

A duality formula + Particle Gibbs for continuous time Feynman-Kac measures

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- ▶ this talk \leadsto Arxiv (18)
- ▶ Discrete time duality + Taylor expansions [+ Kohn-Patras]:
Arxiv (14), CRAS (15), IHP (16)
- ▶ Sharp 1st order analysis [+ Jasra]: SPA I + SPA-II (18)
- ▶ Influenced by the pioneering article

PMCMC - Andrieu, Doucet, Holenstein JRSS-B-10

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Integration (Lebesgue or Riemann)

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In this talk: Measure estimates w.r.t. total variation distance

$$\mu^\epsilon = \mu + \epsilon \iff \|\mu^\epsilon - \mu\|_{tv} \leq c \epsilon \quad \text{for some constant } c \perp \epsilon$$

X_t Markov \in metric space S and potential $V(X_t) \geq 0$

Historical processes

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  - ~~> Ito-functional calculus ∈  
[Dupire-09, Cont-Fournié-13, Jazaerli-Saporito-17, Saporito-14]

# FK-measures on path space := $FK(\widehat{X}, \widehat{V})$

$$d\mathbb{Q}_t = \frac{1}{Z_t} \exp \left[ - \int_0^t \widehat{V}_s(\widehat{X}_s) ds \right] d\mathbb{P}_t$$

with the normalizing constant

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$\Updownarrow$

*Weak formulation on test functions  $F$  on  $S_t$ :*

$$\mathbb{Q}_t(F) = \int F(x) \mathbb{Q}_t(dx) \propto \mathbb{E} \left( F(\widehat{X}_t) \exp \left[ - \int_0^t \widehat{V}_s(\widehat{X}_s) ds \right] \right)$$

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$X_t = \hat{Y}_t := (Y_s)_{0 \leq s \leq t} \rightsquigarrow$  *Path-space again (even more general):*

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# $FK(\hat{X}, \hat{V})$ - particle sampler = $N$ -path particles

$\xi_t := (\xi_t^i)_{1 \leq i \leq N} \in S_t^N$  with ancestral lines  $\xi_t^i = (\xi_{s,t}^i)_{1 \leq s \leq t}$

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$$\mathbb{X}_t \sim m(\xi_t) := \frac{1}{N} \sum_{1 \leq i \leq N} \delta_{\xi_t^i}$$

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In the talk : All particle  $N$ -product spaces up to index permutations (= empirical measures)

## Equivalent interpretations

Genetic algorithm [Turing - 1950], Spatial branching processes [Galton-Watson - 1873, Yaglom - 1947, Harris - 1951], Moran and/or Nanbu type particle system, Mean field particle interpretations of nonlinear Markov processes [Fermi - 1948, McKean - 1966],...

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- ▶ Requires *symmetric/neutral* branching/selection functions/rates

## Some key results (when $FK(X, V)$ stable)

**Fluctuation/clt + expo. concentration + Idp :**

$$m(\xi_t) = \mathbb{Q}_t + \frac{1}{\sqrt{N}} \mathbb{V}_t \quad \text{with} \quad \mathbb{V}_t \simeq \text{Gaussian field}$$

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**Bias/propagation of chaos:**

$$\text{Law}(\mathbb{X}_t) = \mathbb{Q}_t + \frac{t}{N} \quad \text{and} \quad \text{Law}(\mathbb{X}_{s,t}) = \mathbb{Q}_t + \frac{1 + (t - s)}{N}$$

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↳ *not so new:* ↵ [discrete time MPRF-96, continuous time SPA-00]

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- *Independent Metropolis-Hastings - obvious*

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⇓

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- ▶ i.i.d. initial conditions.
- ▶ Between jumps : independent + same evolutions as  $\widehat{X}_t$ .
- ▶ Jump/killing rate  $(1 - 1/N)$   $\widehat{V}_s$ .
- ▶ Jump onto/offspring birth - uniformly on the path-pool.
- ▶ **Extra jump rate**  $(2/N)$   $\widehat{V}_s$  onto  $\mathbb{X}_s$ .

