

Spreading processes on networks

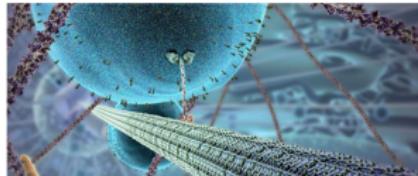
Liubov Tupikina, Marc Santolini

CRI, Bell labs
Big data course

November 26, 2019



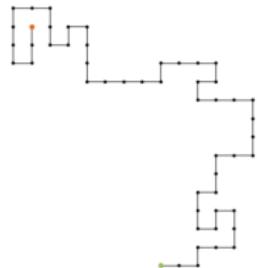
Spreading processes in nature



Transport of macromolecules, organelles and vesicles in living cells is a very complicated process that essentially determines and controls many biochemical reactions, growth and functioning of cells.



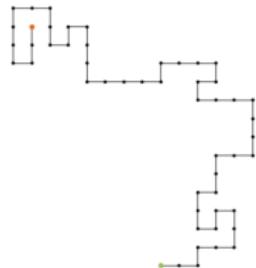
What we will talk about?



▶ **Part 1:**
Spreading processes and diffusion



What we will talk about?

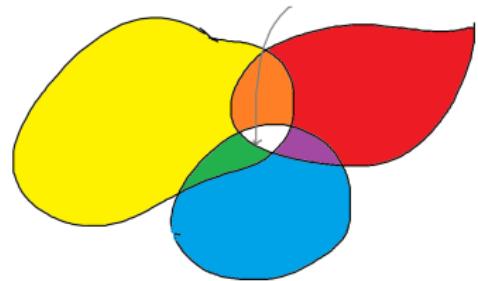


- ▶ **Part 1:**
Spreading processes and diffusion



- ▶ **Part 2:**
Spreading processes on networks

Scientists working on random walk theory:
Pearson, Smoluchowski,
Langevin, Fermat, Bernoulli,
Einstein, Kolmogorov, Pascal,
Hughes, Brown, Wiener,
Montroll, Weiss...

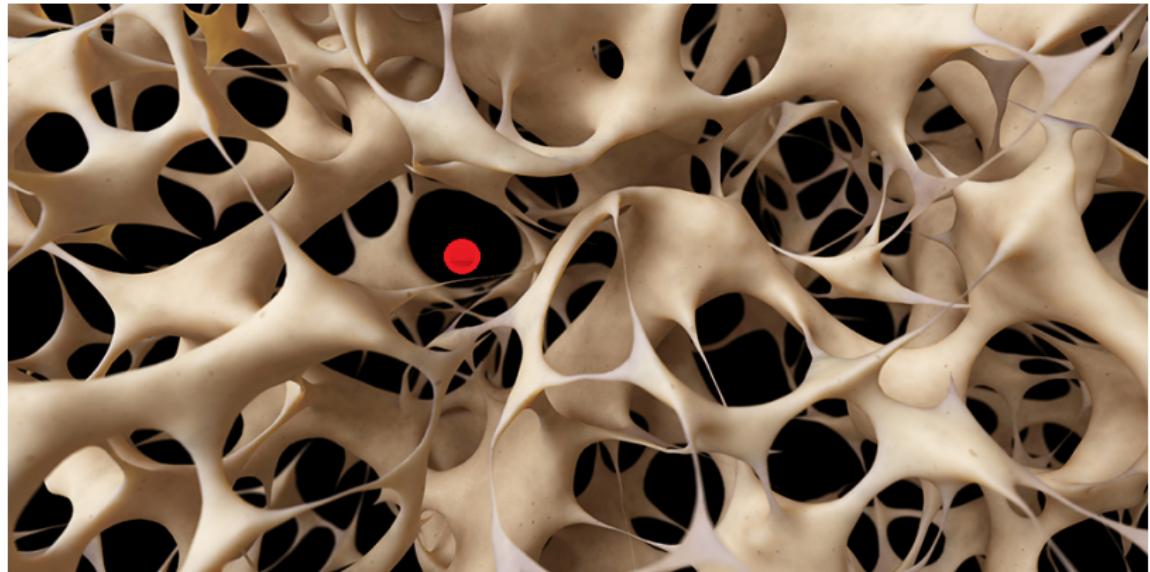


Intersection of Markov chain theory, Stochastic processes, Statistical physics, Disordered systems, Probability theory...

