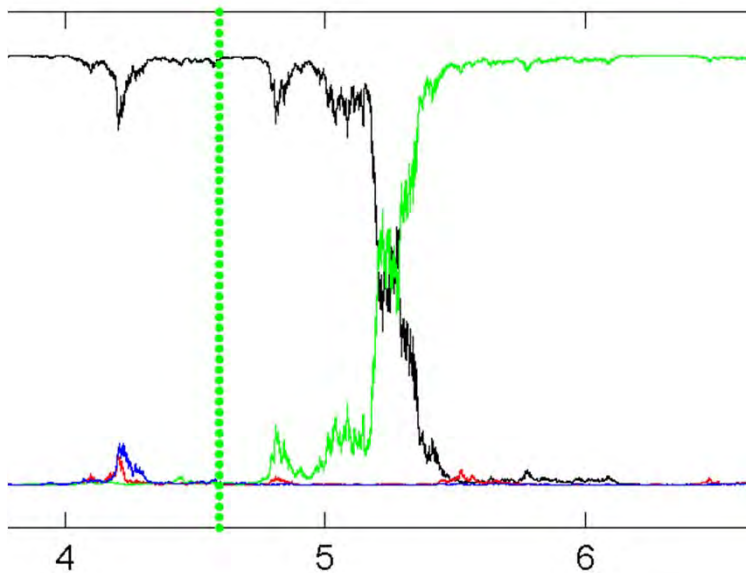
maximize $\langle \psi_{T+\varepsilon} | \psi_0 \rangle$

“know E as well as possible”



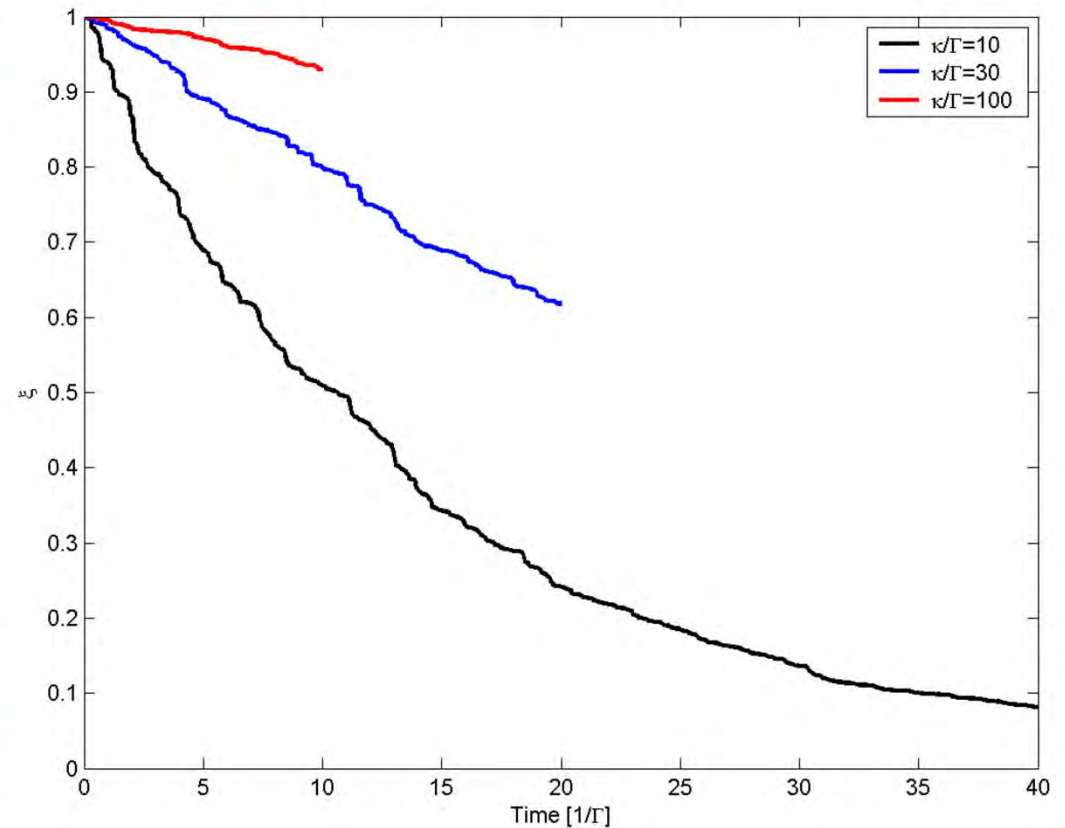
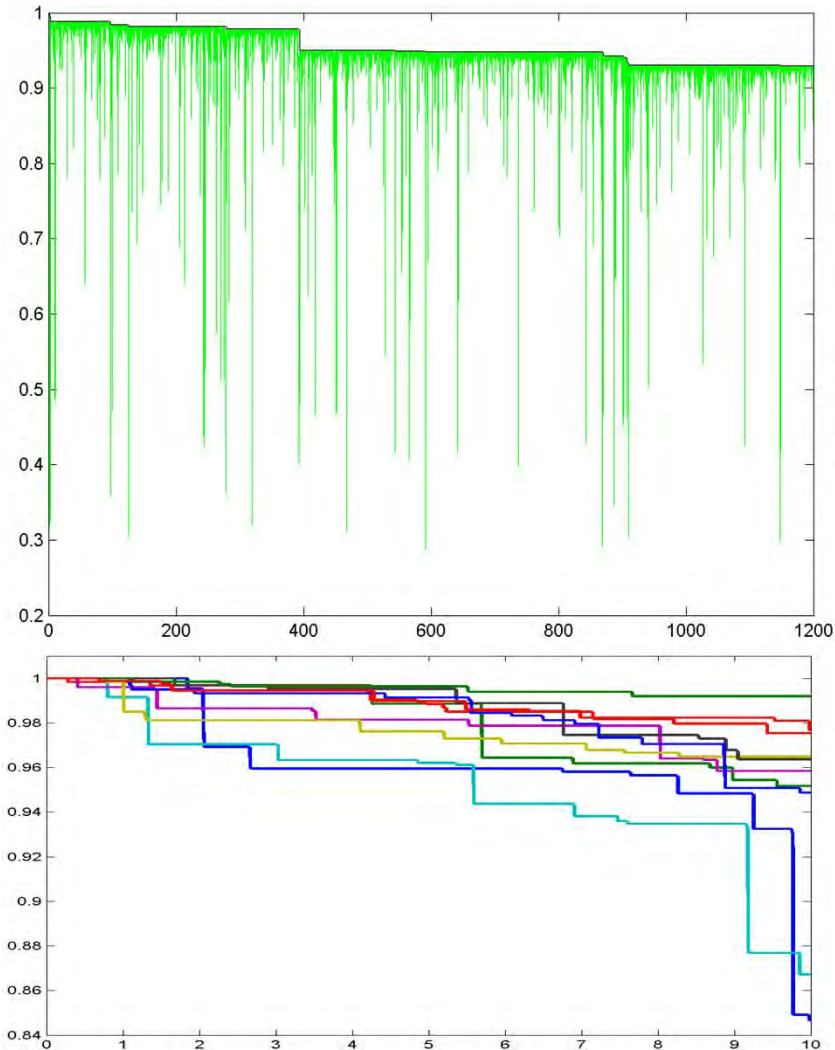
entropy or purity of conditional distribution for the error state

