

Stochastic Representations

and

Navier-Stokes equations

IPAM Workshop: Transport Systems

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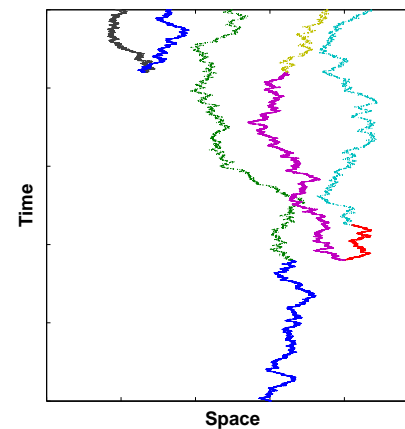
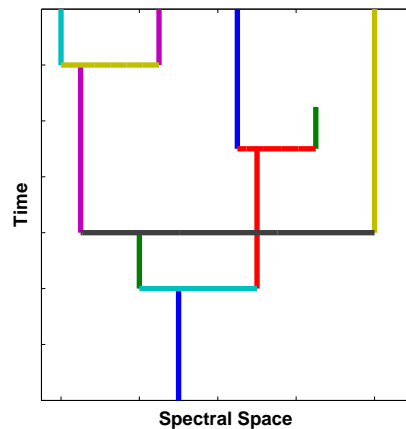
May 5, 2008

GOAL: Illustrate connection between

deterministic PDE's

$$\frac{\partial u}{\partial t} + u \cdot \nabla u = \nu \Delta u - \nabla p + g \quad \nabla \cdot u = 0 \quad u_0(x) = u(x, 0)$$

and stochastic processes



Portray solutions of some **partial differential equations** as **expectations of stochastic processes** with the aim of garnering insight into properties of solutions. Particular example examined: the Navier-Stokes equations in 3-d. Descriptions will be given of two types of representations. First: Fourier-transformed solutions as given by LeJan and Sznitman (1997). Second: physical space representations. Both give existence and uniqueness of solutions for all time for 'small' initial data and on short time intervals for 'large' initial data.

OUTLINE

Background and illustration

- two simple examples

Navier-Stokes: **an extra twist**

- Fourier space
- Physical space

Comments

ANTECEDENTS:

Skorohod (1964)

Ikeda, Watanabe, Dawson and others, 1968 and on

Savits (1969)

McKean (1975)

LeJan and Sznitman (1997)

Bhattacharya et al (2003)

OTHER DIRECTIONS:

Constantin (2001), Constantin and Iyer (2008), Iyer (2006)

Albeverio and Belopolskaya (2003), (2005), (2007)

Busnello, Flandoli, Romito (2005)

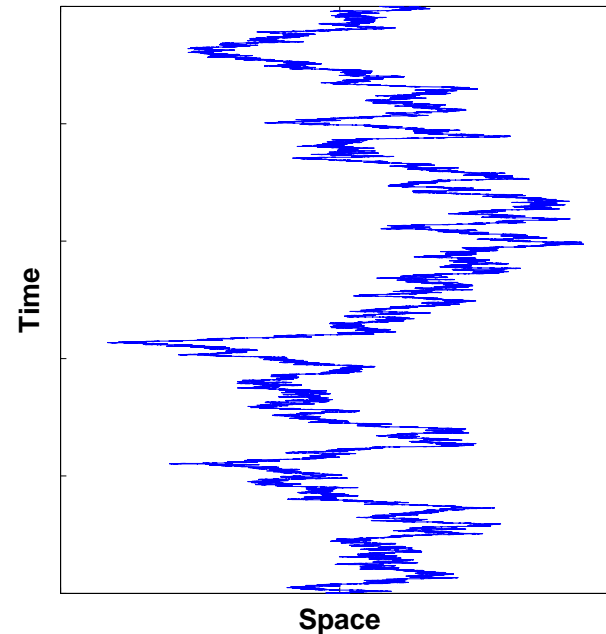
Fontana (2006)

Birnir (2007)

Brzezniak and Neklyudov (2008)

