

# Fluid limit-based tuning of some hybrid MCMC samplers

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We introduce

- ▶ a transformation of the Markov chain  $\longrightarrow$  continuous time process
- ▶ such that the *stability* of this process, is related to the ergodicity of the Markov chain.

When applied to MCMC,

- ▶ the dynamic of this transformation depends upon the design parameters of the algorithm ;
- ▶  $\Rightarrow$  learn from the dynamic, how to tune the design parameters.

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Theoretical results on the fluid limit

Tuning the design parameters for the Metropolis within Gibbs

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$\hookrightarrow$  Outline of the talk

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## Fluid Limit : definition (I)

Let  $\{\Phi_k, k \geq 0\}$  be a Markov chain, on  $X$  ( $X = \mathbb{R}^d$ ).

↔ Family of rescaled process  $\eta_r$ , for  $r > 0$

(i) in the initial point :  $\eta_r(0; x) = \frac{1}{r}\Phi_0 = x, \quad \Phi_0 = rx$

(ii) in time and space :  $\eta_r(t; x) = \frac{1}{r}\Phi_{\lfloor tr \rfloor}$

i.e.  $\eta_r(\cdot; x) = \frac{1}{r}\Phi_k$  on  $\left[\frac{k}{r}; \frac{(k+1)}{r}\right)$ .

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↔ Distributions

- $\mathbb{P}_x$  : distribution of  $\{\Phi_k, k \geq 0\}$  with initial distribution  $\delta_x$ .
- $\mathbb{Q}_{r;x}$  : image prob. of  $\mathbb{P}_{rx}$  by  $\eta_r$ ,

prob. on  $\mathbb{D}(\mathbb{R}^+, X)$  the space of cad-lag functions  $\mathbb{R}^+ \rightarrow X$

## Fluid Limit : definition (II)

↪ Definition  $\mathbb{Q}_x$  probability on  $\mathbb{D}(\mathbb{R}^+, X)$  - for  $x \in X$ , is a **fluid limit** if there exist scaling factors  $r_n \rightarrow +\infty$  such that

$$\mathbb{Q}_{r_n; x} \text{ converges weakly to } \mathbb{Q}_x.$$

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↪ **Existence**

$$\begin{aligned} \Phi_{k+1} &= \Phi_k + \mathbb{E}[\Phi_{k+1} | \mathcal{F}_k] - \Phi_k + \Phi_{k+1} - \mathbb{E}[\Phi_{k+1} | \mathcal{F}_k] \\ &= \Phi_k + \underbrace{\mathbb{E}_x[\Phi_{k+1} - \Phi_k | \mathcal{F}_k]}_{\Delta(\Phi_k)} + \underbrace{(\Phi_{k+1} - \mathbb{E}_x[\Phi_{k+1} | \mathcal{F}_k])}_{\epsilon_{k+1} \text{ martingale increment}}. \end{aligned}$$

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► **Result** (Fort et al, 2007)

If

- $\exists p > 1, \quad \lim_{K \rightarrow +\infty} \sup_{x \in \mathbb{X}} \mathbb{E}_x [|\epsilon_1|^p \mathbb{1}_{|\epsilon_1| > K}] \rightarrow 0.$
- $0 < \sup_{x \in \mathbb{X}} |\Delta(x)| < \infty.$

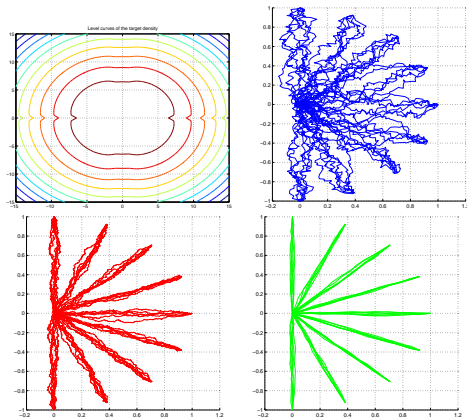
Then

- $\forall x, r_n \rightarrow +\infty, \exists$  subsequ.  $\{r_{n_j}\}$  s.t.  $\mathbb{Q}_{r_{n_j}; x} \Rightarrow \mathbb{Q}_x$
- $\mathbb{Q}_x$  prob. on the space of continuous functions.