- If we know the error state with certainty we can recover perfectly
- Largest conditional probability $\max_i\{p_i\}$ directly related to decoded fidelity
- Purity lost with time because of multiple errors within detection delay
- Purity not monotonic because of transitions and quiescent fluctuations



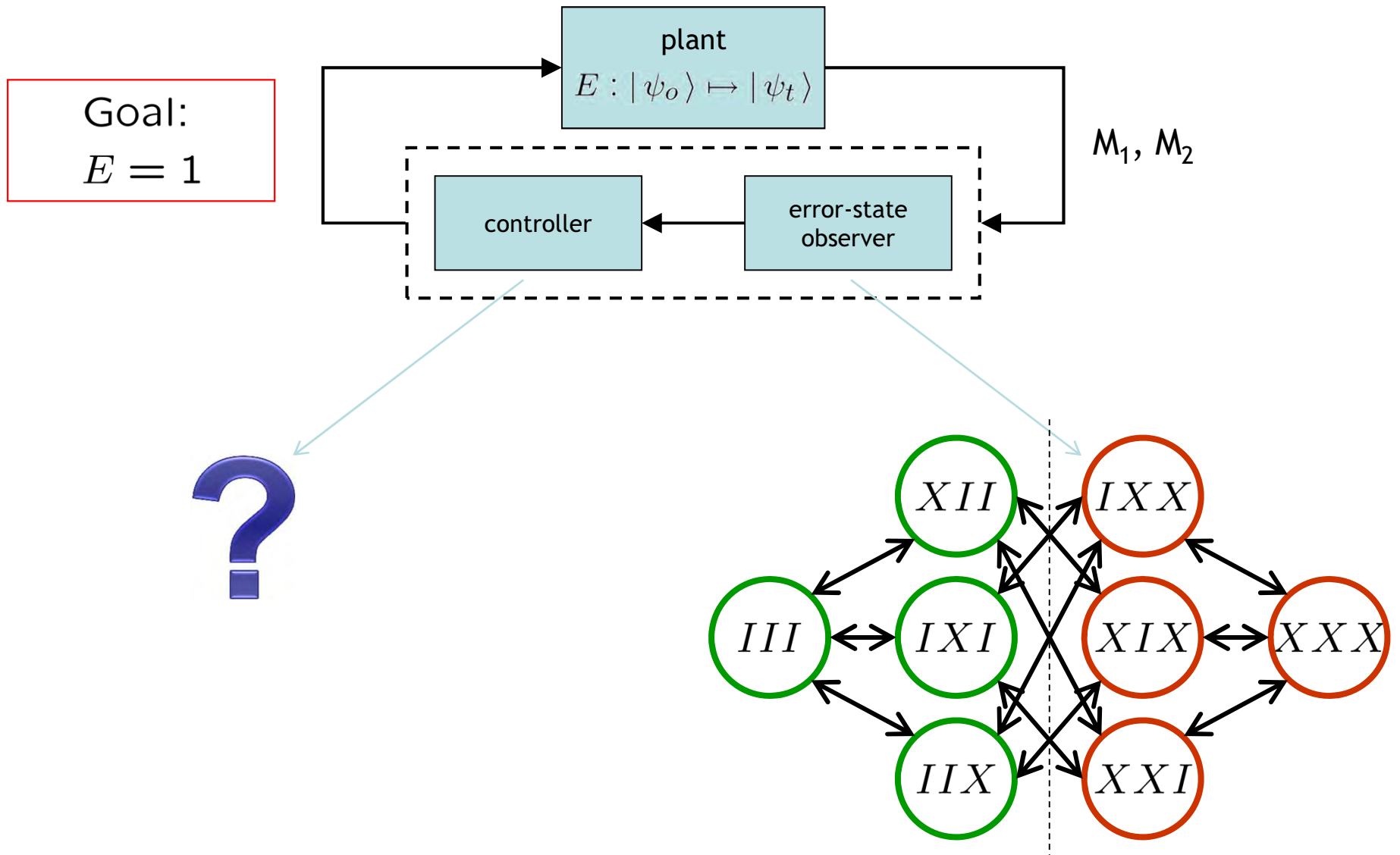
$\xi \equiv \max_i\{p_i / \Pr[\vec{a}_i]\}$ “holds” future supremum of $\max_i\{p_i\}$

Simulation results for symmetric Pauli decoherence



- Protocol: when data is “recalled,” wait for $\max_i\{p_i\}$ to approach ξ and then decode
- Further work required to derive optimal decision policy

Quantum memory with separated control strategy



$$dp_j = (\sum_{i \neq j} \nu_{ij} p_i - \nu_j p_j) dt + \sum_k \beta_k^{-2} p_j (a_{jk} - \langle M_k \rangle) (M_k - \langle M_k \rangle)$$

Correction Strategies

An **error correction strategy** consists of the following:

1. An increasing sequence of times $\{\vartheta_n\}$ at which we correct.
2. A sequence of corrections $\{\zeta_n\}$ to perform at time n .
3. **Constraint:** ϑ_n and ζ_n may depend on the syndrome observations but only in a *causal* manner (i.e., the decision to correct at a certain time may only depend on the past observation history).