EXAMPLE : The heat equation and Brownian motion

$$\frac{\partial u}{\partial t} = \frac{1}{2} \Delta u$$
$$u(x, 0) = f(x)$$



Brownian motion W

$$u(x, t) = \int f(x - y) K(y, t) dy \quad K(y, t) = (2\pi t)^{-d/2} \exp^{-|y|^2/2t}$$

$$u(x, t) = E_x f(W(t))$$

Fourier transform:

$$\hat{u}(\xi, t) = (2\pi)^{-d/2} \int_{\mathbb{R}^d} e^{-i\xi \cdot x} u(x, t) dx, \quad \xi \in \mathbb{R}^d$$

$$\frac{\partial \hat{u}}{\partial t} = -\frac{|\xi|^2}{2} \hat{u}$$

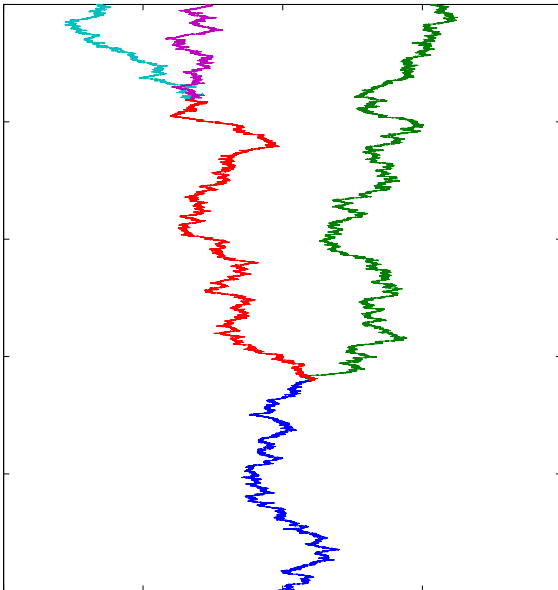
$$\hat{u}(\xi, t) = \hat{f}(\xi) e^{-|\xi|^2 t/2} = \hat{f}(\xi) P(S_\xi > t)$$

$e^{-|\xi|^2 t/2}$ is the Fourier transform of $K(y, t)$

$$u(x, t) = \int f(x - y) K(y, t) dy = E_x f(W(t))$$

KPP reaction-diffusion equation

$$\frac{\partial u}{\partial t} = \frac{1}{2} \frac{\partial^2 u}{\partial x^2} + u^2 - u \quad \text{with} \quad u(x, 0) = f(x)$$



McKean(1975): If $|f(x)| \leq 1$

$$u(x, t) = E_x \prod_{i \leq N(t)} f(W^{(i)}(t))$$

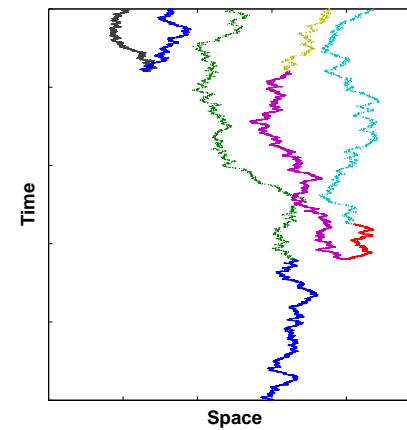
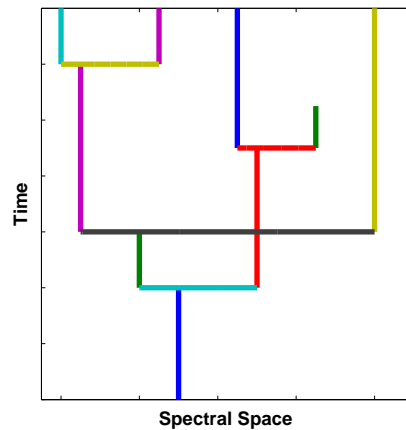
Markov process W : branching Brownian motion

Omnibus idea:

- Mild formulation
- Recognize transition probabilities

USE IDEA TO: Connect

random processes:



solutions to the Navier-Stokes equations:

$$\frac{\partial u}{\partial t} + u \cdot \nabla u = \nu \Delta u - \nabla p + g \quad \nabla \cdot u = 0 \quad u_0(x) = u(x, 0)$$