## Ex. : Plot of $\eta_r(\cdot, x)$ on $[0, T]$ , for diff. $x$ on the unit sphere

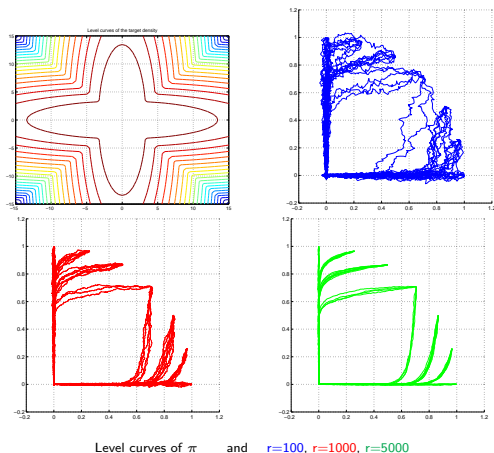
- ▶ Random Walk Hastings-Metropolis on  $\mathbb{R}^2$
- ▶ Target distribution  $\pi(x_1, x_2) \propto (1 + x_1^2 + x_2^2 + x_1^8 x_2^2) \exp(-(x_1^2 + x_2^2))$
- ▶ Gaussian proposal :  $q \sim \mathcal{N}_2(0, 4\mathbb{I})$



Level curves of  $\pi$  and  $r=100, r=1000, r=5000$

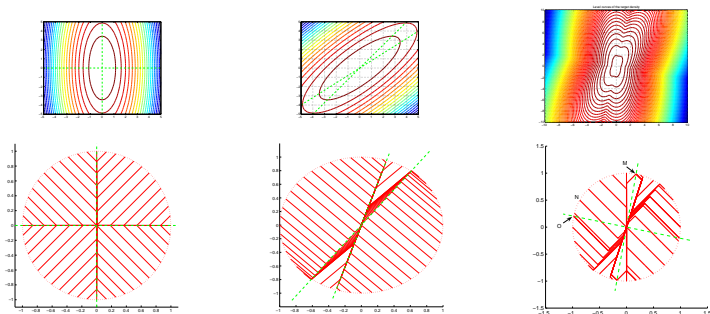
## Ex. : Plot of $\eta_r(\cdot, x)$ on $[0, T]$ , for diff. $x$ on the unit sphere

- ▶ Random Walk Hastings-Metropolis on  $\mathbb{R}^2$
- ▶ Target distribution  $\pi \propto \mathcal{N}_2(0, \Gamma_1) + \mathcal{N}(0, \Gamma_2)$
- ▶ Gaussian proposal :  $q \sim \mathcal{N}_2(0, \mathbb{I})$



## Ex. : Draws from $Q_x$ for diff. $x$ on the unit sphere

- ▶ Metropolis within Gibbs on  $\mathbb{R}^2$
- ▶ Target distribution  $\pi =$  (mixture of)  $\mathcal{N}_2$
- ▶ Gaussian proposal :  $q_i \sim \mathcal{N}(0, c)$  and  $\omega_i = 0.5$ .



Level curves of  $\pi$  and realizations of the fluid limit

# Existence (I)

$$\Phi_{k+1} = \Phi_k + \mathbb{E}[\Phi_{k+1} - \Phi_k | \mathcal{F}_k] + \epsilon_{k+1} = \Phi_k + \Delta(\Phi_k) + \epsilon_{k+1}.$$

↪ Sufficient conditions :

- ▶  $\exists p > 1, \quad \lim_{K \rightarrow +\infty} \sup_{x \in \mathcal{X}} \mathbb{E}_x [|\epsilon_1|^p \mathbf{1}_{|\epsilon_1| > K}] \rightarrow 0.$
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↪ Extension :

- ▶  $\exists \beta \in [0, 1 \wedge (p-1)[, \quad 0 < \sup_{x \in \mathcal{X}} |x|^\beta |\Delta(x)| < \infty.$

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↪ Extension :

- ▶  $\exists \beta \in [0, 1 \wedge (p-1)[, \quad 0 < \sup_{x \in \mathcal{X}} |x|^\beta |\Delta(x)| < \infty.$

In that case, change the definition of the re-scaled process :

$$\eta_{r,\beta}(t; x) = \frac{\Phi_{\lfloor tr^{1+\beta} \rfloor}}{r} \qquad \eta_{r,\beta}(0; x) = x.$$

↔ Example : RWHM chains

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- ▶  $\mathbb{E}[|\epsilon_{k+1}|^p] \longleftrightarrow \mathbb{E}[|\Phi_{k+1} - \Phi_k|^p]$  :  
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► Since

$$\Delta(x) = \int_{\{y, \pi(x+y) < \pi(x)\}} y \left( \frac{\pi(x+y)}{\pi(x)} - 1 \right) q(y) dy,$$

$\beta = 0$ , and  $p$ -moment of the proposal distribution  $q$ .