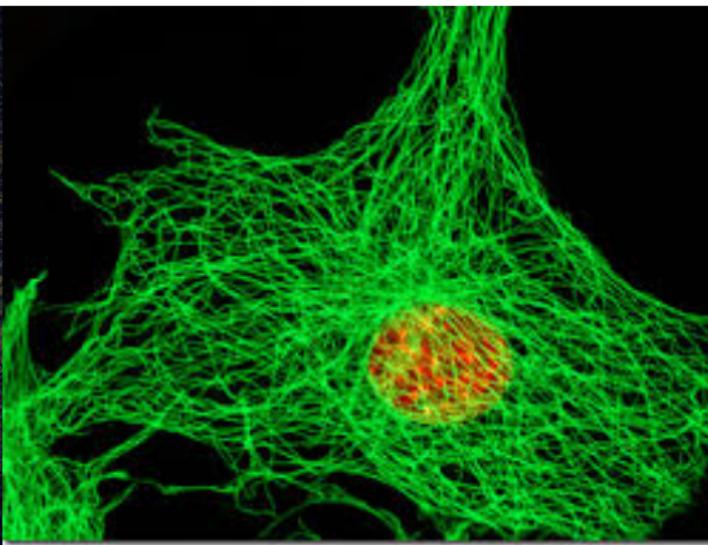
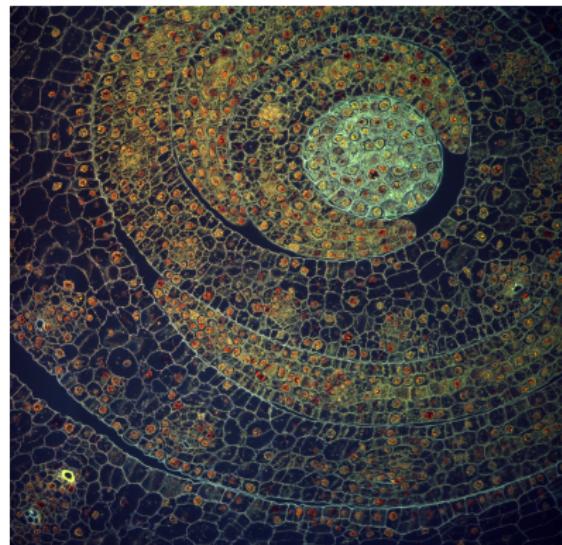
Real-world examples



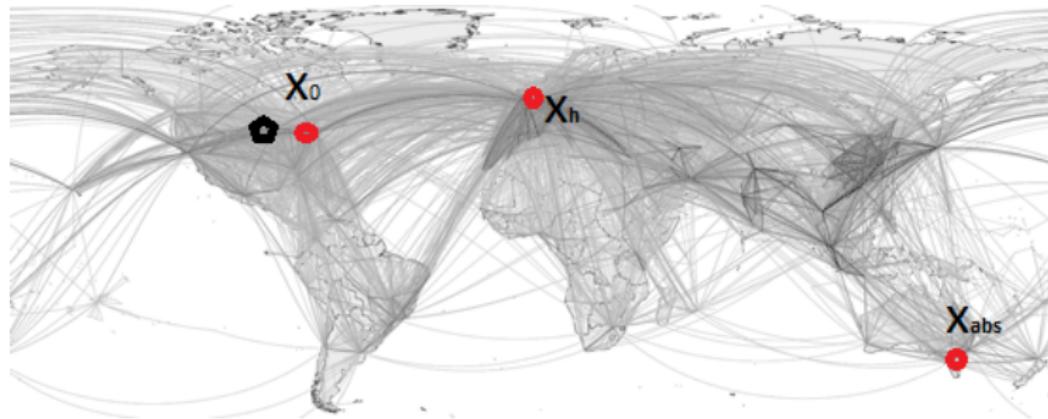
Real-world examples



Real-world examples



E.Katrukha, microtubules, imaginarycellrepresentation



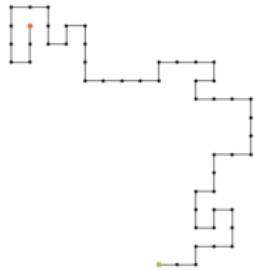
networkrepository/openflights



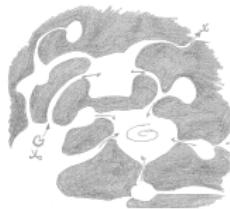


©BabyRoutes.co.uk

Part 1



- ▶ Spreading processes and diffusion history
- random walks
- diffusion models

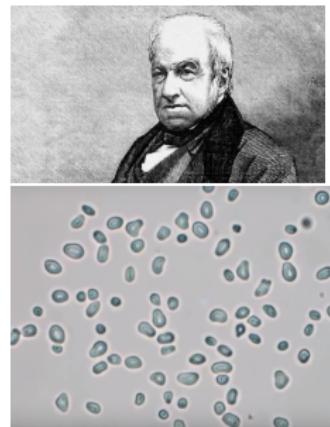


Spreading processes and diffusion: history

Experiment:

Brownian motion was discovered in 1785 by Jan Ingenhousz: irregular motion of coal dust particles on the surface.