Navier-Stokes equations in 3-d

- $\frac{\partial u}{\partial t} + u \cdot \nabla u = \nu \Delta u - \nabla p + g$, **pressure p and forcing g**
- **incompressible fluids:** $\nabla \cdot u = 0$
- **initial data:** $u_0(x) = u(x, 0)$
- **mild solutions:**
$$u = e^{\nu t \Delta} u_0 - \int_0^t e^{\nu(t-s)\Delta} \mathcal{P} \nabla \cdot (u \otimes u)(s) ds + \int_0^t e^{\nu(t-s)\Delta} \mathcal{P} g(s) ds$$

\mathcal{P} : Leray projection onto divergence free vector fields

LeJan and Sznitman (1997): Fourier NS

$$\begin{aligned}\hat{u}(\xi, t) = & e^{-\nu|\xi|^2 t} \hat{u}_0(\xi) + \int_{s=0}^t (2\pi)^{-3/2} e^{-\nu|\xi|^2 s} \\ & \left\{ -i\mathbf{P}_\xi \int \hat{u}(\eta, t-s) \xi \cdot \hat{u}(\xi - \eta, t-s) d\eta \right. \\ & \left. + \mathbf{P}_\xi \hat{g}(\xi, t-s) \right\} ds\end{aligned}$$

Theorem: (LJ-S, 1997) Suppose that

$$\sup_{\xi} |\xi|^2 |\hat{u}_0(\xi)| < (2\pi)^{3/2} \nu / 2 \quad \text{and} \quad \sup_{\xi \neq 0, t \geq 0} 2|\hat{g}(\xi, t)| \leq (2\pi)^{3/2} \nu / 4.$$

Then there is a unique mild solution $\{u(x, t) : x \in \mathbf{R}^3, t \geq 0\}$ to the Navier-Stokes equations with

$$\sup_{\xi \neq 0, t \geq 0} |\xi|^2 |\hat{u}(\xi, t)| \leq (2\pi)^{3/2} \nu.$$

KEY: Weight \hat{u} by $|\xi|^2$

The crucial magical realization:

$$\text{for } h(\xi) = |\xi|^{-2}, \quad h * h(\xi) = c |\xi| h(\xi),$$

This kernel is used to give Markov spatial transition probability

$$dH_\xi(\eta) = \frac{h * h(\xi)}{|\xi| h(\xi)} = \frac{c|\xi|}{|\eta|^2 |\xi - \eta|^2}$$

\mathbf{P}_ξ projection matrix

$$w \otimes_\xi z = -i(e_\xi \cdot z) \mathbf{P}_\xi w$$

Rewriting the Fourier NSE in terms of

$$\chi(\xi, t) = |\xi|^2 \hat{u}(\xi, t) \quad \text{and} \quad \phi(\xi, t) = \frac{2}{\nu} \hat{g}(\xi, t)$$

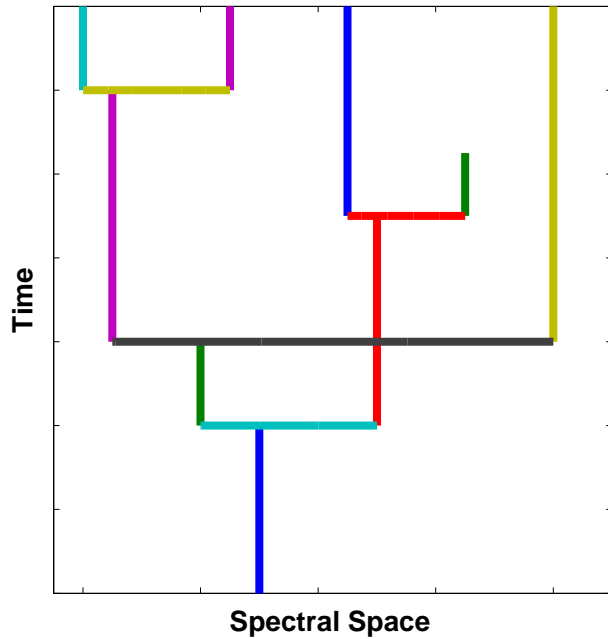
$$\begin{aligned} \chi(\xi, t) = & e^{-\nu t |\xi|^2} \chi_0(\xi) + \int_0^t \nu |\xi|^2 e^{-\nu |\xi|^2 s} \\ & \left\{ \frac{1}{2} c_\nu \int_{\mathbf{R}^3} \chi(\eta, t-s) \otimes_\xi \chi(\xi - \eta, t-s) dH_\xi(\eta) \right. \\ & \left. + \frac{1}{2} \phi(\xi, t-s) \right\} ds \end{aligned}$$