## Stability

↔ **Stable fluid limit model** : if there exist  $T > 0$  and  $\rho < 1$  s.t. for any  $x$  on the unit sphere,

$$\mathbb{Q}_x \left( \eta \in \mathbb{D}(\mathbb{R}^+, \mathbf{X}), \inf_{[0, T]} |\eta(t)| \leq \rho \right) = 1.$$

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► **Result** (Fort et al, 2007)

If

- $\{\Phi_k, k \geq 0\}$  is phi-irreducible and aperiodic; and compact sets are petite.
- fluid limit model exists and is stable,

Then the Markov chain is  $(f, r)$ -ergodic

$$(n+1)^{q-1} \sup_{\{f, |f| \leq 1 + |x|^{p-q}\}} |\mathbb{E}_x[f(\Phi_n)] - \pi(f)| \xrightarrow{n \rightarrow +\infty} 0, \quad 1 \leq q \leq p.$$

## Characterisation : case 1

► If

- $\exists h$  continuous s.t. for any compact set  $H$  of  $X \setminus \{0\}$ ,

$$\lim_{r \rightarrow +\infty} \sup_{x \in H} |\Delta(rx) - h(x)| = 0.$$

- the ODE  $\dot{\mu} = h(\mu)$  is stable.

Then the fluid limit model is stable.

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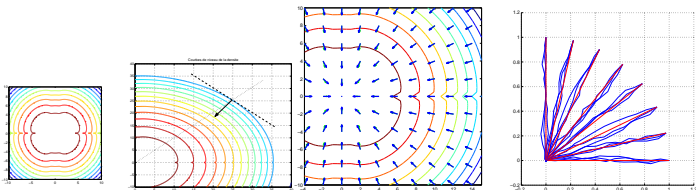
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Level curves of  $\pi$

and

Rejection area

and

the fields  $\Delta$ ,  $h$

and

realizations of the fluid limit

## Characterisation : case 2a

► If

- $\exists h$  continuous s.t. for any compact set  $H$  of cone of  $X \setminus \{0\}$ ,

$$\lim_{r \rightarrow +\infty} \sup_{x \in H} |\Delta(rx) - h(x)| = 0.$$

- the ODE  $\dot{\mu} = h(\mu)$  started in the cone is stable.
- the cone is “attractive”.

Then the fluid limit model is stable.

## Characterisation : case 2a

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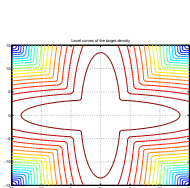
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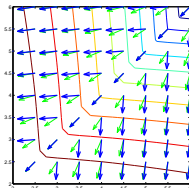
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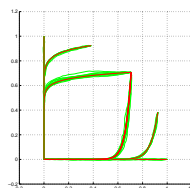
### ► Example : RWHM



Level curves of  $\pi$  and



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realizations of the fluid limit

## Characterisation : case 2b

$(X = \mathbb{R}^2)$



- $X = \bigcup_{\alpha=1}^a O_{\alpha} \cup \bigcup_{\beta=1}^b \{x, f'_{\beta}x = 0\}$ .
- $\exists \Sigma_{\alpha}$  s.t. for any compact set  $H$  of  $O_{\alpha}$ ,

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- “attractive” hyperplanes.

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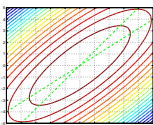
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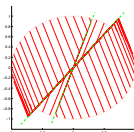
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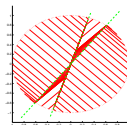
▶ Example : RW Metropolis within Gibbs



Level curves of  $\pi$  and



fluid limits when  $\omega_1 = 0.25$



and fluid limits when  $\omega_1 = 0.5$



## Characterisation : Conclusion / Perspectives

- ▶ Characterisation based on the radial behavior of  $\Delta(x) = \mathbb{E}_x[\Phi_1 - \Phi_0]$ ,

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for all compact subset  $H \subset \mathbb{R}^d$  ?  $\longrightarrow$  quite complex.

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for all compact subset  $H \subset \mathbb{R}^d$  ?  $\longrightarrow$  quite complex.

- ▶ Deterministic fluid limit more or less everywhere since

$$\eta_{r,\beta}(t; x) = \frac{\Phi_{\lfloor tr^{1+\beta} \rfloor}}{r} \quad \eta_{r,\beta}(0; x) = x,$$

with  $0 \leq \beta < 1$ .

When  $\beta = 1 \longrightarrow$  diffusion (work in progress, M. Bédart and E. Moulines).

# The design parameters

↔ At each iteration,

- ▶ Choose a component  $i \in \{1, \dots, d\}$  with probability  $\omega_i$ .
- ▶ Update the  $i$ -th component with a RW move, with distribution  $q_i$ .