In 1827 Robert Brown, a British botanist, is observing a suspended pollen grain in water.



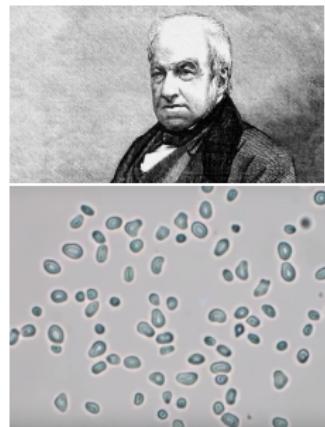
Spreading processes and diffusion: history

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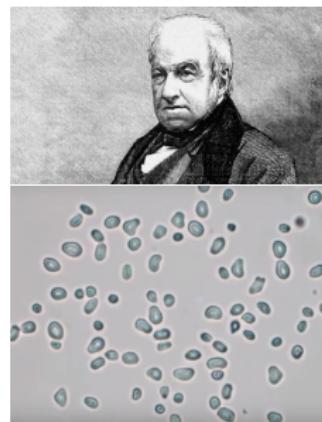
In 1827 Robert Brown, a British botanist, is observing a suspended pollen grain in water.
(video)

How to explain the experiment?



Random walks: How to explain the experiment?

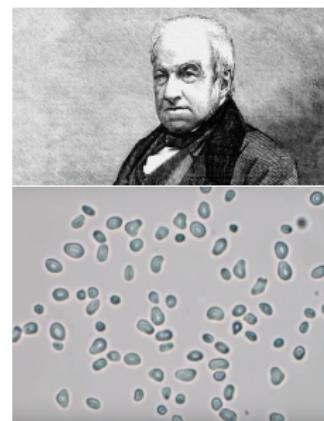
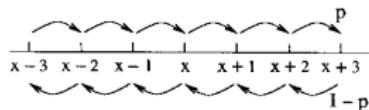
Assumptions:



Random walks: How to explain the experiment?

Assumptions:

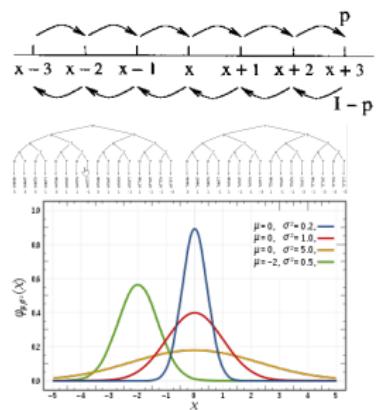
1. Random walk in 1 dimension
2. Memoryless: random walk forgets what he did last step
3. Symmetry: walk to the left of right with equal probability



Diffusion: building a model

Assumptions:

1. Random walk in 1 dimension
2. Memoryless: random walk forgets what he did last step
3. Symmetry: random walk jumps to the left or right with equal probabilities p, q

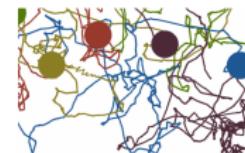
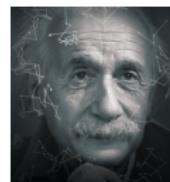


Examples of diffusion

In 1905 the paper "About particles suspended in the liquid(...)" appeared. Molecular motion was suggested the explanation.

Diffusion equations:

$$\frac{dT}{dt} = D \frac{d^2 T}{dx^2} \rightarrow D(x, t) \frac{d^2 T}{dx^2}$$



INVESTIGATIONS ON
THE THEORY OF THE
BROWNIAN MOVEMENT
BY
ALBERT EINSTEIN, PH.D.

This new Dover edition, first published in 1956, is an unabridged and unaltered re-creation of the translation first published in 1926. It is published through special arrangement with Methuen and Co., Ltd., and the estate of Albert Einstein.

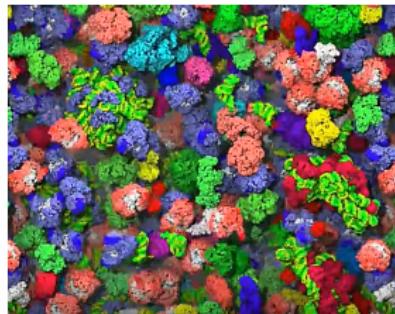
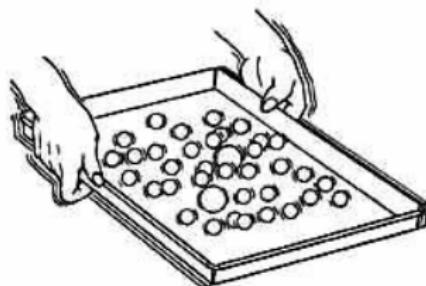
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of America.

