

Quantum Repeated Measurements, Continuous Time Limit

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- **I) Discrete time Model: Quantum Repeated measurement**
 - H. Maassen: ergodic properties, purification properties.
 - L. Bouten, R. Van Handel: discrete quantum filtering
 - M. Bauer, D. Bernard, T. Benoist: large time behaviour, rate of convergence, quantum nondemolition measurement
 - P. Rouchon and all: control (direct collaboration with the serge Haroche team at LKB)
 - M. Merkli: quantum measurement of scattered particles
 - S. Attal, N. Guillotin-Plantard, C. Sabot: Central Limit Theorem for OQRW
- **II) Continuous Time model: Stochastic Schrödinger Equations, Stochastic Master Equations.**
 - Using linear stochastic master equations and change of measure (A.Barchielli, M.Gregoratti...)
 - Quantum Filtering Theory based on Quantum stochastic differential equation (V.P.Belavkin...)
 - Positive Operator Valued Measure and Instruments (E.B Davies...)
 - From discrete to continuous time model (adapting the results of S.Attal-Y.Pautrat...)

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- **III) Continuous Time Limit**

- Continuous time limit of Quantum Repeated Measurements (P., also M. Bauer, D. Bernard, T. Benoist...)

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- Estimation: work in progress with P. Rouchon And H. Amini
- Temperature (with I. Nechita and S. Attal)

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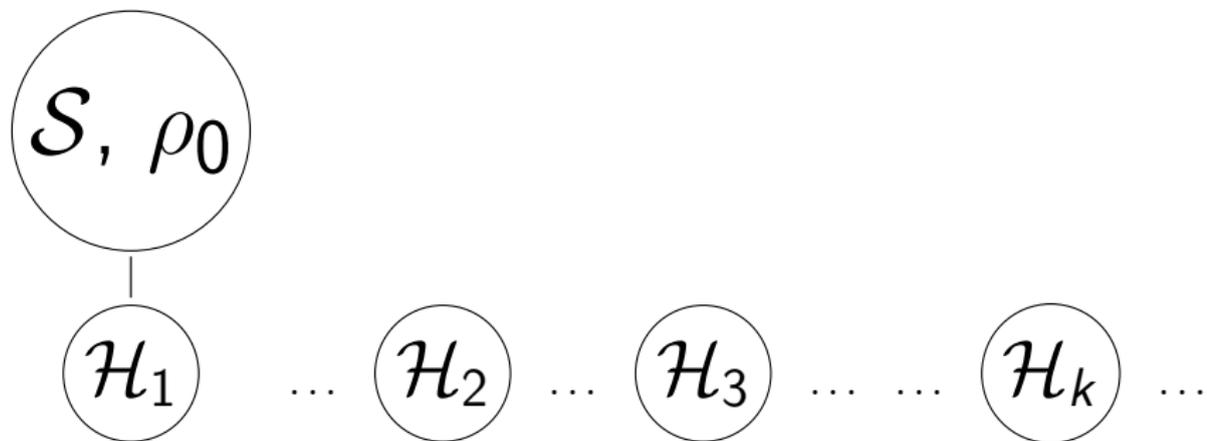
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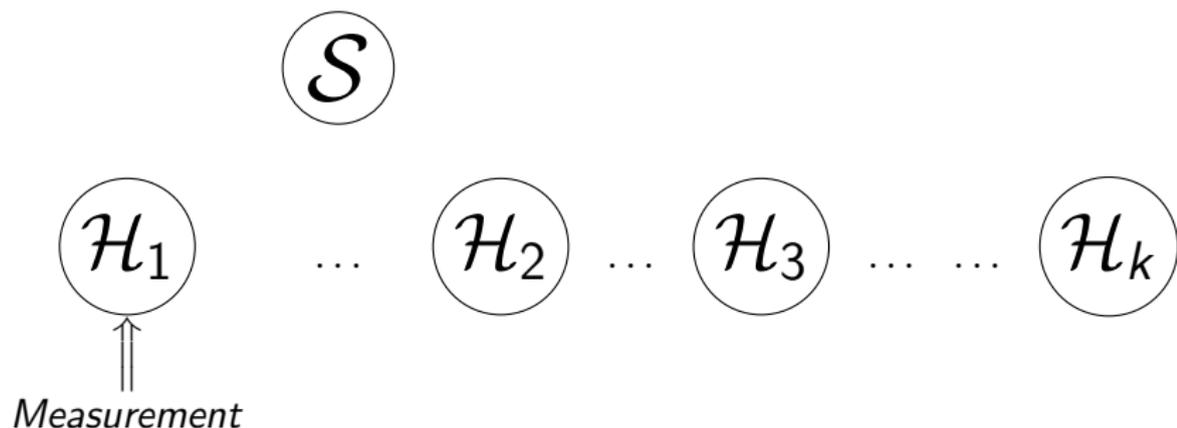
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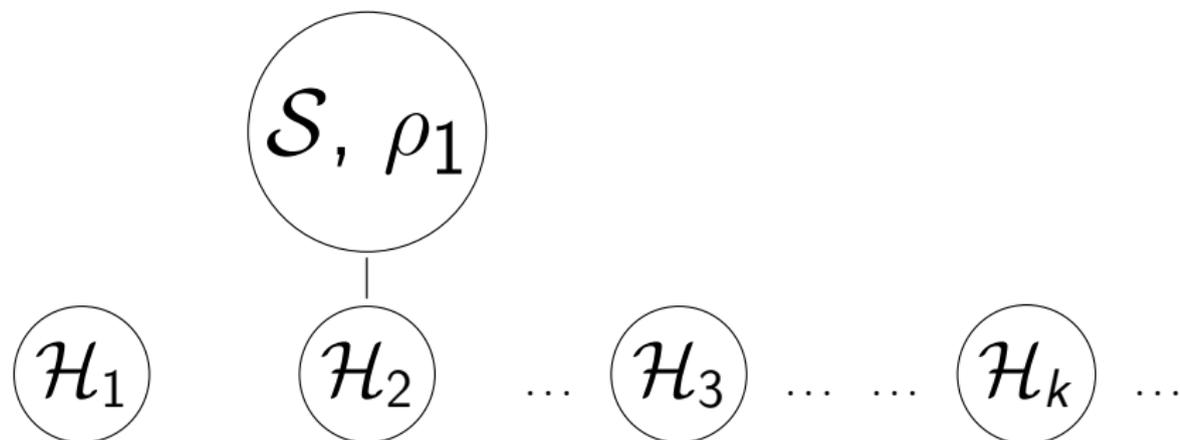
I) Discrete Time Model