Queueing model: we do not know in advance when the memory will be accessed, so we presume that it will be read out at a *random* time τ .

The **cost** of a correction strategy balances our conflicting goals:

$$J_C[\{\vartheta_n, \zeta_n\}] = \mathbf{P}[\text{Wrong syndrome at time } \tau] + C \mathbf{E}[\#\{n : \vartheta_n \leq \tau\}].$$

If $C > 0$ is large, then we give more weight to minimizing the total number of corrections. When $C > 0$ is small, we give more weight to being in the correct syndrome at the readout time.

Optimal Control

Optimal Control Problem

Given a fixed choice for the tradeoff parameter C , find an error correction strategy $\{\vartheta_n^*, \zeta_n^*\}$ which minimizes $J_C[\{\vartheta_n, \zeta_n\}]$.

Can be solved using *quantum filtering* and *dynamic programming*.

What does the optimal strategy look like? Separates into several steps:

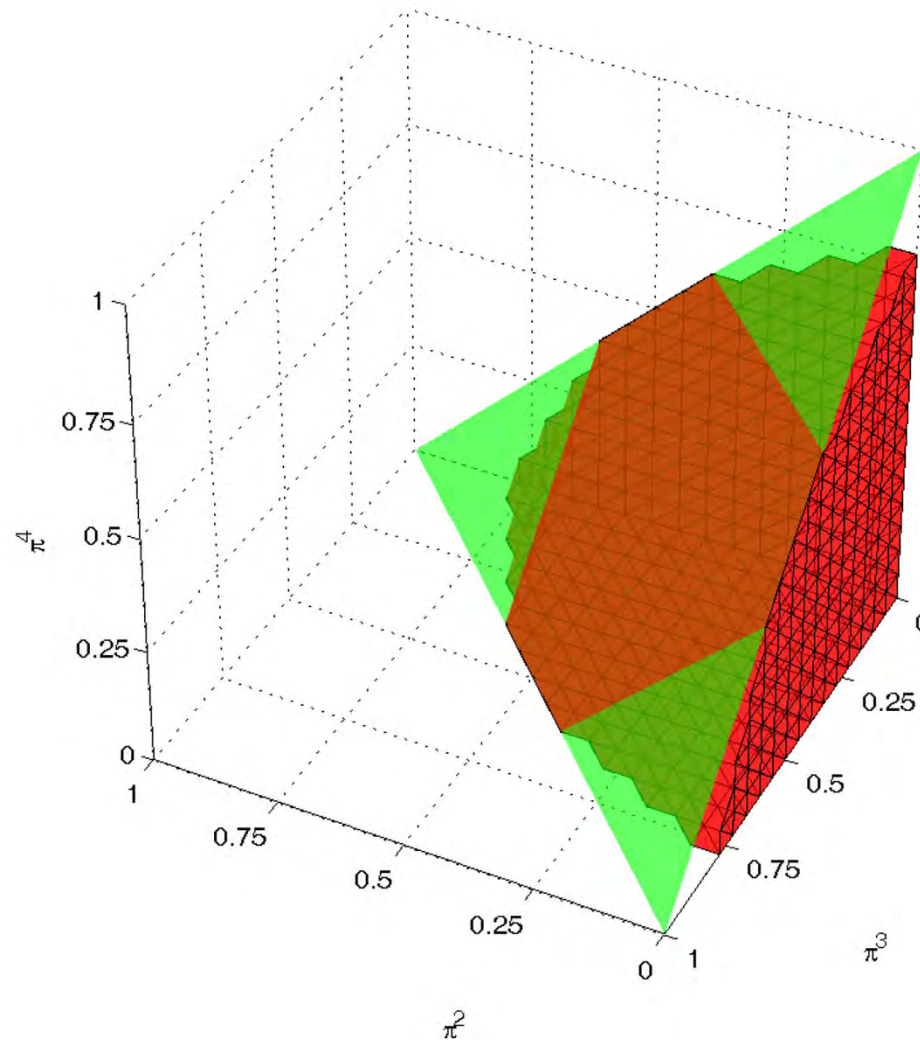
1. First, the syndrome observations are **filtered**. The filter computes the conditional probabilities π_t^i of being in the i th syndrome at time t .
2. The space of all probabilities Π is partitioned into one **continuation region** Π_0 and a **correction region** Π_j for each possible correction.
3. The **optimal strategy**: we do nothing as long as $\pi_t \in \Pi_0$. As soon as π_t enters one of Π_j , $j \geq 1$ we perform the corresponding correction. This brings us back to Π_0 , and the procedure repeats.