EDITED WITH NOTES BY
R. FÜRTH

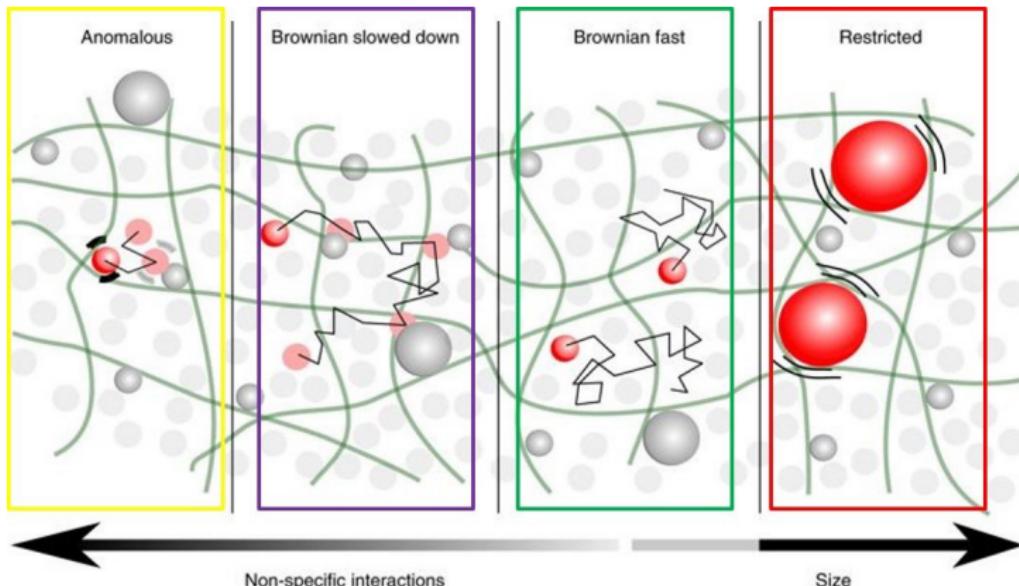
TRANSLATED BY
A. D. COWPER

WITH 3 DIAGRAMS

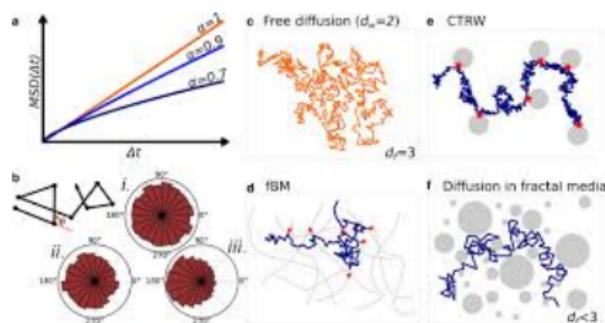
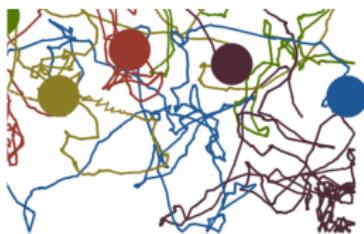
Applications of diffusion theory



- Simulations of the crowded cell environment [McGuffee, Elcock, 2010] (video)
- Estimation of the properties for biological cells (first passage times distribution for the medical substance, etc.) [Klafter, Sokolov, 2010], [Metzler, Chechkin et al.]



Applications of random walk theory



Modelling of the intracellular environment (R.Metzler, A.Cherstvyj, D.Grebenkov, A.Chechkin)

Random walks: What did we learn?

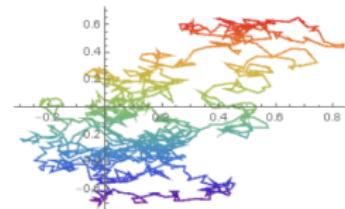
How to analyze motion inside
the heterogeneous media?

What is the full distribution of
the first passage time
probability?



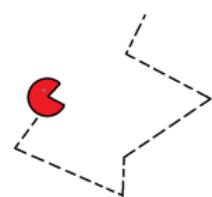
BrownianBridgeProcess

```
sample = RandomFunction[WienerProcess[], {0, 1, .001}, 2]["States"];
ListLinePlot[Transpose@sample, ColorFunction -> "Rainbow", ImageSize -> 300]
```



Types of random walks

- Continuous or discrete (networks) underlying system [Redner, 2002]
- Discrete or continuous time [Montroll, Weiss, 1965]
- Probability distributions of jumps from each node [Hedges, 2002]
- Random walks with resetting, self-avoiding random walk, adaptive random walks...
[Masuda, Lambiotte et al.2017]



Random walks properties

Probability density $P(x, t)$ is probability that we will find random walk in x at time t