QUANTUM REPEATED MEASUREMENT and DISCRETE QUANTUM TRAJECTORIES

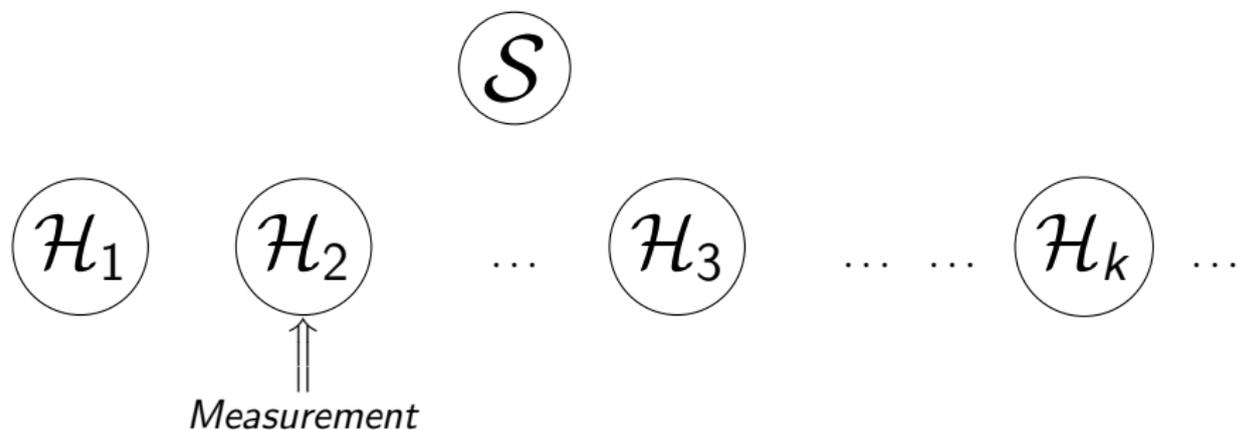


1st measurement





2nd measurement



and so on \Rightarrow discrete quantum trajectory (ρ_n)

Proposition

Let A be an observable of the form $A = \sum_{i=0}^p \lambda_i P_i$.

Then there exists a probability space (Ω, \mathcal{C}, P) , where the *the discrete quantum trajectory* (ρ_k) , describing the quantum repeated measurement of A , is a *Markov chain*.