Numerical Solution

Ramon van Handel

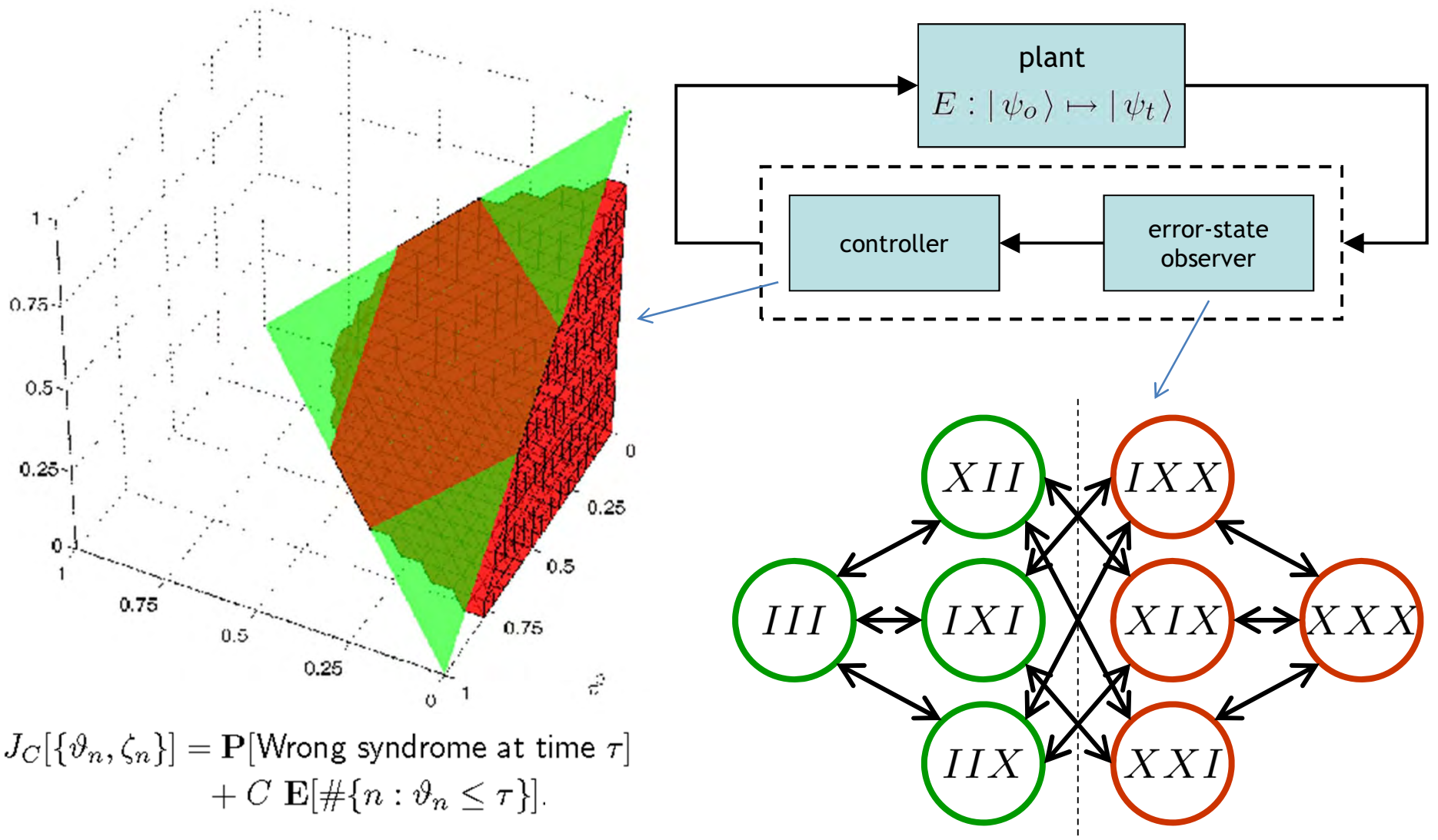
Finding the optimal strategy comes down to computing the continuation and correction regions Π_0, Π_j . This can be done numerically.

A simple three qubit code example (red region is Π_0):



Quantum memory with separated control strategy

HM (and R. van Handel), New J. Phys. **11**, 105044 (2009)

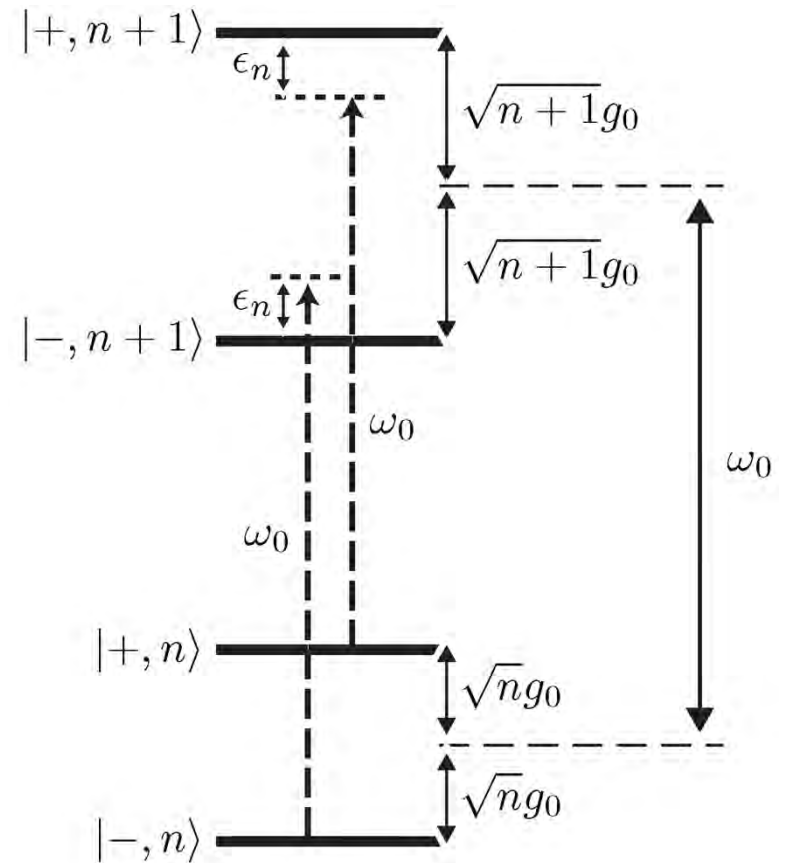
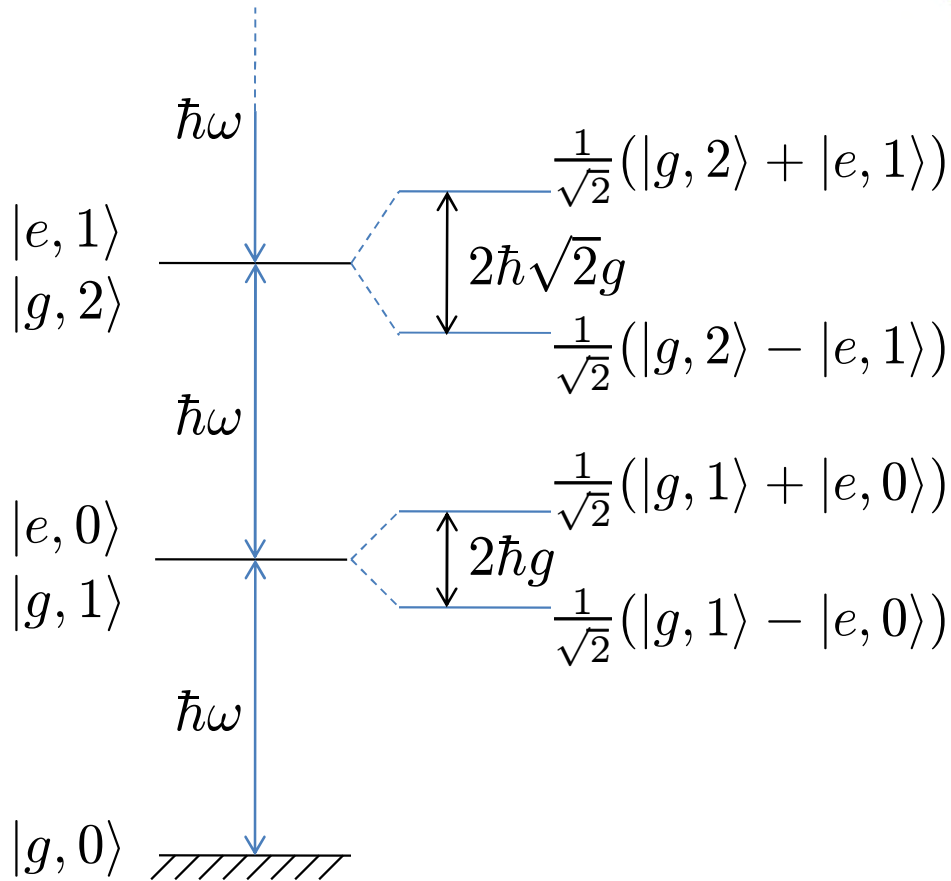
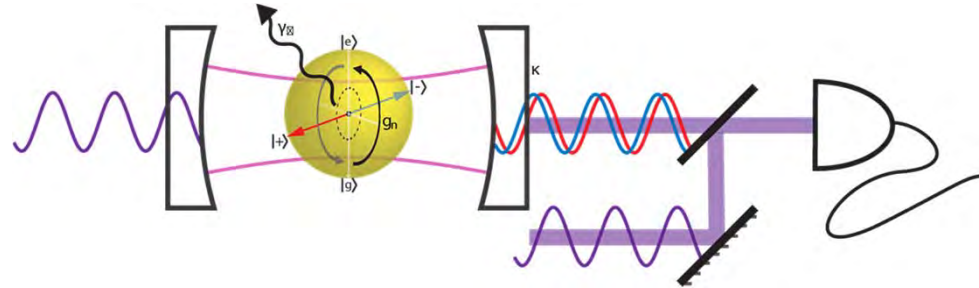