First-passage time $F(x, t)$ is probability that at time t random walk is at x for the first time
[Metzler et al., Redner, 2002]

Network measures based on random walks

Check reversed classroom topics (something, which is NOT in wikipedia)



Random walks: discrete space and time

Master equation describing probability

$$P(x, n) = pP(x - 1, n - 1) + qP(x + 1, n - 1)$$

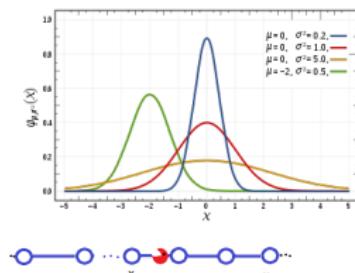
using the generating function and Fourier transformation

$$P(k, z) = \sum_{n=0}^{\infty} z^n \sum_{x=-\infty}^{\infty} e^{ikx} P(x, n)$$

$$\text{We get } P(k, z) = 1 / (1 - z(pe^{ik} + qe^{-ik}))$$

Then the **final probability distribution** for DTRW

$$P(x, n) = \frac{N!}{(N+x)/2!(N-x)/2!} p^{(N+x)/2} (1-p)^{(N-x)/2}$$



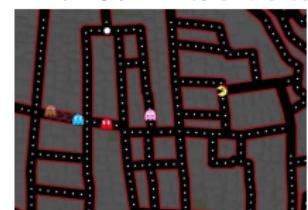
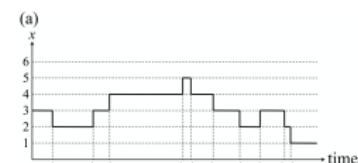
Gaussian in the long-time limit

$$P(x, n) \rightarrow \frac{1}{\sqrt{2\pi npq}} e^{-(x-np)^2/2npq}$$

Random walks: discrete space, continuous time

Continuous time random walk in 1D

At each x , where CTRW jumps, RW waits for τ distributed with $\psi(t)$.



[N.Masuda et al.
2017]

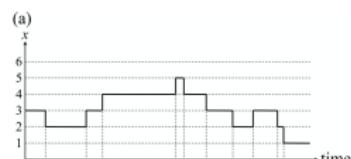
Random walks: discrete space, continuous time

Continuous time random walk in 1D

At each x , where CTRW jumps, RW waits for τ distributed with $\psi(t)$.

Probability that no event occurred till time t :

$$P^0(t) = \int_t^\infty \psi(t')dt'$$



[N.Masuda et al.
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Random walks: discrete space, continuous time

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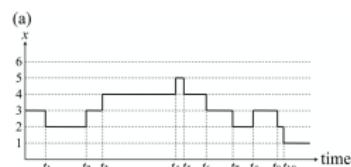
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$$P^1(t) = \int_t^\infty \psi(t')p(0, t - t')dt'$$



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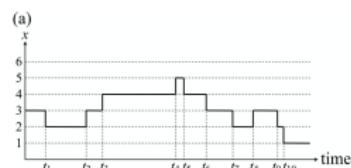
Probability that one event occurred till time t :

$$P^1(t) = \int_t^\infty \psi(t')p(0, t - t')dt'$$

Apply trick to time: $t \rightarrow s$

$$\tilde{P}^1(s) = \tilde{\psi}(s)\tilde{P}^0(s)$$

...



[N.Masuda et al.
2017]

Random walks: discrete space, continuous time

Continuous time random walk in 1D

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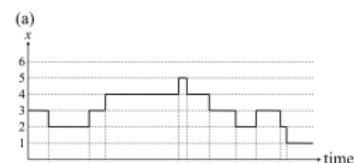
Apply trick to time: $t \rightarrow s$

$$\tilde{P}^1(s) = \tilde{\psi}(s) \tilde{P}^0(s)$$

...

Probability that n events occurred till time t :

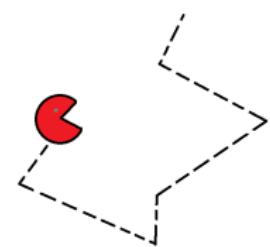
$$\tilde{P}^n(s) = \tilde{\psi}^n(s) \tilde{P}^{n-1}(s) = \tilde{\psi}^n(s) \frac{1 - \tilde{\psi}(s)}{s}$$



[N.Masuda et al.
2017]

Random walks: continuous space and time

Continuous time random walk model
(CTRW): CTRW waits at each x , where
CTRW jumps, it waits time step for τ
distributed with $\psi(t)$. The length of the steps
are distributed with $f(x)$ function.