More precisely if $\rho_k = \theta$ is a state on \mathcal{H}_0 , then ρ_{k+1} takes the values

$$\frac{\mathcal{L}_i(\theta)}{\text{Tr}[\mathcal{L}_i(\theta)]}, \quad i \in \{0, \dots, p\}$$

where $\mathcal{L}_i(\theta) = \text{Tr}_{\mathcal{H}}[(I \otimes P_i) U(\theta \otimes \beta)U^* (I \otimes P_i)]$. Each state appears with probability

$$p_i(\theta) = \text{Tr}[\mathcal{L}_i(\theta)].$$

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$$p_i(\theta) = \text{Tr}[\mathcal{L}_i(\theta)].$$

- The previous result can be summarized by writing the following evolution equation:

$$\rho_{k+1} = \sum_{i=0}^P \frac{\mathcal{L}_i(\rho_k)}{\text{Tr}[\mathcal{L}_i(\rho_k)]} \mathbf{1}_i^{k+1},$$

with $\mathcal{L}_i(\theta) = \text{Tr}_{\mathcal{H}}[(I \otimes P_i) U(\theta \otimes \beta) U^* (I \otimes P_i)]$.

- **Remark:** The operator U depends on the time interaction τ

$$U = e^{-i\tau H_{\text{tot}}}.$$

Questions

What gives the limit τ goes to 0?

What is the limit evolution?

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II) Stochastic Master Equations

Continuous Quantum Trajectories

- Let us start with model where only one measurement apparatus is concerned
- Evolution of \mathcal{H}_0 without measurement : Master Equation in Lindblad form

$$d\rho_t = L(\rho_t)dt.$$

- Effect of measurement = perturbation of this ode under the form of stochastic differential equations

Stochastic Master Equations

- Diffusive Equation

$$d\rho_t = L(\rho_t)dt + [C\rho_t + \rho_t C^* - \text{Tr}[\rho_t(C + C^*)]\rho_t]dW_t$$

- ① The process (W_t) is a standard **Brownian motion**.
- ② C is an arbitrary operator appearing in the Lindblad operator, L .
- Often this equation appears on the following form

$$d\rho_t = L(\rho_t)dt + [C\rho_t + \rho_t C^* - \text{Tr}[\rho_t(C + C^*)]\rho_t](dy_t - \text{Tr}[\rho_t(C + C^*)]dt),$$

where

$$dy_t = dW_t + \text{Tr}[\rho_t(C + C^*)]dt$$

- The process (y_t) represents the measurement process recorded by the measurement apparatus (Homodyne/Heterodyne detection in Quantum Optics).

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- Jump Equation

$$d\rho_t = L(\rho_t)dt + \left[\frac{\mathcal{J}(\rho_t)}{\text{Tr}[\mathcal{J}(\rho_t)]} - \rho_t \right] (d\tilde{N}_t - \text{Tr}[\mathcal{J}(\rho_t)] dt)$$

- ① The process (\tilde{N}_t) is a **counting process** of stochastic intensity

$$t \longrightarrow \int_0^t \text{Tr}[\mathcal{J}(\rho_s)] ds.$$

- ② $\mathcal{J}(\rho) = C\rho C^*$.

- The process (\tilde{N}_t) represents the number of photon detected up to time t by a photo detector.

- First fact: $d\mathbb{E}[\rho(t)] = L(\mathbb{E}[\rho(t)])dt$. That is $(\mathbb{E}[\rho(t)])$ reproduces the solution of the Lindblad master equation
- In the previous cases if

$$L(\rho) = -i[H, \rho] - \frac{1}{2}\{C^*C, \rho\} + C\rho C^*$$

and if at time 0 $\rho_0 = |\psi_0\rangle\langle\psi_0|$ then there exists ψ_t such that

$$\rho(t) = |\psi_t\rangle\langle\psi_t|, \forall t$$

The equation satisfied by ψ_t is called a **Stochastic Schrödinger Equation**

- In general we have $\rho(t) = \rho^*(t)$ and $\text{Tr}[\rho(t)] = 1$, then if there is a solution and if the initial condition is a density matrix then the solution is self-adjoint and of trace 1. What is very difficult to show is that the solution is positive.

Theorem

Let (Ω, \mathcal{F}, P) a probability space where (W_t) is a standard Brownian motion.

the equation

$$d\rho_t = L(\rho_t)dt + \left[C\rho_t + \rho_t C^* - \text{Tr}[\rho_t(C + C^*)]\rho_t \right] dW_t$$

admits a unique solution (ρ_t) with values in the set of states of \mathcal{H}_0 .