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↔ Design parameters when Gaussian proposal

- ▶ Selection weights :  $\omega_1, \dots, \omega_d$ .
- ▶ Variances of the Gaussian proposals :  $\sigma_1^2, \dots, \sigma_d^2$ .

## Radial behavior of $\Delta$ (I)

$$\Delta_i(x) = \omega_i \int_{\{y \in \mathbb{R}, \pi(x+ye_i) < \pi(x)\}} y \left( \frac{\pi(x+ye_i)}{\pi(x)} - 1 \right) q_i(y) dy.$$

↔ For the target densities  $\pi$  in the class

- $\lim_{r \rightarrow +\infty} |\nabla \ln \pi(rx)| = +\infty.$

- $\ell$  given by  $\lim_{r \rightarrow +\infty} \frac{\nabla \ln \pi(rx)}{|\nabla \ln \pi(rx)|} = \ell(x)$  is continuous.

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↔ As  $r \rightarrow +\infty$

$$\Delta_i(rx) \longrightarrow \text{sign}(\ell_i(x)) \omega_i \int_{\mathbb{R}^+} y q_i(y) dy = \text{sign}(\ell_i(x)) \frac{\omega_i \sigma_i}{\sqrt{2\pi}}.$$

## Radial behavior of $\Delta$ (II)

$$\Delta_i(rx) \longrightarrow \text{sign}(\ell_i(x)) \frac{\omega_i \sigma_i}{\sqrt{2\pi}}.$$

↔ This implies

- ▶ The radial limit depends upon the design parameters through the product  $\omega_i \sigma_i$ .
- ▶ The radial limit is constant on the sets

$$O_\alpha = \{x, \text{sign}(\ell(x)) = \gamma_\alpha\}$$

where  $\gamma_\alpha \in \{-1, 1\}^d$ .

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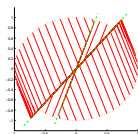
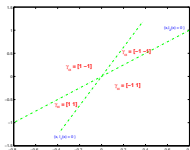
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↪ Example : RW MwG,  $\pi \sim \mathcal{N}_2(0, \Gamma)$

$$\implies \ell(x) = -\frac{\Gamma^{-1}x}{|\Gamma^{-1}x|}.$$





## Piecewise Linear Fluid limits

↔ Linear till the first time it reaches  $\partial[\cup_{\alpha=1}^a O_\alpha]$

▶  $\forall x \in O_\alpha,$

$$\forall t \leq T(x) \quad \eta(t) = x + t \gamma_\alpha \circ \omega \circ \sigma, \quad \mathbb{Q}_x - \text{a.s.}$$

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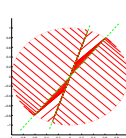
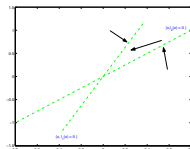
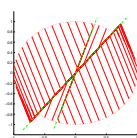
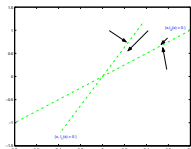
where  $T(x)$  : hitting-time of  $\partial\mathcal{O}_\alpha$ .

↪ Attractive boundaries

(Results in the case :  $d = 2$  and boundaries are hyperplanes)

▶ If the reached boundary is “attractive”, the fluid limit is trapped on the boundary.

▶ Example : RW MwG,  $\pi \sim \mathcal{N}_2(0, \Gamma)$

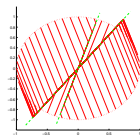
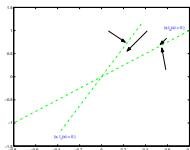


## Stability of the fluid limit

↔ Stable attractive boundaries

(Results in the case :  $d = 2$  and boundaries are hyperplanes)

► The fields in the neighborhood of the boundaries, “point” towards the origin.



► Example : RW MwG,  $\pi \sim \mathcal{N}_2(0, \Gamma)$ .  
Any attractive boundary is stable.