Random walks: continuous space and time

Continuous time random walk model

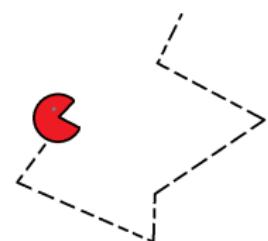
(CTRW) $P(x, t)$ is probability to find RW at x coordinate at time t .

We can use the probability $P(x, n)$ from discrete time random walk

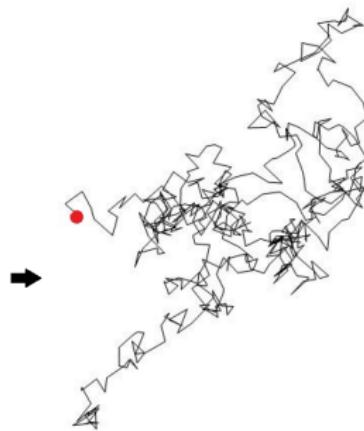
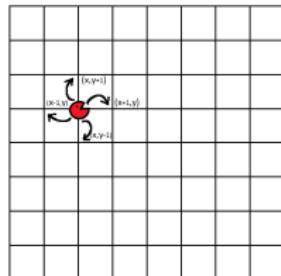
$$P(x, t) = \sum_{n=0}^{\infty} P(x, n)p(n, t)$$

Then using renewal theory, Green function we get the final equation

$$P(k, s) = \frac{1 - \tilde{\psi}(s)}{s} \frac{1}{1 - f(k)\tilde{\psi}(s)}$$



Generalization of random Walk framework



Homogeneous random walk
(RW) on a discrete lattice in
discrete time

R.Metzler, J.Klafter, Phys.Rep. [2000]

**Homogeneous continuous time
random walk (CTRW) in
continuous space**



P.Holme petterhol.me



With some probability or rate



Susceptible
meets
Infectious

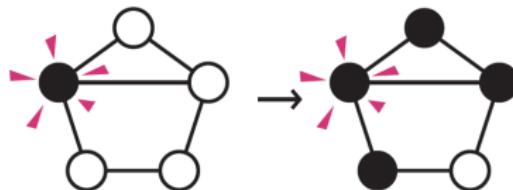
With some rate or after some time



Infectious

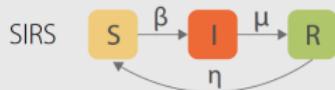


Susceptible or
Recovered



P.Holme, LT, NJP (2018)

SI (viral spreading [see notebook](#)), SIS, SIR model

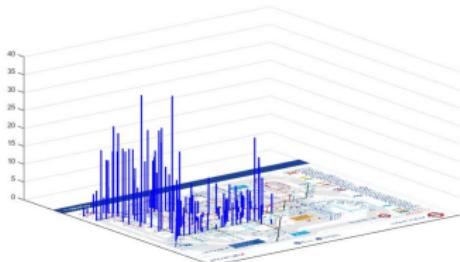
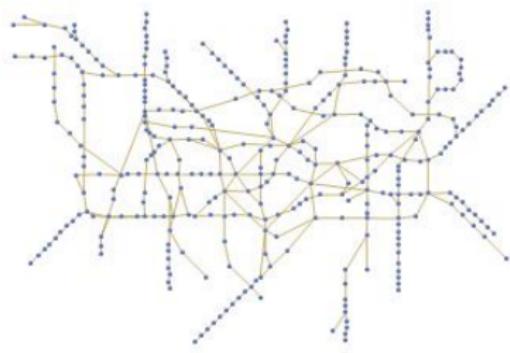


$$\frac{dS}{dt} = -\frac{\beta IS}{N},$$

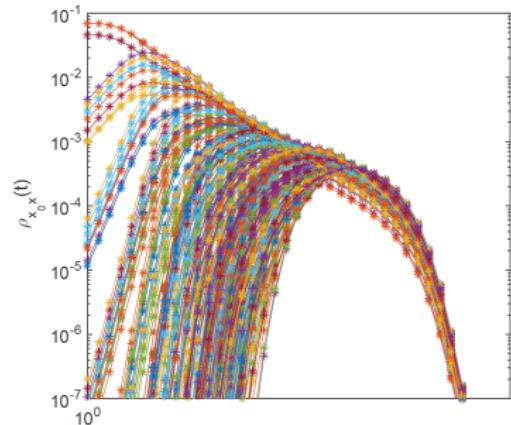
$$\frac{dI}{dt} = \frac{\beta IS}{N} - \gamma I,$$

$$\frac{dR}{dt} = \gamma I.$$

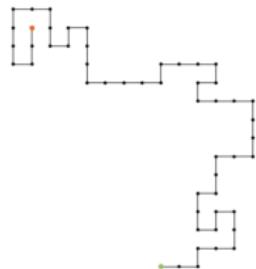
Applications: random walks or spreading on networks



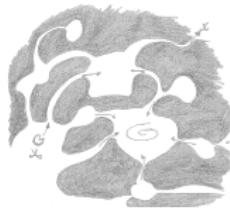
Random walks properties $\rho_{x_0 x}(t)$
on a network of London metro.
Dependence on the initial node x_0 .