$$J_C[\{\vartheta_n, \zeta_n\}] = \mathbf{P}[\text{Wrong syndrome at time } \tau] + C \mathbf{E}[\#\{n : \vartheta_n \leq \tau\}].$$

$$dp_j = (\sum_{i \neq j} \nu_{ij} p_i - \nu_j p_j) dt + \sum_k \beta_k^{-2} p_j (a_{jk} - \langle M_k \rangle) (M_k - \langle M_k \rangle)$$

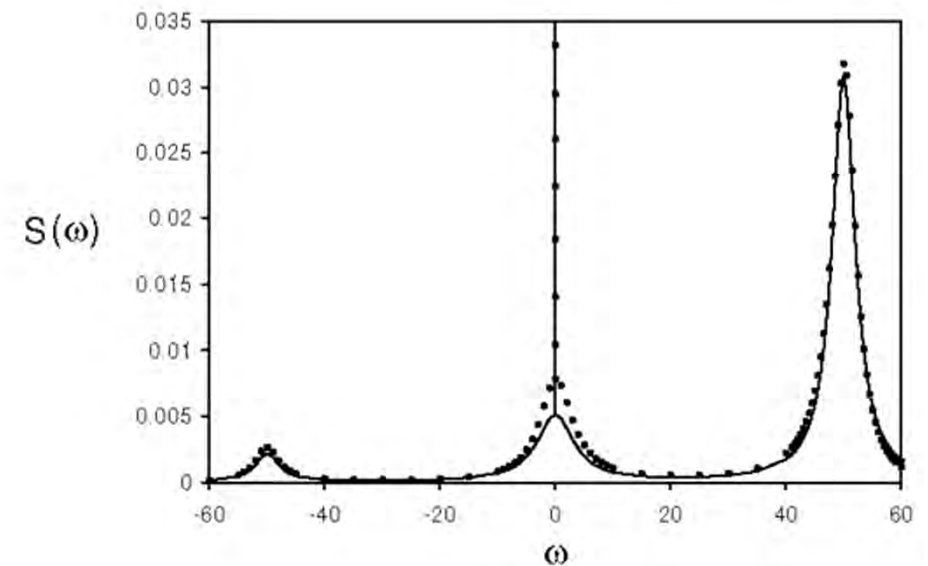
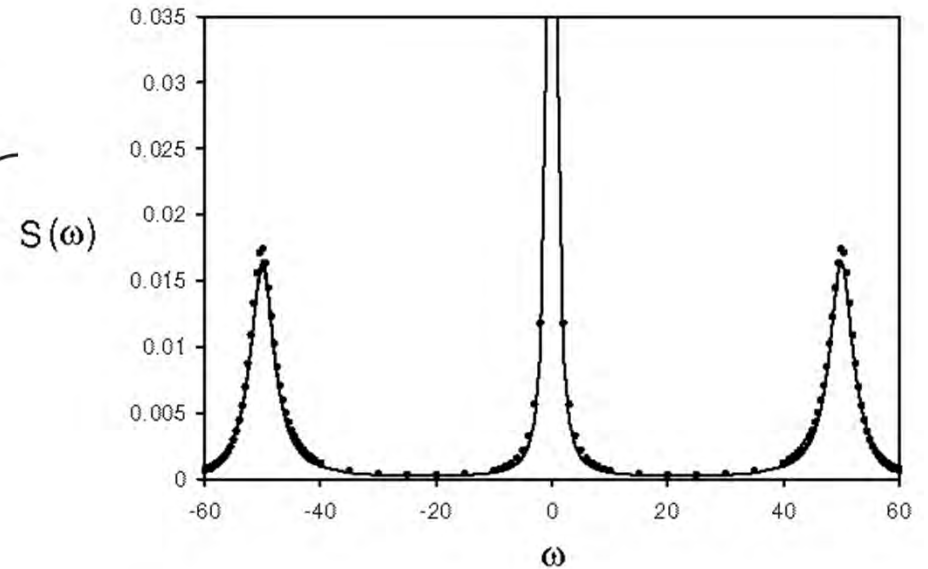
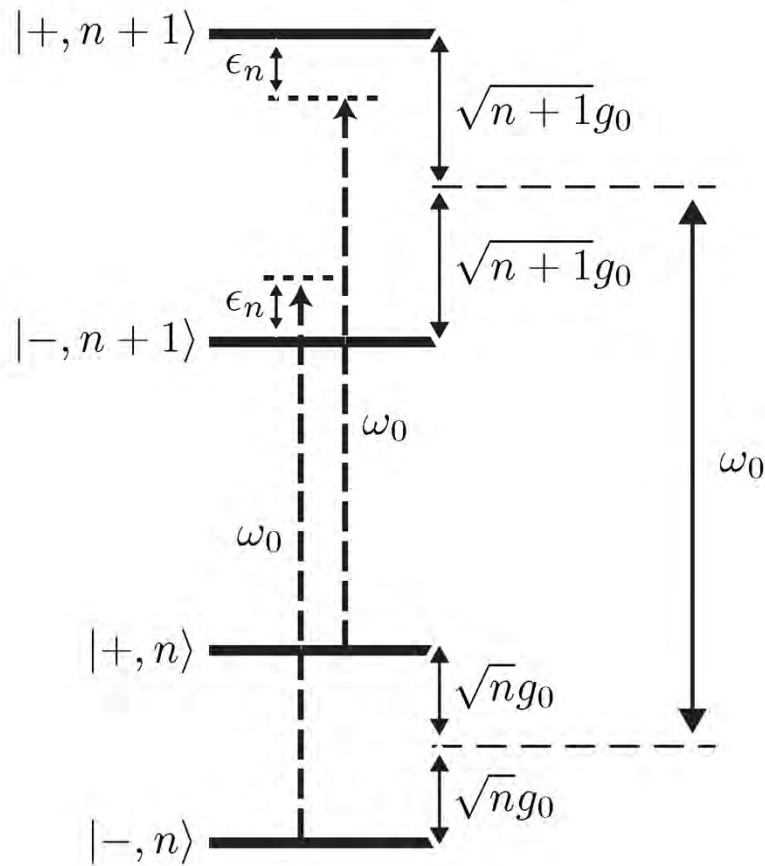
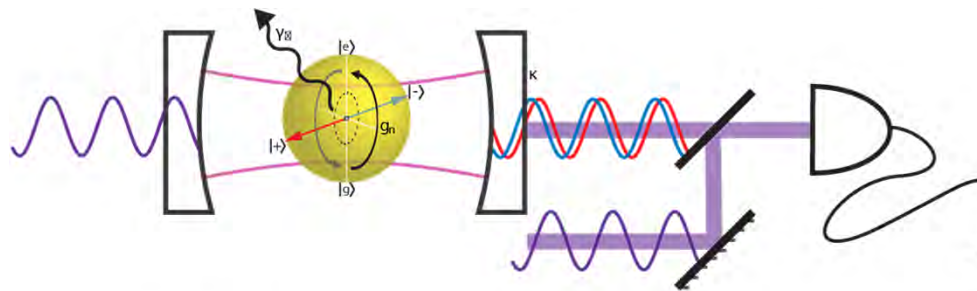
Spontaneous phase switching in cavity QED