- Non Lipschitz coefficients

One can use a truncature method

- We show that this equation preserves the property of being a state.
- Here we can use the approximation procedure to show the positivity of the solution

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Processus de comptage (\tilde{N}_t)?

- Jump equation

$$d\rho_t = L(\rho_t)dt + \left[\frac{\mathcal{J}(\rho_t)}{\text{Tr}[\mathcal{J}(\rho_t)]} - \rho_t \right] (d\tilde{N}_t - \text{Tr}[\mathcal{J}(\rho_t)]dt)$$

Recall: (\tilde{N}_t) is a counting process of intensity $\int_0^t \text{Tr}[\mathcal{J}(\rho_s)]ds$.

Questions

What is the meaning of this equation?

How can we define (\tilde{N}_t) and (ρ_t)?

Process-solution de

$$d\rho_t = L(\rho_t)dt + \left[\frac{\mathcal{J}(\rho_t)}{\text{Tr}[\mathcal{J}(\rho_t)]} - \rho_t \right] (d\tilde{N}_t - \text{Tr}[\mathcal{J}(\rho_t)]dt)$$

Definition

Let $(\Omega, \mathcal{F}, \mathcal{F}_t, P)$ be a probability space. A **process-solution** of the jump equation is a couple (ρ_t, \tilde{N}_t) such that

$$\begin{aligned} \rho_t &= \rho_0 + \int_0^t \left[L(\rho_{s-}) - \mathcal{J}(\rho_{s-}) + \text{Tr}[\mathcal{J}(\rho_{s-})]\rho_{s-} \right] ds \\ &+ \int_0^t \left[\frac{\mathcal{J}(\rho_{s-})}{\text{Tr}[\mathcal{J}(\rho_{s-})]} - \rho_{s-} \right] d\tilde{N}_s \quad \text{a.s} \end{aligned}$$

and such that

$$\tilde{N}_t - \int_0^t \text{Tr}[\mathcal{J}(\rho_{s-})]ds$$

is a \mathcal{F}_t -martingale.

Theorem

Let $(\Omega, \mathcal{F}, \mathcal{F}_t, P)$ be a probability space equipped with a *random Poisson measure* μ on $\mathbb{R}^+ \times \mathbb{R}$ whose the *intensity measure* is the Lebesgue measure $ds \otimes dx$. The SDE

$$\begin{aligned} \rho_t = & \rho_0 + \int_0^t \left[L(\rho_{s-}) - \mathcal{J}(\rho_{s-}) + \text{Tr}[\mathcal{J}(\rho_{s-})]\rho_{s-} \right] ds \\ & + \int_0^t \int_{\mathbb{R}} \left[\frac{\mathcal{J}(\rho_{s-})}{\text{Tr}[\mathcal{J}(\rho_{s-})]} - \rho_{s-} \right] \mathbf{1}_{0 < x < \text{tr}[\mathcal{J}(\rho_{s-})]} \mu(ds, dx) \end{aligned}$$

admits a unique solution (ρ_t) . The process (\tilde{N}_t) defined by

$$\tilde{N}_t = \int_0^t \int_{\mathbb{R}} \mathbf{1}_{0 < x < \text{tr}[\mathcal{J}(\rho_{s-})]} \mu(ds, dx)$$

is a counting process of intensity $\int_0^t \text{Tr}[\mathcal{J}(\rho_{s-})] ds$.

One can generalise

$$\begin{aligned}\rho_t &= \rho_0 + \int L(\rho_{s-}) ds + \sum_{i=0}^l \int_0^t h_i(\rho_{s-}) dW_i(s) \\ &\quad + \sum_{i=0}^q \int_0^t \int_{\mathbb{R}} g_i(\rho_{s-}) \mathbf{1}_{0 < x < v_i(\rho_{s-})} [\mu_i(dx, ds) - dx ds],\end{aligned}$$

where $(W_t = (W_0(t), \dots, W_p(t)))$ are a p -dimensional Brownian motion and μ_i are $p + 1$ random measure of intensity $ds \otimes dx$. All the processes are independent.

