

Perturbation theory of non-demolition measurements

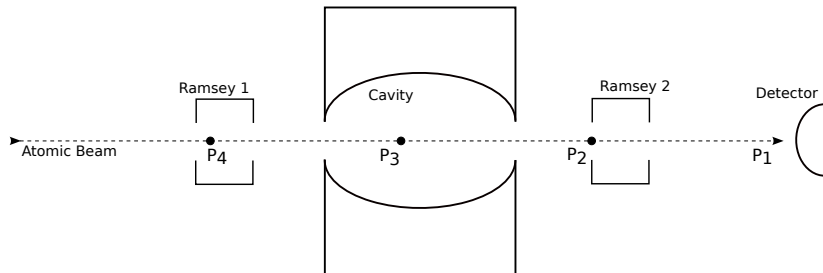
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Based on Two Articles

- ▶ M Ballesteros, MF, J Fröhlich and B Schubnel: "Indirect acquisition of information in quantum mechanics." JSP 162 (2016)
- ▶ M Ballesteros, N Crawford, MF, J Fröhlich and B Schubnel: "Perturbation theory of non-demolition measurements." in preparation

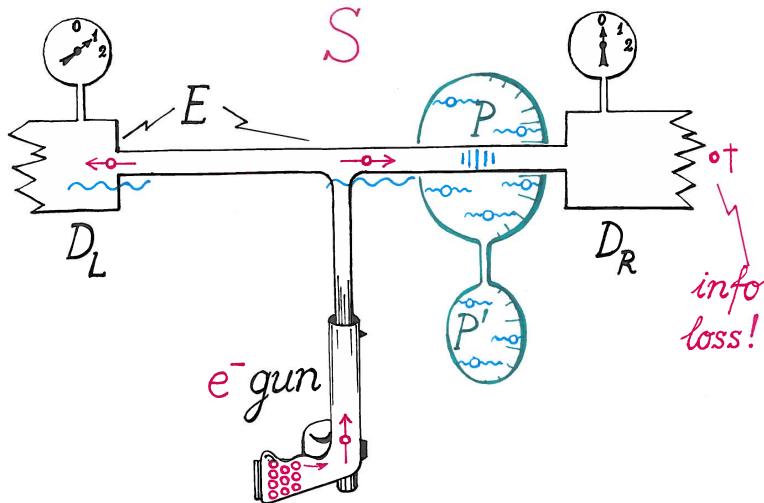
Haroche's experiment



Phenomena: After a long sequence of measurement results ξ_1, ξ_2, \dots the state of light in the cavity is close to a photon number state.

[Haroche et. al. Nature 2007, Siddiqi et. al. Nature 2013]

Electron gun



Mathematical Description

Jump operators V_ξ , $\xi \in \sigma$, with a normalization $\sum_{\xi \in \sigma} V_\xi^* V_\xi = 1$.
Probability to measure ξ is $\text{Tr}(V_\xi^* V_\xi \rho)$ and the state changes as

$$\rho \rightarrow \frac{V_\xi \rho V_\xi^*}{\text{Tr}(V_\xi \rho V_\xi^*)}.$$

For an infinity history $\underline{\xi} = \xi_1, \xi_2, \dots$ we put $V^{(n)}(\underline{\xi}) = V_{\xi_1} \dots V_{\xi_n}$,
the probability of a finite history ξ_1, \dots, ξ_n is then

$$\mathbb{P}_\rho(\xi_1, \dots, \xi_n) = \text{Tr}((V^{(n)}(\underline{\xi}))^* V^{(n)}(\underline{\xi}) \rho)$$

and the state evolves to

$$\rho_n(\underline{\xi}) = \frac{V^{(n)}(\underline{\xi}) \rho (V^{(n)}(\underline{\xi}))^*}{\text{Tr}(V^{(n)}(\underline{\xi}) \rho (V^{(n)}(\underline{\xi}))^*)}.$$

Remarks

1. Products of i.i.d. random matrices studied by [Furstenberg, Kesten Annals of Stat. 1960] and many others with applications to 1D random Schrödinger e.g [Bougerol, Lacroix (1985)]
Measurement in Quantum Mechanics also leads to study of product of matrices but with non i.i.d. measure.
2. The setting is an example of a finitely correlated state [Fannes, Nachtergaele, and Werner CMP 144].