Conditional expectation:

$$\begin{aligned} \chi(\xi, t) = & \mathbf{E}_\xi \left(\chi_0(\xi) \mathbf{1}[S_\xi > t] \right. \\ & + \left\{ \sigma_\xi c_\nu \chi(\eta, t - S_\xi) \otimes_\xi \chi(\xi - \eta, t - S_\xi) \right. \\ & \left. \left. + (1 - \sigma_\xi) \phi(\xi, t - S_\xi) \right\} \mathbf{1}[S_\xi \leq t] \right) \end{aligned}$$

Motivates formulation of stochastic process and iterative functional

$$\Xi(\xi, t) = \begin{cases} \chi_0(\xi) & S > t \\ c_\nu \Xi(\eta, t - S) \otimes_\xi \Xi(\xi - \eta, t - S) & S \leq t, \sigma = 1 \\ \phi(\xi, t - S_\xi) & S \leq t, \sigma = 0 \end{cases}$$



$$\Xi(\xi, t) = \begin{cases} \chi_0(\xi) \\ c_\nu \Xi \otimes_\xi \Xi \\ \phi(\xi, t - S_\xi) \end{cases}$$

Markov walk on tree

PROPOSITION: If $\mathbf{E}_\xi |\Xi(t)| < \infty$, then

$$\hat{u}(\xi, t) = c_\nu |\xi|^{-2} \mathbf{E}_\xi \Xi(t)$$

provides a mild solution to the Navier-Stokes equations.

Non-negative Fourier multiplier majorizing kernel:

$$h * h(\xi) \leq c|\xi|h(\xi)$$

With $\chi = \frac{\hat{u}}{h}$ and $\phi = \frac{|\xi|^2 \hat{g}}{h}$, the probabilistic construction gives:

PROPOSITION: If $\mathbf{E}_\xi |\Xi(t)| < \infty$, then $\hat{u}(\xi, t) = c_\nu h(\xi) \mathbf{E}_\xi \Xi(t)$ provides a mild solution to the Navier-Stokes equations.

THEOREM(BCDGOOTW, 2004): If both

$$\sup_{\xi} \frac{|\hat{u}_0(\xi)|}{h(\xi)} \leq c_\nu \quad \text{and} \quad \sup_{\xi, t} \frac{|\hat{g}(\xi, t)|}{|\xi|^2 h(\xi)} \leq c_\nu,$$

then there is a unique mild solution $\{u(x, t) : x \in \mathbf{R}^3, t \geq 0\}$ to the Navier-Stokes equations with

$$\sup_{\xi \neq 0, t \geq 0} h(\xi) |\hat{u}(\xi, t)| \leq c_\nu.$$

Fourier multiplier majorizing kernels in \mathbf{R}^3 : $h * h(\xi) \leq |\xi|h(\xi)$

Le-Jan Snitzman kernels:

$$h_0(\xi) = \frac{c}{|\xi|^2}, \quad h_1(\xi) = \frac{ce^{-\alpha|\xi|}}{|\xi|}$$

A couple of BCDGOOTW kernels:

$$h_{\alpha,\beta}(\xi) = c|\xi|^{\beta-2}e^{-\alpha|\xi|^\beta}, \quad \alpha > 0, \quad \beta \in [0, 1]$$

Here $0 < \beta < 1$ allows smooth compactly supported initial data.

$$h(\xi) = c_\nu \int_{t>0} (\pi t)^{-3} \prod_{j=1}^3 \left(\frac{1}{1 + (\frac{\xi_j}{t})^2} \right) dt$$

Note: the class of majorizing kernels is log-convex.