Remark The functions h_i et g_i are given by

$$\begin{aligned}h_i(\rho) &= C_i \rho + \rho C_i^* - \text{Tr}[\rho(C_i + C_i^*)] \rho \\ g_i(\rho) &= \frac{\mathcal{J}_i(\rho)}{\text{Tr}[\mathcal{J}_i(\rho)]} - \rho\end{aligned}$$

III) Convergence Result

FROM DISCRETE TO CONTINUOUS QUANTUM TRAJECTORIES

Back to the discrete setup

- Setup

- 1 Case $\mathcal{H}_0 = \mathcal{H} = \mathbb{C}^2$
- 2 An observable of \mathcal{H} is of the form $A = \lambda_0 P_0 + \lambda_1 P_1$.
- 3 The discrete stochastic Schrödinger equation

$$\rho_{k+1} = \frac{\mathcal{L}_0(\rho_k)}{p_0(\rho_k)} \mathbf{1}_0^{k+1} + \frac{\mathcal{L}_1(\rho_k)}{p_1(\rho_k)} \mathbf{1}_1^{k+1}.$$

- Let us introduce $X_{k+1} = \frac{\mathbf{1}_1^{k+1} - p_1(\rho_k)}{\sqrt{p_0(\rho_k)p_1(\rho_k)}}$.

- In terms of X_{k+1} , we get

$$\rho_{k+1} = \mathcal{L}_0(\rho_k) + \mathcal{L}_1(\rho_k) + \left[-\sqrt{\frac{p_0}{p_1}} \mathcal{L}_0(\rho_k) + \sqrt{\frac{p_1}{p_0}} \mathcal{L}_1(\rho_k) \right] X_{k+1}.$$

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- Recall that (ρ_k) is defined through the quantity

$$\mathcal{L}_i(\rho) = \text{Tr}_{\mathcal{H}}[(I \otimes P_i) U(n)(\rho \otimes \beta)U^*(n) (I \otimes P_i)]$$

$$U(n) = e^{i\frac{1}{n}H_{tot}}$$

Naturally **the asymptotic assumptions** are going to appear in $U(n)$.

- Now, we fix a basis $\{\Omega_0, \Omega_1\}$

The reference state of the chain will be $\beta = |\Omega_0\rangle\langle\Omega_0|$.

- Let us write $U(n)$ as

$$U(n) = \begin{pmatrix} U_0^0(n) & U_0^1(n) \\ U_1^0(n) & U_1^1(n) \end{pmatrix}$$

where the $U_{ij}(n)$ are operators \mathcal{H}_0 .

- S. Attal-Y. Pautrat:** From repeated to continuous quantum interactions, "*Annales Henri Poincaré*"

In the previous article, the authors gives a precise description of the asymptotic conditions that we need to impose to $U_{ij}(n) \implies$ in order to obtain a non-trivial limit for the quantum repeated interactions model (interms of quantum stochastic calculus).

- In our context, we naturally adopt their conditions and we need

$$U_0^0(n) = I + \frac{1}{n} \left(-iH_0 - \frac{1}{2} C^* C \right) + o\left(\frac{1}{n}\right)$$

$$U_1^0(n) = \frac{1}{\sqrt{n}} C + o\left(\frac{1}{n}\right)$$

Limit evolution in the case of a diagonal A

- If A is **diagonal** in $\{\Omega_0, \Omega_1\}$. For example $A = 1 \times |\Omega_0\rangle\langle\Omega_0| + 0 \times |\Omega_1\rangle\langle\Omega_1|$, then we have

$$\mathcal{L}_0(\rho) = U_0^0 \rho (U_0^0)^* = \rho + \frac{1}{n} \left[(-iH_0 - \frac{1}{2}C^*C)\rho + \rho(-iH_0 - \frac{1}{2}C^*C) \right]$$

$$\mathcal{L}_1(\rho) = U_1^0 \rho (U_1^0)^* = \frac{1}{n} C \rho C^*$$

The transition probabilities satisfies

$$p_0(\rho_k) = 1 - \frac{1}{n} \text{Tr}[\mathcal{J}(\rho_k)] + o\left(\frac{1}{n}\right)$$

$$p_1(\rho_k) = \frac{1}{n} \text{Tr}[\mathcal{J}(\rho_k)] + o\left(\frac{1}{n}\right).$$

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$$\begin{aligned}\rho_{k+1} &= \rho_k + \frac{1}{n}[L(\rho_k) + o(1)] \\ &\quad + \left(\frac{\mathcal{J}(\rho_k)}{\text{Tr}[\mathcal{J}(\rho_k)]} - \rho_k + o(1) \right) (\mathbf{1}_1^{k+1} - \rho_1(\rho_k)).\end{aligned}$$