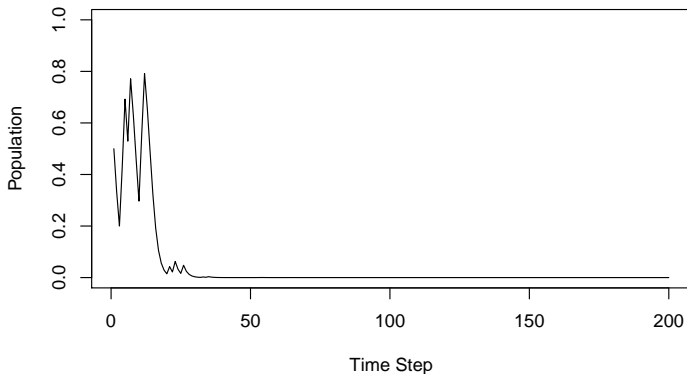
Example $\varepsilon = 0$

For a two level system and $\sigma = \{e, g\}$ we put

$$V_e = e^{-i\varepsilon\sigma_1} \begin{pmatrix} \sqrt{p} & 0 \\ 0 & \sqrt{q} \end{pmatrix}, \quad V_g = e^{-i\varepsilon\sigma_1} \begin{pmatrix} \sqrt{1-p} & 0 \\ 0 & \sqrt{1-q} \end{pmatrix}$$

For $\varepsilon = 0$ the population of $\mathcal{N} = \sigma_z$ approaches an eigenstate,

Population Tracking



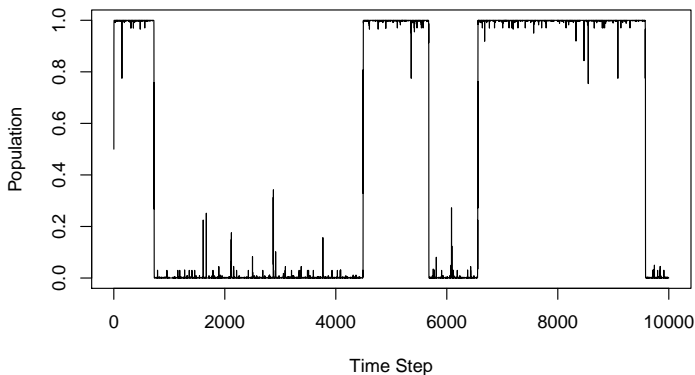
Example $\varepsilon \neq 0$

For a two level system and $\sigma = \{e, g\}$ we put

$$V_e = e^{-i\varepsilon\sigma_1} \begin{pmatrix} \sqrt{p} & 0 \\ 0 & \sqrt{q} \end{pmatrix}, \quad V_g = e^{-i\varepsilon\sigma_1} \begin{pmatrix} \sqrt{1-p} & 0 \\ 0 & \sqrt{1-q} \end{pmatrix}$$

For $\varepsilon \neq 0$ the population of $\mathcal{N} = \sigma_z$ jumps between eigenstates,

Population Tracking



Non-demolition case and its perturbations

For an observable \mathcal{N} and a Hamiltonian H with $[H, \mathcal{N}] \neq 0$ we put

$$V_{\xi}^{(\varepsilon)} = e^{-i\varepsilon H} V_{\xi}(\mathcal{N})$$

for some complex functions $V_{\xi}(\cdot)$.

In the case $\varepsilon = 0$ we have $[V_{\xi}, V_{\xi'}] = 0$ and

$$\mathbb{P}_{\rho}(\xi_1, \dots, \xi_n) = \int_{\sigma(\mathcal{N})} |V_{\xi_1}(\nu)|^2 \dots |V_{\xi_n}(\nu)|^2 d\lambda_{\rho}(\nu),$$

where λ_{ρ} is a spectral measure of \mathcal{N} . Interpreting ν as an unknown we define maximum likelihood estimate

$$\hat{\mathcal{N}}_k(\underline{\xi}) = \operatorname{argmax}_{\nu \in \sigma(\mathcal{N})} l_k(\nu | \underline{\xi}), \quad l_k(\nu | \underline{\xi}) := \frac{1}{k} \sum_{j=1}^k \log |V_{\xi_j}(\nu)|^2.$$

Non-Demolition Case, $\varepsilon = 0$

For set $N \in \sigma(\mathcal{N})$ let $\Pi(N)$ be the associated spectral projection and $S(\nu|N) = \inf_{\nu' \in N} \sum_{\xi \in \sigma} |V_\xi(\nu)|^2 [I(\nu|\xi) - I(\nu'|\xi)]$.

Theorem (Law of Large Numbers)

Suppose $\nu \rightarrow V_\xi(\nu)$ is injective and $V_\xi(\cdot)$ is continuous for all $\xi \in \sigma$, then the maximum likelihood estimator $\hat{\mathcal{N}}_k$ converges almost surely to a random variable $\hat{\mathcal{N}}_\infty$. For any Borel set $N \subset \sigma(\mathcal{N})$,

$$\mathbb{P}_\rho(\underline{\xi} : \lim_{k \rightarrow \infty} \hat{\mathcal{N}}_k \in N) = \text{Tr}(\Pi(N)\rho).$$

Moreover if N is a closed subset of $\sigma(\mathcal{N})$ contained in the support of the measure λ_ρ then we have

$$- \lim_{k \rightarrow \infty} \frac{1}{k} \log \text{Tr}(\Pi(N)\rho_k) = S(\hat{\mathcal{N}}_\infty|N), \quad \mathbb{P}_\rho - \text{almost surely.}$$