P. Alsing and H. J. Carmichael, Quantum Opt. **3**, 13 (1991)



Feedback control: the Mollow doublet

J. E. Reiner, H. M. Wiseman and HM, Phys. Rev. A **67**, 042106 (2003)



'Retroactive' quantum jumps

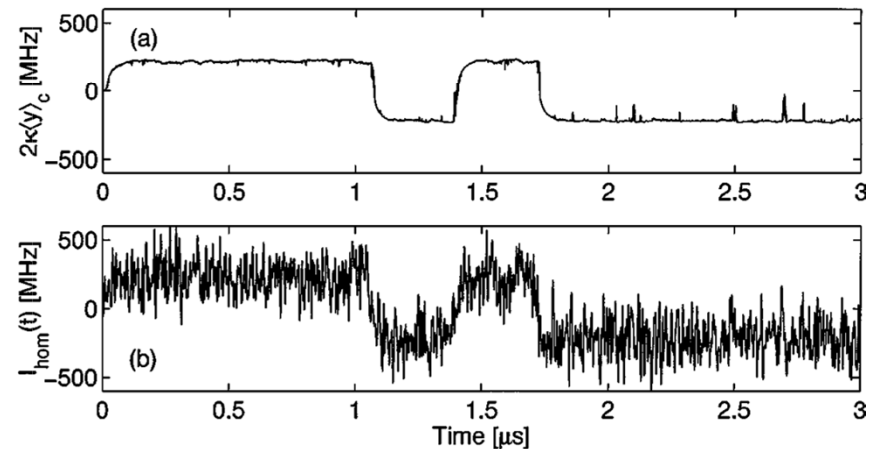
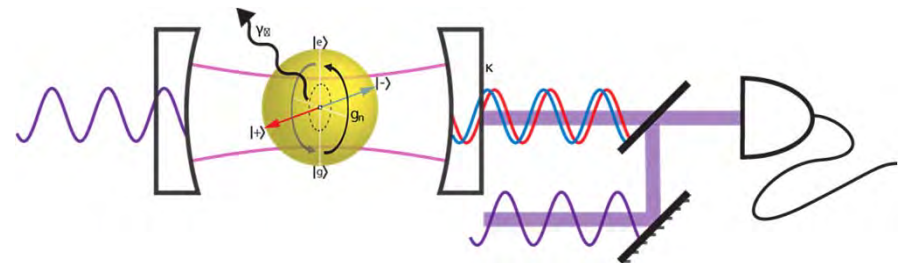
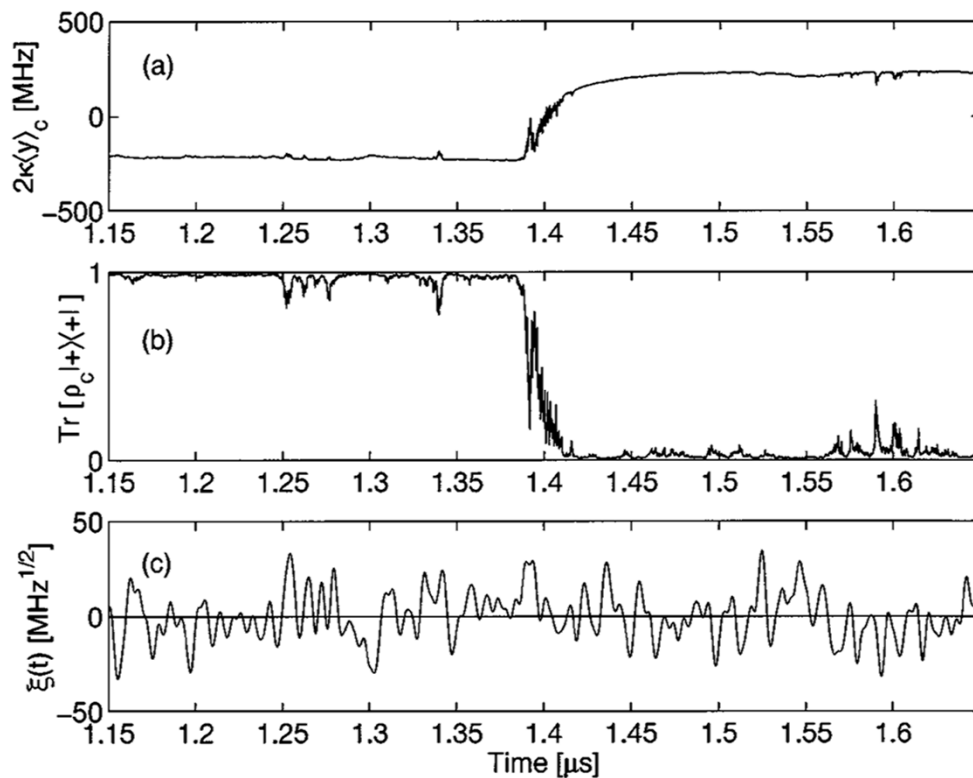
P. Alsing and H. J. Carmichael, Quantum Opt. **3**, 13 (1991)