What we will talk about?



- ▶ **Part 1:**
Spreading processes

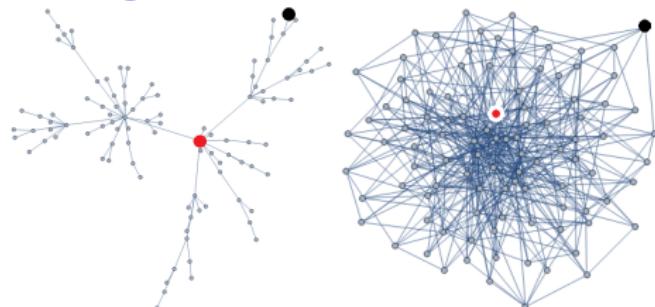


- ▶ **Part 2::**
Spreading processes on networks

- PageRank and other network measures
- Community detection algorithms
- Spreading processes on networks

Hands-on part

Looking at notebooks



For notebooks go to:

<https://github.com/Big-data-course-CRI/>

Applications of random walk theory

What happens if we violate the assumptions?

1. Random walk in $R \rightarrow$ in dimension R^N
2. Memoryless \rightarrow **with memory**: random walk remembers what he did last step
3. **Non-Symmetry**: random walk jumps to the left or right with different probabilities
4. New one??

Applications of random walk theory

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Fractal trajectories of random walk in 2D

Slide after all the maths

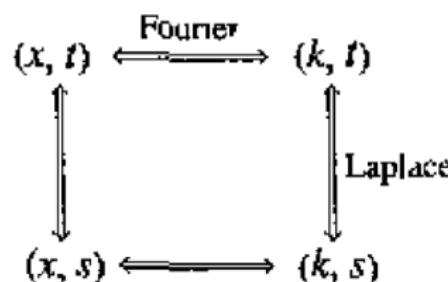
You did it!

Programs and articles on random walks, networks:
<https://github.com/Big-data-course-CRI/>

Main trick

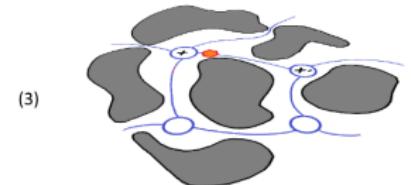
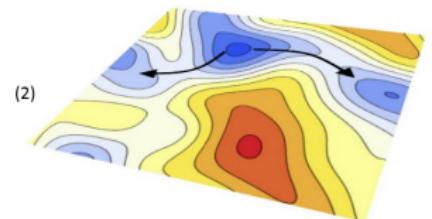
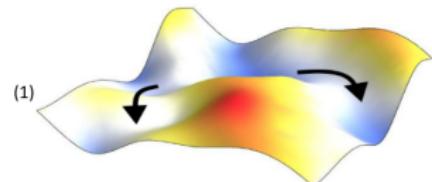
$$P(x, t) = \frac{e^{-(x-vt)^2/4Dt}}{\sqrt{4\pi Dt}}$$

$$P(k, t) = e^{-(ikv + Dk^2 t)}$$



$$P(x, s) = \frac{e^{-(v \mp \sqrt{v^2 + 4Ds}) |x|/2D}}{\sqrt{v^2 + 4Ds}}$$

$$P(k, s) = \frac{1}{s + ikv + Dk^2}$$

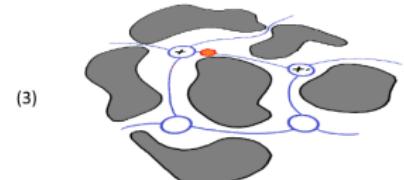
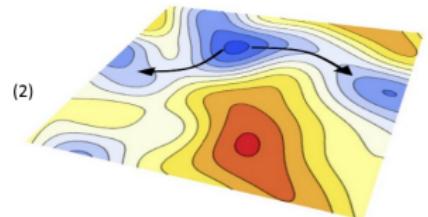
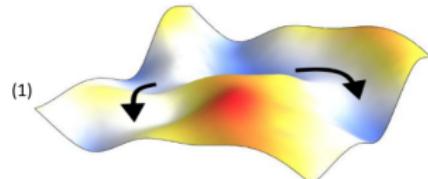


Heterogeneous Continuous Time Random Walk model on a graph:

- graph with transition **matrix Q** ,
- heterogeneous **travel time distributions** $\psi_{xx'}(t)$ between nodes x, x' .