Non-Demolition Case - references

When spectrum of \mathcal{N} is discrete the Law of Large Numbers was proved in

- ▶ Maassen, Kümmerner 2006
- ▶ Bauer, Bernard PRA 2011
- ▶ Mabuchi et. al. IEEE 2004

The large deviation theory

- ▶ Bauer, Benoist, Bernard AHP 2013

Demolition Case

We make measurement times t_1, t_2, \dots of ξ_1, ξ_2, \dots random and distributed by Poisson distribution N_t , then the evolution is

$$\tau_\varepsilon^{(s)}(\underline{t}, \underline{\xi})\rho = e^{-i\varepsilon H(s-t_{N_s})} V_{\xi_{N_s}} \dots e^{-i\varepsilon H t_1} \rho e^{i\varepsilon H t_1} V_{\xi_1} \dots e^{i\varepsilon H(s-t_{N_s})}.$$

This is an unravelling of Lindblad evolution,

$$\mathbb{E}[\tau_\varepsilon^{(s)}] = \exp(s\mathcal{L}_\varepsilon), \quad \mathcal{L}_\varepsilon \rho = -i\varepsilon[H, \rho] + \sum_{\xi \in \sigma} V_\xi \rho V_\xi^* - \rho.$$

For a sampling time $T > 0$ we define

$$\hat{\mathcal{N}}_s := \operatorname{argmax}_{\nu \in \sigma(\mathcal{N})} \frac{1}{N_{s+T} - N_s} \sum_{j=N_s}^{N_{s+T}} I(\nu|\xi_j).$$

To avoid dealing with overlapping data we set $\mathcal{M}_{jT} = \hat{\mathcal{N}}_{jT}$, for $j \in \mathbb{N}$ and extend the definition of \mathcal{M}_t to all $t \geq 0$ by declaring it to be piecewise constant on the intervals $[jT, (j+1)T)$.

Demolition Case

Let $\mathcal{N} = \sum_{\nu} \nu P_{\nu}$ and define

$$\mathcal{P}\rho = \sum_{\nu \in \sigma(\mathcal{N})} P_{\nu} \rho P_{\nu}, \quad P_{\nu} = |\nu\rangle\langle\nu|.$$

We define an operator Q on the range of \mathcal{P} by

$$\varepsilon^2 Q = -\mathcal{P}\mathcal{L}_{\varepsilon}\mathcal{P}_{\perp}\mathcal{L}_0^{-1}\mathcal{P}_{\perp}\mathcal{L}_{\varepsilon}\mathcal{P}.$$

The matrix Q defines a Markov Kernel on $\sigma(\mathcal{N})$ with elements

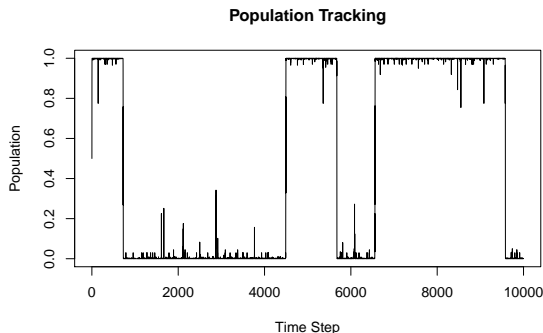
$$\mathrm{Tr}(P_{\nu'} Q P_{\nu}) = \begin{cases} \sum_{\beta \neq \nu} \frac{|\langle\beta|H|\nu\rangle|^2}{\sum_{\xi} V_{\xi}(\beta) \bar{V}_{\xi}(\nu) - 1} + \text{c.c.} & \text{for } \nu = \nu' \\ -\frac{|\langle\nu'|H|\nu\rangle|^2}{\sum_{\xi} V_{\xi}(\nu') \bar{V}_{\xi}(\nu) - 1} + \text{c.c.} & \text{for } \nu \neq \nu'. \end{cases}$$

Let Y_s be the continuous time Markov chain generated by Q started from an initial probability distribution $\pi_{\rho}(\nu) = \mathrm{Tr}(P_{\nu}\rho)$.

Demolition Case

Theorem (Distribution of Jumps)

Suppose $\nu \rightarrow V_\xi(\nu)$ is injective, and pick a positive I strictly smaller than $\min_{\nu, \nu'} S(\nu, \nu')$. Let $T = -\beta \log \varepsilon$, for some $\beta > \max\{2(1 - e^{-I})^{-1}, (1 - e^{-\frac{I}{2}})^{-1}\}$. Then under $\mathbb{P}_\rho^{(\varepsilon)}$, $\mathcal{M}_{\varepsilon^{-2}s}$ converges in law to Y_s , and the posterior density matrix $\rho_{\varepsilon^{-2}s}$ converges in law to P_{Y_s} .



Thank you for your attention!