# Expected Values Estimated via Mean-Field Approximation are $1 / N$-Accurate 

Nicolas Gast

Inria, Grenoble, France

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## Mean-field approximation is widely used in our community

2016 Asymptotics of Insensitive Load Balancing and Blocking Phases - Jonckheere - Prabhu

2016 On the Approximation Error of Mean-Field Models - Ying
2015 Power of d Choices for Large-Scale Bin Packing: A Loss Model - Xie et al
2015 Transient and Steady-state Regime of a Family of List-based Cache Replacement Algorithms - Gast, Van Houdt
2013 Queueing system topologies with limited flexibility. - Tsitsiklis, Xu
2013 A mean field model for a class of garbage collection algorithms in flash-based solid state drives. - Van Houdt
2012 Fluid limit of an asynchronous optical packet switch with shared per link full range wavelength conversion. - Van Houdt, Bortolussi
2011 On the power of (even a little) centralization in distributed processing. -
2010 Randomized load balancing with general service time distributions. Bramson et al.
2010 Incentivizing peer-assisted services: a fluid shapley value approach. - Misra et al
2010 A mean field model of work stealing in large-scale systems. - Gast, Gaujal
2009 The age of gossip: spatial mean field regime. - Chaintreau et al.

## What is mean-field approximation?

We study a population of $N$ interchangeable objects.
$X^{(N)}$ denotes the empirical measure.

$$
X_{i}^{(N)}(t)=\text { fraction of objects in state } i
$$

Idea of mean-field: Some models simplify as $N \rightarrow \infty$
Theorem (Kurtz 70,...)

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Example: $N$ servers


Randomly choose two, and select one


## Idea of mean-field: Some models simplify as $N \rightarrow \infty$

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## How accurate is mean-field approximation?

$$
\begin{aligned}
& \text { Theorem (Kurtz 70... } \\
& \qquad x^{(N)}(t) \approx x(t)+\frac{1}{\sqrt{N}} G_{t}
\end{aligned}
$$

$\uparrow X_{3}(t)$ - Fraction of servers with 3 jobs

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In practice, how accurate is mean-field approximation?
Can we use the approximation for $N=1000$ ? $N=100$ ? $N=10$ ?

| $N$ | 10 | 100 | 1000 | $\infty$ |
| :---: | :---: | :---: | :---: | :---: |
| Average queue length (simulation) | 2.8040 | 2.3931 | 2.3567 | 2.3527 |
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Table: Two-choice model with $\rho=0.9$

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Contributions:
(1) We show that under very general conditions, the error is in $O(1 / N)$
(2) We show that for that, the drift needs to be twice-differentiable.
(3) We study numerically the power-of-two choice.

## Outline

(1) The Kurtz's Population Model: Classical Convergence Results
(2) The $O(1 / N)$-Accuracy of Mean-Field Approximation
(3) Example: two-choice model
(4) Recap and Discussion

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4 Recap and Discussion

## CTMC

A continous-time Markov chain (CTMC) with state-space $\mathbf{E}$ is given by an initial state $x_{0}$ and its transitions $(\ell \in \mathcal{L})$ :

$$
X \mapsto X+\ell \quad \text { at rate } \quad \beta_{\ell}(X)
$$

The drift is $f(x)=\sum_{\ell} \ell \beta_{\ell}(x)$.
${ }^{1}$ We assume $(\mathbf{E},\|\cdot\|)$ is a Banach space, not necessarily $\mathbb{R}^{d}$.

## Population CTMC

Density dependent population process (70s)

A population process is a sequence of CTMC $\mathbf{X}^{N}$, indexed by the population size $N$, with state spaces $\mathbf{E}^{N} \subset \mathbf{E}$, with initial state $x_{0}$ and with transitions (for $\ell \in \mathcal{L}$ ):

$$
X \mapsto X+\frac{\ell}{N} \quad \text { at rate } N \beta_{\ell}(X)
$$

The drift is $f(x)=\sum_{\ell} \ell \beta_{\ell}(x)$.

[^0]
## Transient regime

Let $\Phi_{t}$ denotes the (unique) solution of the ODE :

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\Phi_{t} x=x+\int_{0}^{t} f\left(\Phi_{s} x\right) d s
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Theorem (Kurtz 70s)
If $f$ is Lipschitz-continuous with constant $L$, then for any fixed $T$ :

$$
\sup _{t<T}\left\|X^{N}(t)-\Phi_{t} X^{N}(0)\right\|=O(1 / \sqrt{N}) \quad\left(\lim _{N \rightarrow \infty} \cdot=0\right)
$$



## Stationary regime

If the ODE $\dot{x}=f(x)$ has a unique fixed point $x^{*}$ that is exponentially stable, then:

Theorem (Ying 2016)
If $f$ is Lipschitz-continuous with constant $L$, then for any fixed $T$ :

$$
\mathbb{E}\left[\left\|x^{N}-x^{*}\right\|\right]=O(1 / \sqrt{N}) \quad\left(\lim _{N \rightarrow \infty} \cdot=0\right)
$$

(the uniqueness of the fixed point is not sufficient, see Benaim-Le Boudec 2008).