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$$\begin{aligned}\rho_0(\rho_k) &= 1 - \frac{1}{n} \text{Tr}[\mathcal{J}(\rho_k)] + o\left(\frac{1}{n}\right) \\ \rho_1(\rho_k) &= \frac{1}{n} \text{Tr}[\mathcal{J}(\rho_k)] + o\left(\frac{1}{n}\right).\end{aligned}$$

In case of a non diagonal A

- If $A = \lambda_0 P_0 + \lambda_1 P_1$ de \mathcal{H} **is not diagonal** in $\{\Omega_0, \Omega_1\}$. For example $A = |\Omega_0\rangle\langle\Omega_1| + |\Omega_1\rangle\langle\Omega_0|$ we get the following asymptotic expression

$$\mathcal{L}_0(\rho) = \frac{1}{2} (U_0^0 \rho (U_0^0)^* + U_0^0 \rho (U_1^0)^* + U_1^0 \rho (U_0^0)^* + U_1^0 \rho (U_1^0)^*) \quad (1)$$

$$= \frac{1}{2} \left(\rho + \frac{1}{\sqrt{n}} (C\rho + \rho C^*) + \frac{1}{n} L(\rho) \right) \quad (2)$$

$$\mathcal{L}_1(\rho) = \frac{1}{2} (U_0^0 \rho (U_0^0)^* - U_0^0 \rho (U_1^0)^* - U_1^0 \rho (U_0^0)^* + U_1^0 \rho (U_1^0)^*)$$

- Here the **probabilities** are

$$p_0(\rho_k) = \frac{1}{2} + \frac{1}{\sqrt{n}} \left[\text{Tr}[\rho_k (C + C^*)] + o(1) \right]$$

$$p_1(\rho_k) = 1 - p_0(\rho_k).$$

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$$\begin{aligned}\rho_{k+1} &= \rho_k + \frac{1}{n}[L(\rho_k) + o(1)] \\ &+ \frac{1}{\sqrt{n}}[C\rho_k + \rho_k C^* - \text{Tr}[\rho_k(C + C^*)]\rho_k + o(1)]X_{k+1}.\end{aligned}$$

Here the **probabilities** are

$$\begin{aligned}p_0(\rho_k) &= \xi + \frac{1}{\sqrt{n}}\nu \left[\text{Tr}[\rho_k(C + C^*)] + o(1) \right] \\ p_1(\rho_k) &= 1 - p_0(\rho_k).\end{aligned}$$

Convergence to the diffusive case

- From the previous description, we put

$$\rho_{[nt]} = \rho_0 + \sum_{k=0}^{[nt]-1} \frac{1}{n} [L(\rho_k) + o(1)] + \sum_{k=0}^{[nt]-1} \frac{1}{\sqrt{n}} [\mathcal{H}(\rho_k) + o(1)] X_{k+1}.$$

- Putting

$$\rho_n(t) = \rho_{[nt]}, \quad V_n(t) = \frac{[nt]}{n}, \quad W_n(t) = \frac{1}{\sqrt{n}} \sum_{k=0}^{[nt]-1} X_{k+1}.$$

- We have that $(\rho_n(t))$ satisfies

$$\rho_n(t) = \rho_0 + \int_0^t L(\rho_n(s-)) dV_n(s) + \int_0^t \mathcal{H}(\rho_n(s-)) dW_n(s) + \varepsilon_n(t).$$

Theorem

The process $(W_n(t), V_n(t), \varepsilon_n(t))$ converge in distribution to $(W_t, V_t, 0)$ where (W_t) is a standard Brownian motion and $V_t = t$ for all t .

Moreover, we have

$$\sup_n \mathbf{E} \left[[W_n(t), W_n(t)] \right] < \infty$$

Then the process $(\rho_n(t))$ satisfying

$$\rho_n(t) = \rho_0 + \int_0^t L(\rho_n(s-)) dV_n(s) + \int_0^t \mathcal{H}(\rho_n(s-)) dW_n(s) + \varepsilon_n(t)$$

converge in distribution to (ρ_t) the unique solution of

$$\rho_t = \rho_0 + \int_0^t L(\rho_s) ds + \int_0^t \mathcal{H}(\rho_s) dW_s.$$

The jump case

- Again

$$\begin{aligned}\rho_{[nt]} &= \rho_0 + \sum_{k=0}^{[nt]-1} \frac{1}{n} \left[L(\rho_k) - \mathcal{J}(\rho_k) + \text{Tr}[\mathcal{J}(\rho_k)]\rho_k + o(1) \right] \\ &\quad + \sum_{k=0}^{[nt]-1} \left(\frac{\mathcal{J}(\rho_k)}{\text{Tr}[\mathcal{J}(\rho_k)]} - \rho_k + o(1) \right) \mathbf{1}_1^{k+1}.\end{aligned}$$

