## DYNAMICS IN RANDOM TIME

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The processes that are used as time-change are generally subordinators, or inverse subordinators. Subordinators are non-decreasing Levy processes, i.e., processes with independent and stationary increments having non-decreasing sample paths.
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If $S(t)$ is a subordinator, then we have

$$
E\left[e^{-\lambda S(t)}\right]=e^{-t \Phi(\lambda)},
$$

where $\Phi$ is called the Laplace exponent

## Inverse subordinators

For a subordinator $S(t)$, the first-exit time process is defined by

$$
E(t)=\inf _{s \geq 0}\{S(s)>t\}
$$

and we call this process the inverse subordinator. Note that

$$
P(E(t)>x)=P(S(x) \leq t)
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For every jump of the subordinator $S(t)$ there is a corresponding flat period of its inverse $E(t)$. These flat periods represent trapping events in which the test particle gets immobilized in a trap. Trapping slows down the overall dynamics of the process.

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## Markov case

Let $\mathcal{L}$ be a generic heuristic Markov generator defined on functions $u_{0}(x, t), t>0, x \in \mathbb{R}^{d}$. Consider the evolution equations of the following type

$$
\begin{cases}\frac{\partial u_{0}(x, t)}{\partial t} & =\left(\mathcal{L} u_{0}\right)(x, t)  \tag{1}\\ u_{0}(x, 0) & =\xi(x)\end{cases}
$$

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which we assume a solution $u_{0}(x, \cdot) \in L^{1}\left(\mathbb{R}_{+}\right)$is known.
We are interested in studying the subordination of the solution $u_{0}(x, t)$ by the density $G_{t}(\tau)$ of the inverse subordinator $E(t)$, that is the function $u(x, t)$ defined by

$$
\begin{equation*}
u(x, t):=\int_{0}^{\infty} u_{0}(x, \tau) G_{t}(\tau) d \tau, \quad x \in \mathbb{R}^{d}, t \geq 0 \tag{2}
\end{equation*}
$$

If $X(t)$ be the MP for generator $\mathcal{L}$ and $Y(t)=X(E(t))$ be time changed process then

$$
\begin{aligned}
& u_{0}(x, t)=E^{x}[\xi(X(t))] \\
& u(x, t)=E^{x}[\xi(Y(t))]
\end{aligned}
$$

The subordination principle tells that $u(x, t)$ is the solution of the general fractional differential equation

$$
\begin{cases}\left(\mathbb{D}_{t}^{(k)} u\right)(x, t) & =(\mathcal{L} u)(x, t)  \tag{3}\\ u(x, 0) & =\xi(x)\end{cases}
$$

with the same operator $\mathcal{L}$ acting in the spatial variables $x$ and the same initial condition $\xi$. Here $\mathbb{D}_{t}^{(k)}$ is a generalized fractional (convolutional) derivative corresponding to $E(t)$ cf. Kochubei, Toaldo, ...

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## Dynamical systems as Markov processes

Any DS can be considered as a deterministic MP.
Then all notions related to MP may be applied to DS.
Our aim is to demonstrate this observation.

## DS and Liouville equations

Let $X(t, x), t \geq 0, x \in \mathbb{R}^{d}$ be a dynamical system in $\mathbb{R}^{d}$ starting from $x$, that is, $X(0, x)=x$. This system is also a deterministic Markov process. Given $f: \mathbb{R}^{d} \longrightarrow \mathbb{R}$ we define

$$
u(t, x):=f(X(t, x))
$$

Then we have a version of Kolmogorov equation which is called the Liouville equation in the theory of dynamical systems:

$$
\frac{\partial}{\partial t} u(t, x)=L u(t, x)
$$

where $L$ is the generator of a semigroup.

## Example

## Consider DS

$$
\frac{d X(t)}{d t}=V(X(t))
$$

where $V \in T \mathbb{R}^{d}$ is a vector field.
Liouville operatror

$$
L f(x)=<V(x), \nabla f(x)>=\sum_{k=1}^{d} V_{k}(x) \nabla_{k} f(x)
$$

Liouville equation

$$
\frac{\partial u(t, x)}{\partial t}=<V(x), \nabla u(t, x)>
$$

that is so-called transport equation.

## Random times in DS

Assume we have a random time $E(t)$ as an inverse subordinator. Define a random dynamical systems

$$
Y(t, x, \omega)=X(E(t, \omega), x, \omega)
$$

For suitable functions $f: \mathbb{R}^{d} \longrightarrow \mathbb{R}$ define

Then $v(t, x)$ is the solution to an evolution equation with the same generator $L$ but with generalized fractional derivative, namely


As a result the following subordination formula holds:


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v(t, x):=E[f(Y(t, x))] .
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As a result the following subordination formula holds:

$$
\begin{equation*}
v(t, x)=\int_{0}^{\infty} u(\tau, x) G_{t}(\tau) \mathrm{d} \tau \tag{4}
\end{equation*}
$$

The problem (as in the Markov case) is to see the change of the behavior of $u$ after subordination.