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## $1 / \sqrt{N}$ or $1 / N$ ?



| $N$ | 10 | 100 | 1000 | $+\infty$ |
| :---: | :---: | :---: | :---: | :---: |
| Average queue length $\left(m^{N}\right)$ | 2.81 | 2.39 | 2.36 | 2.35 |
| Error $\left(m^{N}-m^{\infty}\right)$ | 0.46 | 0.039 | 0.004 | 0 |

$1 / \sqrt{N}$ or $1 / N$ ?

$$
O(1 / \sqrt{N})
$$

System property (depends on $\left.\mathbf{X}^{N}\right) \quad$ (dep. on $\left.\mathbb{E}\left[h\left(\mathbf{X}_{i}\right)\right]\right)$ $O(1 / \sqrt{N})($ CLT $)$
$O(1 / N):$
$x_{i}$ (mean-field approx)

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## Steady-state analysis

We say that $\dot{x}=f(x)$ has an exponentially stable attractor $x^{*}$ if for any solution:

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\left\|x(t)-x^{*}\right\| \leq C e^{-\alpha t}\left\|x(0)-x^{*}\right\|
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## Theorem

If $f$ is twice differentiable, if the ODE has an exponentially stable attractor $x^{*}$ and if there exists a bounded set $\mathcal{B}$ such that $\mathbf{P}\left[X^{N} \notin \mathcal{B}\right]=O\left(1 / N^{2}\right)$, then for any bounded function $h$, there exists a constant $K$ such that:

$$
\limsup _{N \rightarrow \infty} N\left|\mathbb{E}\left[h\left(X^{N}\right)\right]-h\left(x^{*}\right)\right| \leq K
$$

- Note: A similar result holds for the transient behavior.


## Main ideas of the proof

1. Comparison of the generators:

$$
\begin{aligned}
\left(L^{(N)} h\right)(x) & =\sum_{\ell \in \mathcal{L}} N \beta_{\ell}(x)\left(h\left(x+\frac{\ell}{N}\right)-h(x)\right) \\
(\Lambda h)(x) & =\sum_{\ell \in \mathcal{L}} \beta_{\ell}(x) D h(x) \cdot \ell=\operatorname{Dh}(x) \cdot f(x)
\end{aligned}
$$

2. Stein's method :

$$
\mathbb{E}\left[h\left(X^{N}\right)-h\left(x^{*}\right)\right] \mathbb{E}\left[\left(\Lambda-L^{(N)}\right)(G h) X^{N}\right]
$$

where $G h(x)=\int_{0}^{\infty}\left(h\left(\Phi_{t} x\right)-h\left(x^{*}\right)\right) d t$ satisfies $(x)-h\left(x^{*}\right)=\Lambda(G h) x$.
3. Perturbation theory: $D^{2}(G h)$ is twice-differentiable.

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## The two choice model ${ }^{2}$



Infinite state-space:

$$
X_{0}(t), X_{1}(t), \ldots
$$

where
$X_{i}(t)=$ fraction with $i$ or more jobs.
Randomly choose two, and select one

## Does this model satisfies our assumptions?

$(\mathbf{E},\|\cdot\|)$ is the set of infinite sequences such that $\|x\|_{w}=\sum_{i=1}^{\infty} w_{i}\left|x_{i}\right|<\infty$.

- Transitions : easy
- Regularity of the drift : easy
- Unique attractor: mitzenmacher 98
- Stationary measure concentrates on a bounded set : coupling argument : 2-choice $\ll$ 1-choice.


## The power of two-choice

Our theory guarantees that the average queue length satisfies:

$$
m^{N}(\rho)=m^{\infty}(\rho)+O(1 / N)
$$

where $m^{\infty}(\rho)=\Theta_{\rho \rightarrow 1}\left(\log \frac{1}{1-\rho}\right)$.

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where $m^{\infty}(\rho)=\Theta_{\rho \rightarrow 1}\left(\log \frac{1}{1-\rho}\right)$.
By simulation, we observe that $N\left(m^{N}(\rho)-m^{\infty}\right)=d(\rho) \approx \frac{\rho^{2}}{2(1-\rho)}$

| $N$ | 10 | 20 | 30 | 50 | $+\infty$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $m^{N}(\rho)$ | 2.804 | 2.566 | 2.491 | 2.434 | - |
| $m^{\infty}(\rho)+\frac{\rho^{2}}{2 N(1-\rho)}$ | 2.758 | 2.555 | 2.488 | 2.434 | 2.353 |

Table: Average queue length in the two-choice model ( $\rho=0.9$ ).

The quality of the approximation degrades as $\rho$ goes to 1

Simulation results suggest that:

$$
\begin{aligned}
& m^{N}(\rho) \approx \underbrace{m^{\infty}(\rho)}+\frac{1}{N} \underbrace{d(\rho)}+O\left(\frac{1}{N^{2}}\right) \\
& \approx \log \frac{1}{1-\rho} \\
& \approx \frac{\rho^{2}}{2(1-\rho)}
\end{aligned}
$$

Conjecture: the power of two-choice holds if $N=\Omega\left(\frac{1}{1-\rho}\right)$

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## Recap

(1) Convergence of mean-field model is $O(1 / N)$.

- Works for transient and steady-state
- Works for infinite-dimensional state space.
(2) Our approach is to focus on the expected values


Our result: the difference is $O(1 / N)$

## In practice

For many mean-field models :

$$
\mathbb{E}\left[X^{N}\right] \approx x+\frac{C}{N}
$$

- $C$ can be computed for one $N$ and then interpolated.

This provides a new light for the two-choice.

## Does it always work?

- Works for the model of Kurtz
- Also works for the "Benaim-Le Boudec" by using uniformization

But: it requires the drift to be twice-differentiable.

- (see counter-example on the paper)


## Extension and open questions

- Multistable equilibria.
- Can we go to the order $O\left(1 / N^{2}\right)$ ? It is useful?
- I assumed twice differentiable (and it is needed).
- Can we do something in between for the steady-state?
- Non-homogeneous population.
- e.g., caching

Paper is reproducible:
https://github.com/ngast/meanFieldAccuracy

## Thank you!

http://mescal.imag.fr/membres/nicolas.gast

nicolas.gast@inria.fr

## Mean-field and decoupling

## Benaïm,

Le Boudec 08

Le Boudec 10

Darling Norris 08
G. 16
G. 16

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Applications: caches
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