- Again, we put

$$\rho_n(t) = \rho_{[nt]}, \quad V_n(t) = \frac{[nt]}{n}, \quad N_n(t) = \sum_{k=0}^{[nt]-1} \mathbf{1}_1^{k+1}.$$

- We get the discrete SDE

$$\rho_n(t) = \rho_0 + \int_0^t \Theta(\rho_n(s-)) dV_n(s) + \int_0^t \Phi(\rho_n(s-)) dN_n(s) + \varepsilon_n(t).$$

- Kurtz-Protter?
- In the jump case, **we can not directly show that** $(N_n(t))$ converge in distribution to the process (\tilde{N}_t) .
- Method:
 - ① Coupling.
 - ② Comparison with a Euler scheme.

Theorem

*The process $(\rho_n(t))$ defined from the quantum repeated measurement of a **diagonal observable** converge in distribution to (ρ_t) solution of the jump equation.*

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Theorem

*The process $(\rho_n(t))$ defined from the quantum repeated measurement of a **diagonal observable** converge in distribution to (ρ_t) solution of the jump equation.*

- How can we compare the two methods?
- How can we show that

$$\rho_{k+1} = \sum_{i=0}^p \frac{\mathcal{L}_i(\rho_k)}{\text{Tr}[\mathcal{L}_i(\rho_k)]} \mathbf{1}_i^{k+1}$$

converges to

$$\begin{aligned} \rho_t &= \rho_0 + \int L(\rho_{s-}) ds + \sum_{i=0}^l \int_0^t h_i(\rho_{s-}) dW_i(s) \\ &\quad + \sum_{i=0}^q \int_0^t \int_{\mathbb{R}} g_i(\rho_{s-}) \mathbf{1}_{0 < x < v_i(\rho_{s-})} [\mu_i(dx, ds) - dx ds]. \end{aligned}$$

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- **Martingale problem.**

- The discrete model can be describe by the transition Kernel

$$\Pi_n(\rho, \mu) = \sum_{i=0}^p p^i(\rho) \delta_{\mathcal{L}_i^{(n)}(\rho)/\text{Tr}[\mathcal{L}_i^{(n)}(\rho)]}(\mu), \quad (3)$$

- We can then define the Markov generator

$$\begin{aligned} \mathcal{A}_n f(\rho) &= n \int (f(\mu) - f(\rho)) \Pi_n(\rho, d\mu) \\ &= n \sum_{i=0}^p \left(f(\mathcal{L}_i^{(n)}(\rho)/\text{Tr}[\mathcal{L}_i^{(n)}(\rho)]) - f(\rho) \right) p^i(\rho). \end{aligned} \quad (4)$$

- Defining $\rho_n(t) = \rho_{[nt]}$ and $\mathcal{F}_t^n = \sigma(\rho_n(s), s \leq t)$

$$f(\rho_n(k/n)) - f(\rho_0) - \sum_{j=0}^{k-1} \frac{1}{n} \mathcal{A}_n f(\rho_n(j/n)) \quad (5)$$

is a $(\mathcal{F}_{k/n}^n)$ martingale

- You compute the limit of \mathcal{A}_n denoted by \mathcal{A}
- In distribution, there exists a unique Markov process $\rho(t)$ such that

$$f(\rho(t)) - f(\rho(0)) - \int_0^t \mathcal{A}f(\rho(s))ds$$

is a martingale with respect to the natural filtration of $(\rho(t))$

- After identifying \mathcal{A} , one can show that this Markov generator is the same as the one of the solution of the generalization of the stochastic master equation.
- This gives the expected convergence in distribution

IV) Estimation, Temperature

- The problem of estimation concerns similar models where we do not know the initial state. Nevertheless we have access to the results of the measurement (for example the value 0 or 1).
- If ρ design the true initial state (that we do not know), we know that 0 appears with probability $p_0(\rho)$ and 1 with probability $p_1(\rho)$.
- Now let $\tilde{\rho}$ an arbitrary state, conditionally to the result of the measurement, we put

$$\tilde{\rho}_1(i) = \frac{\mathcal{L}_i(\tilde{\rho})}{\text{Tr}[\mathcal{L}_i(\tilde{\rho})]}$$

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Estimation

- What can we expect?
- We can expect that the distance between $\tilde{\rho}_1$ and ρ_1 is smaller than the one between $\tilde{\rho}$ and ρ .
- Roughly speaking this means that knowing the result of the measurement allows us to estimate the true quantum trajectory.
- Within the previous procedure we can produce a random sequence $\tilde{\rho}_k$ whose transition probability are given by the one of the "true" quantum trajectory .