HM and H. M. Wiseman, Phys. Rev. Lett. **81**, 4620 (1998)

$$d\rho = \mathcal{L}\rho dt + i\sqrt{2\kappa\eta} \left\{ a\rho - \rho a^\dagger - \text{Tr} [\rho (a - a^\dagger)] \right\} dW_t$$

$$I_{\text{hom}}(t)dt = 2\eta \text{Tr} [(-ia + ia^\dagger)\rho] dt + \sqrt{2\kappa\eta} dW_t$$

$$dW_t \equiv \left\{ I_{\text{hom}}(t)dt - 2\eta \text{Tr} [(-ia + ia^\dagger)\rho] dt \right\} / \sqrt{2\kappa\eta}$$

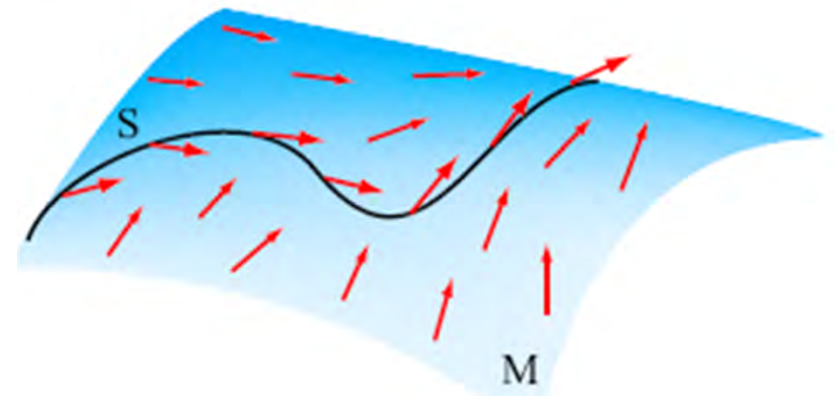
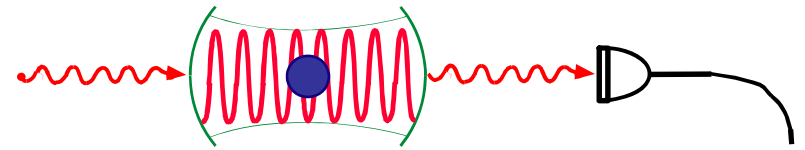
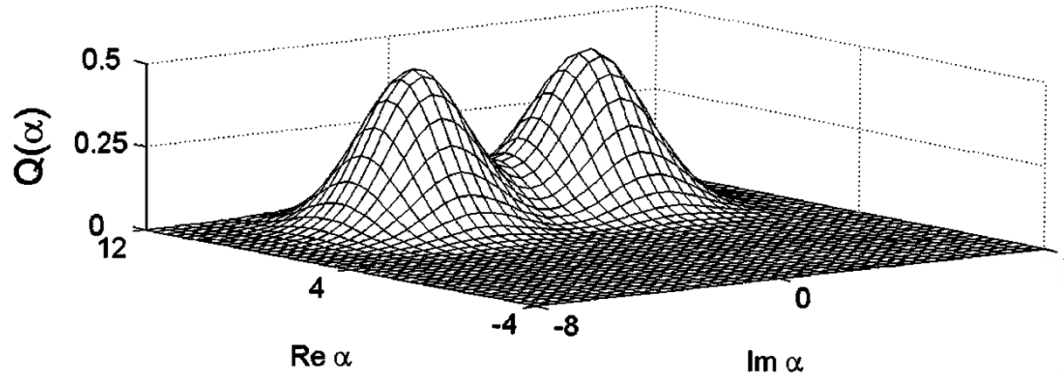