## Example

Consider the simplest evolution equation in $\mathbb{R}^{d}$

$$
\frac{\mathrm{d}}{\mathrm{~d} t} X(t)=v \in \mathbb{R}^{d}, \quad X(0)=x_{0} \in \mathbb{R}^{d}
$$

The corresponding dynamics is

$$
X(t)=x_{0}+v t, \quad t \geq 0
$$

We consider $x_{0}=0$ just for simplicity. Assume that the certain assumptions on $E$ are satisfied. Take $f(x)=e^{-\alpha|x|}, \alpha>0$. The corresponding solution to the Liouville equation is

$$
u(t, x)=e^{-\alpha t|v|}, \quad t \geq 0
$$

Using properties of densities of certain class of inverse subordinators we obtain

$$
v(t, x) \sim \frac{1}{\alpha|v| \Gamma(\gamma)} t^{\gamma-1} Q(t), \quad t \rightarrow \infty
$$

In particular, for the $\alpha$-stable subordinator considered we obtain
$v(t, x) \sim C t^{-\alpha}, C$ is a constant.

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## Example

For $d=1$ consider the dynamics

$$
\alpha \frac{\mathrm{d}}{\mathrm{~d} t} X(t)=\frac{1}{X^{\alpha-1}(t)}, \quad \alpha \geq 1 .
$$

It is clear that the solution is given by

$$
X(t)=(t+C)^{1 / \alpha} .
$$

Take the function $f(x)=\exp \left(-a|x|^{\alpha}\right), a>0$, then the long time behavior of the subordination $v(t, x)$ is given by

$$
v(t, x) \sim \frac{e^{-a C}}{a} \frac{t^{\gamma-1}}{\Gamma(\gamma)} Q(t), \quad t \rightarrow \infty
$$

## Potentials for DS

In our framework above for a function $f: \mathbb{R}^{d} \rightarrow \mathbb{R}$ consider the solution to the Cauchy problem

$$
\begin{aligned}
\frac{\partial}{\partial t} u(t, x) & =L u(t, x), \\
u(0, x) & =f(x)
\end{aligned}
$$

Then as above

$$
u(t, x)=\left(e^{t L} f\right)(x)
$$

Define a potential for the function $f$ as

$$
U(f, x)=\int_{0}^{\infty} u(t, x) d t=\int_{0}^{\infty}\left(e^{t L} f\right)(x) d t=-\left(L^{-1} f\right)(x)
$$

We would like to have an integral representation

$$
U(f, x)=\int_{\mathbb{R}^{d}} f(y) \mu^{x}(d y)
$$

with a Radon measure on $\mathbb{R}^{d}$. This measure we will call the Green measure for our dynamical system. The existence of this measure is a non-trivial question for each particular model.

## Trajectories of random DS

Now we will analyze the transformation of trajectories of dynamical systems under random times. As above we have the Liouville equation for

$$
u(t, x):=f(X(t, x)), \quad t \geq 0, x \in \mathbb{R}^{d}
$$

that is,

$$
\frac{\partial}{\partial t} u(t, x)=L u(t, x), \quad u(0, x)=f(x)
$$

where $L$ is the generator of a semigroup.

In addition, let $E(t), t \geq 0$ be the inverse subordinator process, then we can consider the time changed random dynamical systems

$$
Y(t, x)=X(E(t), x), \quad t \geq 0, x \in \mathbb{R}^{d}
$$

Define

$$
v(t, x):=E[f(Y(t, x)] .
$$

The subordination formula gives

$$
v(t, x)=\int_{0}^{\infty} u(\tau, x) G_{t}(\tau) \mathrm{d} \tau
$$

Now we will take the vector-function

$$
f(x)=x \in \mathbb{R}^{d} .
$$

Then the average trajectories of $Y(t, x)$ is given by

$$
\int_{0}^{\infty}(X(\tau, x)) G_{t}(\tau) \mathrm{d} \tau=\int_{0}^{\infty} X(\tau, x) G_{t}(\tau) \mathrm{d} \tau
$$

If we consider the dynamical system of Example 1, that is, $X(t, x)=v t$, then we obtain

$$
\mathbb{E}[Y(t, x)]=v \int_{0}^{\infty} \tau G_{t}(\tau) \mathrm{d} \tau
$$

Computing the Laplace transform of each component yields

$$
v \mathcal{K}(p) \int_{0}^{\infty} \tau e^{-\tau p \mathcal{K}(p)} \mathrm{d} \tau=\frac{v p^{-2}}{\mathcal{K}(p)} \sim v p^{-(2-\gamma)} L\left(\frac{1}{p}\right)
$$

where $L(x)=(Q(x))^{-1}$ is a slowly varying function. It follows from Karamata-Feller Tauberian theorem. that

$$
\mathbb{E}[Y(t, x)] \sim v \frac{t^{1-\gamma}}{\Gamma(2-\gamma)} \frac{1}{Q(t)}
$$

## Diffusion approximation

Consider the following SDE in $\mathbb{R}^{d}$ :

$$
d X_{\varepsilon}(t)=v\left(X_{\varepsilon}\right) d t+\varepsilon d W(t)
$$

for each $\varepsilon>0$. The generator of this diffusion is

$$
L_{\varepsilon} f(x)=\varepsilon \Delta f(x)+<v(x), \nabla f(x)>.
$$

We can consider $X_{\varepsilon}$ as small perturbation of the DS

$$
d X(t)=v(X(t)) d t
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and the convergence $X_{\varepsilon} \rightarrow X$ under certain Lipschitz assumptions is well understood (Freidlin, Wentzel, Kutoyants,....). But the class of admissible vector fields $v(x)$ is quite restricted.

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An open problem is to have an info about transport equations

$$
d / d t u(t, x)=<v(x), \nabla u(t, x)>
$$

by means of detailed info about the heat type equations (Fokker-Planck)

$$
d / d t u_{\varepsilon}(t, x)=\varepsilon \Delta u(t, x)+<v(x), \nabla u(t, x)>.
$$