Majorizing kernel formulation:

Go to a scalar comparison equation

$$U(\xi, t) = e^{-\nu|\xi|^2 t} U_0(\xi) + c \int_{s=0}^t |\xi| e^{-\nu|\xi|^2 s} U * U(\xi, t - s) ds$$

Comparison equation: set $U(\xi, t) = |\hat{u}(\xi, t)|$

The steady state solution:

$$U_\infty(\xi) = c|\xi|^{-1} U_\infty * U_\infty(\xi)$$

gives candidates for the majorizing kernels.

Comparison equation: Blomker, Romito, Tribe (2007)

Physical space analogue

$$\begin{aligned}
 u(x, t) &= \int_{\mathbf{R}^3} u_0(x - y) K(y, 2\nu t) dy \\
 &+ \int_{s=0}^t \int_{\mathbf{R}^3} \left\{ \left[\frac{|z|}{4\nu s} K(z, 2\nu s) \mathbf{b}_1(z; u(x - z, t - s), u(x - z, t - s)) \right. \right. \\
 &- \left. \left(\frac{1}{4\pi^2 \nu s |z|^4} \int_{\{y: |y| \leq |z|\}} |y|^2 K(y, 2\nu s) dy \right) \right. \\
 &\quad \left. \left. \mathbf{b}_2(z; u(x - z, t - s), u(x - z, t - s)) \right] \right. \\
 &+ \left. \left[K(z, 2\nu s) \mathbf{P}_z \right. \right. \\
 &- \left. \left. \frac{1}{4\pi |z|^3} \int_{\{y: |y| \leq |z|\}} K(y, 2\nu s) dy (I - 3e_z e_z^t) \right] g(x - z, t - s) \right\} dz ds
 \end{aligned}$$

$$K(y, t) = (2\pi t)^{-3/2} e^{-|y|^2/2t}$$

$$e_y = y/|y|, \quad \mathbf{P}_y = I - e_y e_y^t$$

$$\mathbf{b}_1(y; u, v) = (u \cdot e_y) \mathbf{P}_y v + (v \cdot e_y) \mathbf{P}_y u$$

$$\mathbf{b}_2(y; u, v) = (\mathbf{b}_1(y; u, v) + u \cdot (I - 3e_y e_y^t) v e_y) / 2$$

$$\mathbf{c}_1 = \mathbf{P}_z \quad \mathbf{c}_2 = (I - 3e_z e_z^t) / 2$$

Lemma:

$$|\mathbf{b}_k(y; u, v)| \leq |u||v| \quad \text{and} \quad \mathbf{b}_k(y; u, v) = h^2 \mathbf{b}_k(y; \frac{u}{h}, \frac{v}{h})$$

Scalar comparison steady state equation:

$$U(x) = c \int_{y \in \mathbf{R}^3} |y|^{-2} U^2(x - y) dy$$

Motivation for majorizing kernel:

$$h(x) \geq c \int_{y \in \mathbf{R}^3} |y|^{-2} h^2(x - y) dy$$

Introduce **majorizing kernel pairs** (h, \tilde{h}) :

$$(A) \quad \int_{\mathbf{R}^3} h^2(x-y)|y|^{-2}dy \leq h(x) \quad h \text{ scales } u, \text{ dominates } |u|$$

and

$$(\tilde{A}) \quad \int_{\mathbf{R}^3} \tilde{h}(x-y)|y|^{-1}dy \leq h(x) \quad \tilde{h} \text{ scales } g, \text{ dominates } |g|$$

$$\chi(x, t) = \frac{u(x, t)}{h(x)} \quad \text{scaled solution} \quad \chi_0(x) = \chi(x, 0) \quad \text{initial data}$$

$$\varphi(x, t) = \frac{g(x, t)}{\tilde{h}(x)} \quad \text{scaled forcing}$$

Building blocks of probabilistic representation:
spatial and temporal transition densities

spatial: built from the majorizing kernels and based on

$$\frac{h^2(x-y)|y|^{-2}}{\int_{\mathbf{R}^3} h^2(x-y)|y|^{-2} dy} \quad \text{iterative term}$$

and

$$\frac{\tilde{h}(x-y)|y|^{-1}}{\int_{\mathbf{R}^3} \tilde{h}(x-y)|y|^{-1} dy} \quad \text{forcing term}$$