Phase bistability and quantum filter projection

P. Alsing and H. J. Carmichael, Quantum Opt. **3**, 13 (1991)

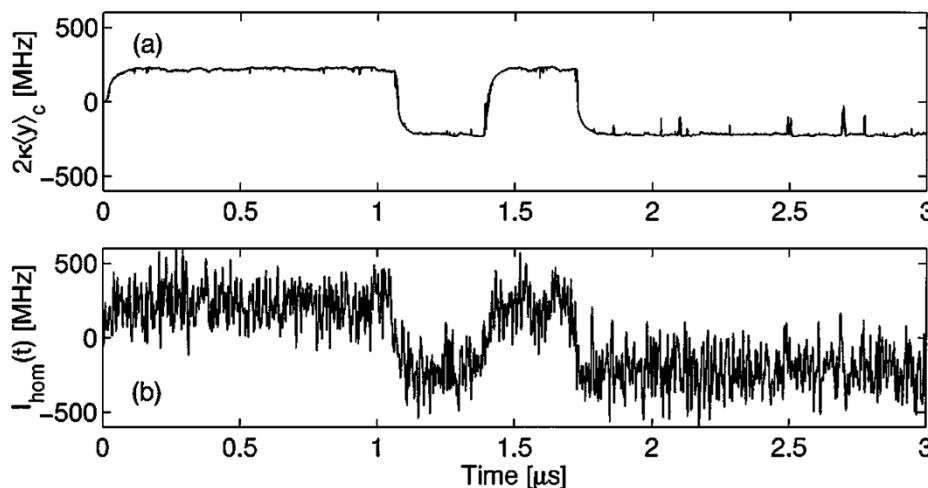
Ramon van Handel and HM, J. Opt. B: Quantum Semiclass. Opt. **7**, S226 (2005)

H. Mabuchi, Phys. Rev. A **78**, 015801, (2008)



$$d\rho = \mathcal{L}\rho dt + i\sqrt{2\kappa\eta} \left\{ a\rho - \rho a^\dagger - \text{Tr}[\rho(a - a^\dagger)] \right\} dW_t$$

$$I_{\text{hom}}(t) = 2\eta \text{Tr} [(-ia + ia^\dagger)\rho] + \sqrt{2\kappa\eta} dW_t/dt$$



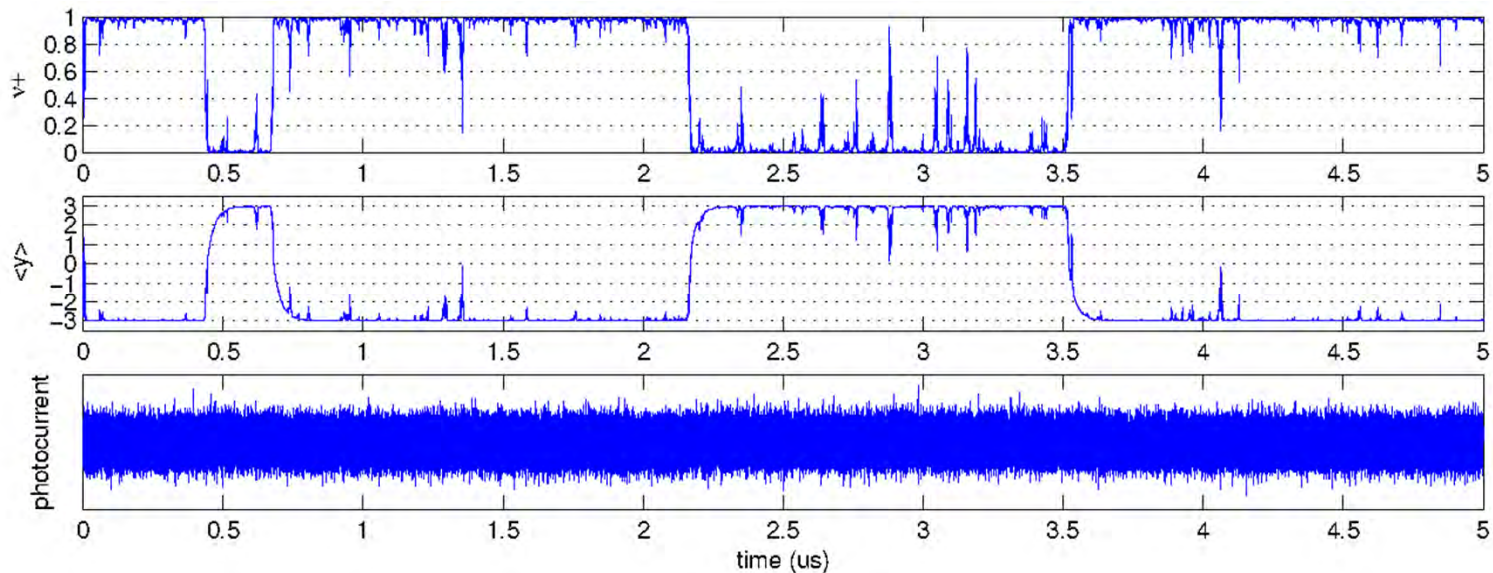
- parameterize sub-manifold of states
- intuition: two coupled “line segments”
- project stochastic equations of motion
- *e.g.*, Hilbert-Schmidt inner product

Bi-Gaussian approximate filter

Ramon van Handel and HM, J. Opt. B: Quantum Semiclass. Opt. **7**, S226 (2005)

Physical intuition motivates Gaussian *ansatz*;
restriction by geometric methods (*D. Brigo et al., M. H. Vellekoop and J. M. C. Clark, ...*)

$$\begin{aligned} d\tilde{\nu}_t^+ &= -\gamma_{\perp}(\tilde{\nu}_t^+ - \frac{1}{2})dt + \sqrt{2\kappa\eta}\tilde{\nu}_t^+(1 - \tilde{\nu}_t^+)(\mu_t^+ - \mu_t^-)(dY_t - \sqrt{2\kappa\eta}(\mu_t^+\tilde{\nu}_t^+ + \mu_t^-(1 - \tilde{\nu}_t^+))dt) \\ \frac{d\mu_t^+}{dt} &= -g - \kappa\mu_t^+ + \frac{\gamma_{\perp}}{2}\frac{1 - \tilde{\nu}_t^+}{\tilde{\nu}_t^+}(\mu_t^- - \mu_t^+) \\ \frac{d\mu_t^-}{dt} &= +g - \kappa\mu_t^- + \frac{\gamma_{\perp}}{2}\frac{\tilde{\nu}_t^+}{1 - \tilde{\nu}_t^+}(\mu_t^+ - \mu_t^-) \end{aligned}$$



accomplishes $\sim 10^5 \rightarrow 1$ reduction, but relies on knowing sub-manifold