## Adaptive strategies : state-dependent design parameters

Since the fluid limit depends upon the design parameters through  $\omega_i \sigma_i$ ,

**Strategy 1.** Fix  $\omega_i = 1/d$  and choose  $\sigma_i(x)$ .

**Strategy 2.** Fix  $\sigma_i = c$  and choose  $\omega_i(x)$ .

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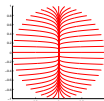
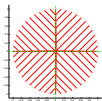
$$[\omega_i \sigma_i](x) = c |\ell_i(x)| \quad \ell_i(x) = \lim_r \frac{\nabla_i \ln \pi(rx)}{|\nabla \ln \pi(rx)|}.$$

so that in both strategies, the fluid limit  $\longleftrightarrow$  ODE  $\dot{\mu} = h(\mu)$  with

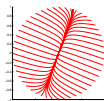
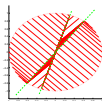
$$h(x) = \frac{c}{\sqrt{2\pi}} \ell(x) \quad \ell_i(x) = \lim_r \frac{\nabla_i \ln \pi(rx)}{|\nabla \ln \pi(rx)|}.$$

## Ex. : Fluid limits [left] non-adaptive [right] adaptive

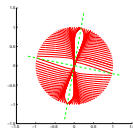
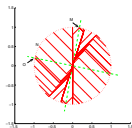
► When  $\pi \sim \mathcal{N}_2(0, \Gamma_1)$   $\Gamma_1$  diagonal



► When  $\pi \sim \mathcal{N}_2(0, \Gamma_2)$   $\Gamma_2$  non-diagonal



► When  $\pi \sim \mathcal{N}_2(0, \Gamma_1) + \mathcal{N}_2(0, \Gamma_2)$



# Assessing efficiency (I)

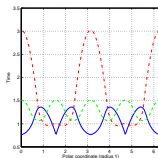
↪ **Criterion 1** : Based on the **Limit fluid** and on the time the fluid limit started on the unit sphere, enters a ball of radius  $\rho \in ]0, 1[$ .

## ► Example

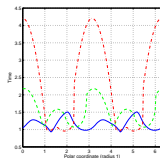
*x*-axes : polar coordinate of the initial value.

*y*-axes : hitting-time.

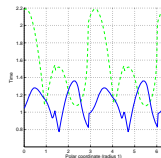
for the three algorithms **Adaptive strategy** **Non-Adaptive,  $\omega_1 = 0.25$**  **Non-Adaptive,  $\omega_1 = 0.5$**



$\pi \sim \mathcal{N}_2(0, \Gamma_1)$   $\Gamma_1$  diagonal



$\pi \sim \mathcal{N}_2(0, \Gamma_2)$   $\Gamma_2$  non diagonal



$\pi \sim \mathcal{N}_2(0, \Gamma_1) + \mathcal{N}_2(0, \Gamma_2)$

## Assessing efficiency (II)

↪ **Criterion 2** : Based on the **Markov chain** and the hitting-time of the “center of the space” when started “far” from the center.

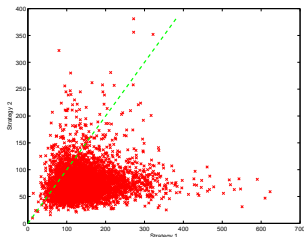
► **Example** :  $\pi \sim \mathcal{N}_8(0, \Gamma)$        $d = 8$

$\Gamma$  : diagonal, with entries  $\Gamma_{i,i} \sim \mathcal{E}(1)$ .

5000 adaptive chains, started from  $x \in \{z' \Gamma^{-1} z = d\}$ .

$x$ -axes : hitting-time of the ball of radius  $\sqrt{d}$  with the Strat 1 (adapt  $\sigma$ )

$y$ -axes : hitting-time of the ball of radius  $\sqrt{d}$  with the Strat 2 (adapt  $\omega$ )





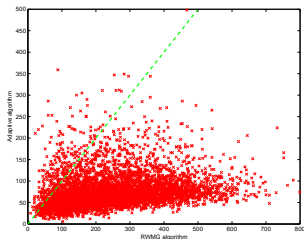
► Example :  $\pi \sim \mathcal{N}_8(0, \Gamma)$        $d = 8$

$\Gamma$  : diagonal, with entries  $\Gamma_{i,i} \sim \mathcal{E}(1)$ .

5000 adaptive chains, started from  $x \in \{z' \Gamma^{-1} z = d\}$ .

$x$ -axes : hitting-time of the ball of radius  $\sqrt{d}$  with the non-adaptive strategy

$y$ -axes : hitting-time of the ball of radius  $\sqrt{d}$  with the Strat 2 (adapt  $\omega$ )



## Conclusion

1. Normalisation : how does the chain behave when started far in the tails?

normalisation NOT as in Roberts et al. (1997), Roberts and Rosenthal (2001), Neal et al. (2007),

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### Talk based on the papers

- G. Fort, S. Meyn, E. Moulines and P. Priouret. The ODE method for the stability of skip-free Markov Chains with applications to MCMC. To be published, Ann. Appl. Probab. (2007)
- G. Fort. Fluid limit-based tuning of some hybrid MCMC samplers. Submitted (2007).