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- In order to evaluate if the "true" and the "estimate" quantum trajectory get closer we use the fidelity distance

$$F(\rho, \mu) = \text{Tr}[\sqrt{\sqrt{\rho}\mu\sqrt{\rho}}]^2$$

- We have $F(\rho, \mu) = F(\mu, \rho)$ and if for example $\rho = |\psi\rangle\langle\psi|$

$$F(\rho, \mu) = \text{Tr}[\rho\mu]$$

- $F(\rho, \mu) = 1$ if and only if $\rho = \mu$.

- We have the following result

$$\mathbb{E}[F(\tilde{\rho}_{k+1}, \rho_{k+1}) | (\tilde{\rho}_k, \rho_k)] \geq F(\tilde{\rho}_k, \rho_k),$$

- This means that $F(\tilde{\rho}_k, \rho_k)$ is a sub-martingale.
- It is not a trivial result since there exist distance where this property is not satisfied.
- Problem: What do we have to impose on the system to have $\lim_{k \rightarrow \infty} F(\tilde{\rho}_k, \rho_k) = 1$
- Similar result for continuous time models (the discrete approach is really useful to show the sub-martingale result).

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- Similar result for continuous time models.

$$\begin{aligned}d\rho(t) &= L(\rho(t-))dt + \sum_{i=0}^p H_i(\rho(t-))(dy_i(t) - \text{Tr}[(C_i + C_i^*)\rho(t-)]dt) \\ &\quad + \sum_{i=p+1}^n \left(\frac{J_i(\rho(t-))}{v_i(\rho(t-))} - \rho(t-) \right) (dN_i(t) - v_i(\rho(t-))dt). \\ \rho(0) &= \rho_0\end{aligned}$$

$$\begin{aligned}d\tilde{\rho}(t) &= L(\tilde{\rho}(t-))dt + \sum_{i=0}^p H_i(\tilde{\rho}(t-))(dy_i(t) - \text{Tr}[(C_i + C_i^*)\tilde{\rho}(t-)]dt) \\ &\quad + \sum_{i=p+1}^n \left(\frac{J_i(\tilde{\rho}(t-))}{v_i(\tilde{\rho}(t-))} - \tilde{\rho}(t-) \right) (dN_i(t) - v_i(\tilde{\rho}(t-))dt). \\ \tilde{\rho}(0) &= \tilde{\rho}_0,\end{aligned}$$

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- The fidelity is still a sub-martingale
- Almost impossible to estimate the term

$$dF(\tilde{\rho}(t), \rho(t)) = d\text{Tr}[\sqrt{\sqrt{\tilde{\rho}(t)}\rho(t)\sqrt{\tilde{\rho}(t)}}}]^2$$

- Use of the sub-martingale property for the discrete time result and the convergence result

Temperature

- The state of the environment $\beta = |\Omega_0\rangle\langle\Omega_0|$ represents the vacuum state. Considering such a state is crucial in the approach of Attal-Pautrat.
- In the work of Attal-Joye, they consider a modelization of a heat bath by taking

$$\beta = \frac{e^{-\beta H}}{Z} = \begin{pmatrix} \beta_0 & 0 \\ 0 & \beta_1 \end{pmatrix},$$

where β is the inverse of a temperature. this is the usual Gibbs state. Using a GNS representation and adapting the work of Attal-Pautrat they manage to derive Quantum Langevin Equation for heat bath.

- What are the limit equation in the context of QRM.
- Following the guideline of Attal-Joye we use the GNS representation and we adapt the general result (with multiple noise). Surprisingly (for me) no jumping processes remain.
- For example in the case of two results, in the case of the diagonal observable, the limit equations is deterministic: just the Lindblad master equation (with temperature parameter)
- In the case of non diagonal the Wiener process remains

- Alternative: from a physical point of view, you can consider that the Gibbs state $\begin{pmatrix} \beta_0 & 0 \\ 0 & \beta_1 \end{pmatrix}$, is either

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix},$$

with probability β_0 or

$$\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix},$$

with probability β_1 .

- One can consider the evolution without measurement, the random aspect disappears at the limit and we recover the Lindblad equation (as if the random aspect is averaged)
- But with measurement, in the diagonal case it appears two different jumps and in the non diagonal case two different Wiener processes.

THANK YOU