The actual spatial transition densities:

$$f(y, z|x) = \frac{|y|^{-1}|z|^{-4}h^2(x-z)\mathbf{1}[|z| > |y|]}{2\pi \int_{\mathbf{R}^3} |z|^{-2}h^2(x-z)dz} \quad (\text{branch point})$$

$$\tilde{f}(y, z|x) = \frac{|y|^{-1}|z|^{-3}\tilde{h}(x-z)\mathbf{1}[|z| > |y|]}{2\pi \int_{\mathbf{R}^3} |z|^{-1}\tilde{h}(x-z)dz} \quad (\text{input forcing})$$

Temporal transition densities:

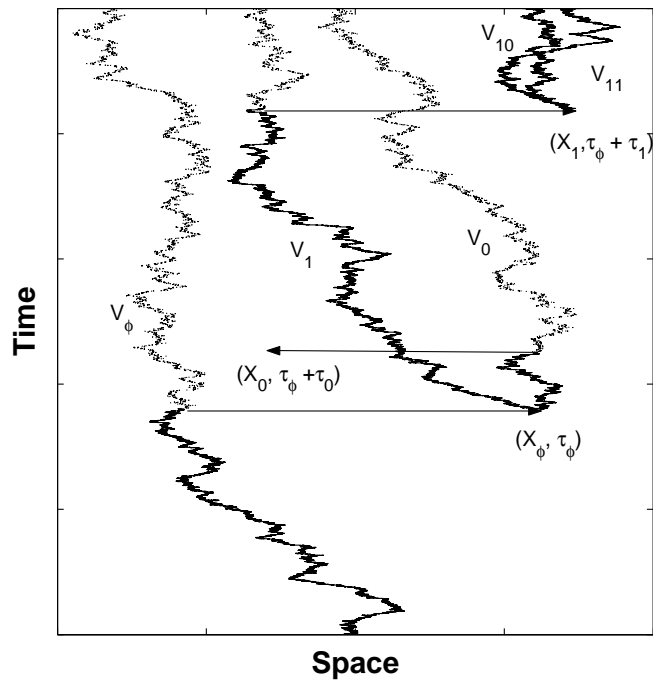
densities for waiting times between branches; conditional on spatial location

$$f_0(s|z) = c_\nu s^{-5/2} |z|^3 e^{-|z|^2/4\nu s}$$

$$f_1(s|y) = c_\nu s^{-3/2} |y| e^{-|y|^2/4\nu s}$$

If the kernel h is **excessive**

$$J(y, t|x) = \frac{h(y)K(x - y, 2\nu t)}{h(x)} \quad \text{h-Brownian motion}$$



- V_v 's h -Brownian motion
- X_v 's Markov branching rw
- τ_v 's waiting times
- σ_v 's 0 or 1

Stochastic iterative functional Υ on tree; start with leaves and work down branches:

$$m(X_{\bar{v}}) \mathbf{B}_v(\Upsilon_{v*0}(t - \tau_v), \Upsilon_{v*1}(t - \tau_v))$$

if $\sigma_v = 1, \tau_v \leq t$

$$\Upsilon_v(t) = \chi_0(V_v(t)) +$$

$$\tilde{m}(X_{\bar{v}}) \mathbf{c}_v \varphi(X_v, t - \tau_v) \quad \mathbf{if} \quad \sigma_v = 0, \tau_v \leq t$$