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**Nb.:** without the blue

**the above process coincides with the FK-particle sampler**

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Given  $\mathbb{X}_t = x \sim \mathbb{Q}_t$

Coincides with the particle sampler  $\widehat{\zeta}_t^- = (\zeta_t^i)_{2 \leq i \leq N}$  of FK measures:

$$\widehat{V}_s \rightsquigarrow (1 - 1/N) \widehat{V}_s$$

$$\widehat{X}_s \rightsquigarrow \widehat{X}_s \oplus \text{extra jump rate } (2/N) \widehat{V}_s \text{ onto } \zeta_s^1 := x_s$$

# Equivalent formulation - Duality formula

**Theo.:** [+Arnaudon ↠ Arxiv (18)]

$$\mathbb{E} \left( F(\mathbb{X}_t, \widehat{\xi}_t) e^{- \int_0^t m(\xi_s)(\widehat{V}_s) ds} \right) = \mathbb{E} \left( F(\widehat{\mathbb{X}}_t, \widehat{\zeta}_t) e^{- \int_0^t \widehat{V}_s(\widehat{\mathbb{X}}_s) ds} \right)$$

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- = random sample from  $\mathbb{Q}_t$
- = many-body/twisted distribution on  $\widehat{\xi}_t$  and

$$(\widehat{\xi}_t \mid \mathbb{X}_t = x) = (\widehat{\zeta}_t \mid \widehat{\mathbb{X}}_t = x) \quad \mathbb{Q}_t - a.e. \ x$$

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- *discrete time version*

↪ +Kohn+Patras, Arxiv (14), CRAS (15)+IHP (16)

## ⇒ Reversible Gibbs-sampler with target $\mathbb{Q}_t$

$$\mathbb{X} = x \longrightarrow \left[ \begin{array}{l} \hat{\zeta}_t^x \sim \left( \hat{\zeta}_t \mid \hat{X}_t = x \right) \end{array} \right] \longrightarrow \overline{\mathbb{X}} = \bar{x} \sim m(\hat{\zeta}_t^x)$$

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$\mathbb{Q}_t$ -reversible Markov transition:

$$\mathbb{K}_t(f)(x) = \mathbb{E} \left[ m(\hat{\zeta}_t^x)(f) \right]$$

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Key observation for the convergence analysis:

$$m(\widehat{\zeta}_t^x) = \frac{1}{N} \delta_x + \left(1 - \frac{1}{N}\right) m(\widehat{\zeta}_t^{x,-})$$

with  $\widehat{\zeta}_t^{x,-} = (\zeta_s^{x,-})_{0 \leq s \leq t}$  = particle sampler of an FK measure:

$$\widehat{V}_s \rightsquigarrow (1 - 1/N) \widehat{V}_s$$

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# The FK-bias/propagation of chaos

$$\mathbb{K}_t(f)(x) = \mathbb{E} \left[ m(\widehat{\zeta}_t^x)(f) \right] = \mathbb{Q}_t^x(f) + \frac{t}{N}$$

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## "A little" perturbation analysis

$$\mathbb{Q}_t^x = \mathbb{Q}_t + \frac{t}{N} \implies \delta_x \mathbb{K}_t = \mathbb{Q}_t + \frac{t}{N} \implies \delta_x \mathbb{K}_t - \delta_y \mathbb{K}_t = \frac{t}{N}$$

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## More refined (discrete time)

**Theo.** [+Kohn+Patras, Arxiv (14), CRAS (15)+IHP (16)]

$$\mathbb{K}_t = \mathbb{Q}_t + \sum_{1 \leq k < l} \left( \frac{t}{N} \right)^k \partial^{(k)} \mathbb{K}_t + \left( \frac{t}{N} \right)^l \bar{\partial}^l \mathbb{K}_t$$

( $\forall l \geq 1$ )  $\oplus$  Differential operators  $\sim$  coalescent/colored trees and s.t.

$$\left\| \partial^{(k)} \mathbb{K}_t \right\| \vee \left\| \bar{\partial}^{(k)} \mathbb{K}_t \right\| \leq k^{2k} \quad \text{as soon as } l^2 t / N < 1.$$

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$\Downarrow$  [direct]

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Continuous time ?  $\rightsquigarrow$  Open problem/question

"Hint/solution"  $\rightsquigarrow$  Taylor/Bias [+ Rubenthaler-Patras] JTP (09)

$\oplus$  backward analysis using prop 2.1 in  $\rightsquigarrow$  Arxiv (18)