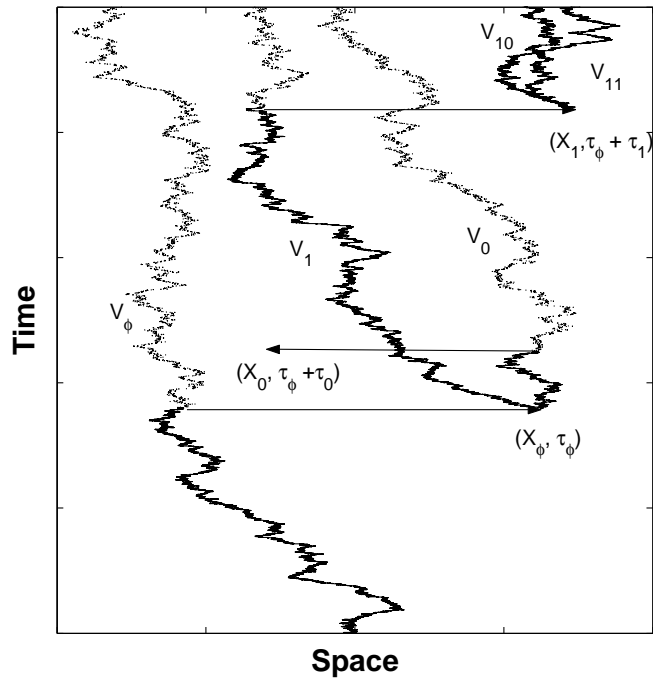
- V_v 's h -Brownian motion
 - X_v 's Markov branching rw
 - τ_v 's waiting times
 - σ_v 's 0 or 1
 - \mathbf{B}_v randomized \mathbf{b}_k 's
 - \mathbf{c}_v randomized:
- $\mathbf{P}_{Z_v} \quad \text{or} \quad - (I - 3e_{Z_v} e_{Z_v}^t) / 2$

Normalization multipliers:

$$m(x) = c_\nu \frac{\int |y|^{-2} h^2(x-y) dy}{h(x)}$$

$$\tilde{m}(x) = c_\nu \frac{\int |y|^{-1} \tilde{h}(x-y) dy}{h(x)}$$

Since (h, \tilde{h}) is a majorizing kernel pair, the magnitude of the multipliers is controlled.



PROPOSITION: If $\mathbf{E}_x |\Upsilon(t)| < \infty$, then

$$u(x, t) = h(x) \mathbf{E}_x \Upsilon(t)$$

provides a mild solution to the Navier-Stokes equations.

Theorem: If h is excessive, (h, \tilde{h}) is a majorizing kernel pair, and

$$\frac{|u_0(x)|}{h(x)} \quad \text{and} \quad \frac{|g(x, t)|}{\tilde{h}(x)}$$

are both uniformly small enough, then

$$u(x, t) = h(x)E_x \Upsilon(t)$$

is an unique mild solution to the NS equations for all time and bounded in magnitude by $c_\nu h(x)$ for each x and all t .

Proof of theorem: Easy contraction argument by induction on **finite binary tree*** shows that $|\Upsilon|$ is uniformly bounded if the magnitudes of both χ_0 and φ are uniformly small enough. It follows that $\chi(x, t) = E_x \Upsilon_\emptyset(t)$ exists and is uniformly bounded in magnitude. Uniqueness takes a little bit more. \square

finite with probability 1!

Regularity (no forcing): Ladyzhenskaya-Prodi-Serrin condition;
If a solution u to the NS equations satisfies

$$\int_0^T \left(\int |u(x, t)|^q dx \right)^r dt < \infty$$

for some $q > 3, r \geq r(q) > 0$, then u is regular on $[0, T]$.

Take a majorizing kernel $h(x)$ proportional to

$$\frac{1}{(1 + |x|)^{4(2-\gamma)}} \quad \text{for any } \gamma \in (3/2, 13/8)$$

Corollary: If $g = 0$ and for some $\gamma \in (3/2, 13/8)$

$$\sup_x (1 + |x|)^{4(2-\gamma)} |u_0(x)|$$

is finite, then there exists a unique regular solution to the NS equations for $t \leq T(\gamma)$.

Proof: Modifies probability model somewhat since these h 's are not excessive.

Comments:

- Short time existence for large initial data: speed up branching in time
- Branching is finite with probability 1; controls 'blow-up'
- Incompressibility is partially captured in the transition probabilities
- Family of branching diffusions connected to physical process