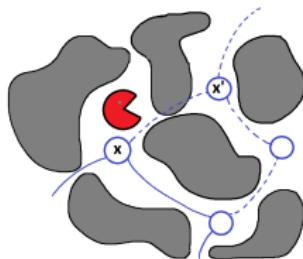
The generalized transition matrix

$$Q_{xx'}(t) = Q_{xx'} \psi_{xx'}(t)$$



HCTRW is a formalism for studying diffusion in heterogeneous structures (1),(2).

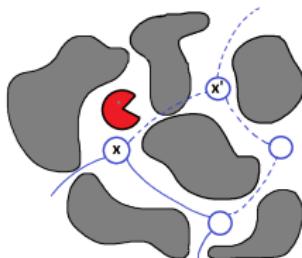
Analytical results for HCTRW model



HCTRW on a
graph:

1. graph with
transition **matrix Q** ,
2. travel times
 $\psi_{xx'}(t)$ between
nodes x, x'

Analytical results for HCTRW model

Analytic formula for HCTRW propagator $\tilde{P}_{x_0 x}(s)$:

$$\tilde{P}_{x_0 x}(s) = \frac{1 - \sum_{x'} \tilde{Q}(s)_{xx'}}{s} [(I - \tilde{Q}(s))^{-1}]_{x_0 x}, \quad (1)$$

where $\tilde{Q}_{xx'}(s) = Q_{xx'} \tilde{\psi}_{xx'}(s)$.Long-time behaviour for HCTRW propagator $P_{x_0 x}(t)$:**HCTRW** on a graph:

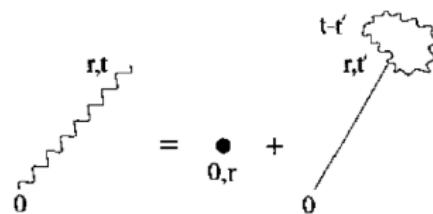
- graph with transition matrix Q ,
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$\psi_{xx'}(t)$ between nodes x, x'

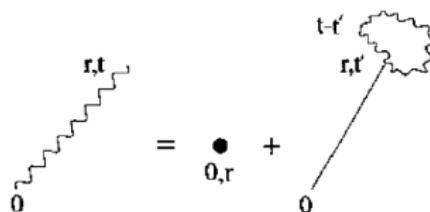
$$\tilde{P}_{x_0 x}(s) \simeq \underbrace{p_{st}(x)}_{\text{stat. dist.}} + t(x) \sum_{k>0} \frac{u_{0k}^\dagger(x) v_{0k}(x_0)}{\lambda_{0k} + s\lambda_{1k}} \quad (2)$$

λ_{0k} are eigenvalues, u_{0k}, v_{0k} eigenvectors of $I - Q$.
 $\lambda_{0k} + s\lambda_{1k}$ is the 1st order correction for $I - Q + sT$,
 $T_{xx'} = Q_{xx'} \langle T_{xx'} \rangle$, $t(x) = \sum_{x'} T_{xx'}$.

First passage time



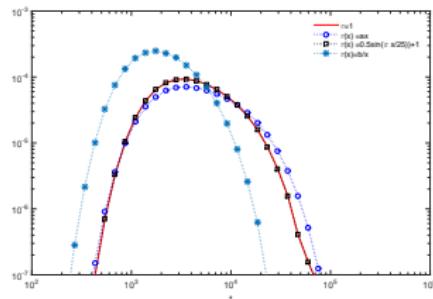
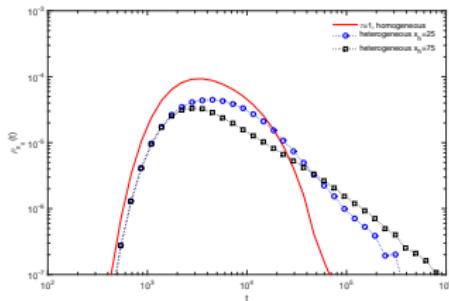
First passage time



$F(x, t)$ - first passage time probability, then using renewal approach, [Redner, 2002]:

$$P(x, t) = \delta_{x,0}\delta_{t_0} + \sum_{t' \neq t} F(r, t') P(0, t - t')$$

Results for Heterogeneous random walks



The **FPT density** of HCTRW on an interval, absorbing $x = 100$: (top) hetero-nodes $x_h = 25, 75$ with heavy-t.distr.
 $\alpha = 0.5$; (bottom) $\tau_{\pm} = 1$; $\tau_{xx \pm 1} = ax$;
 $\tau_{xx+1} = 0.5 \sin(\pi x/25) + 1$; $\tau_{xx+1} = b/x$.

Analytic formula for HCTRW

propagator $P_{x_0 x}(t)$ on a graph is derived.

Analytic framework of HCTRW

links structural graph properties and dynamical RW properties

HCTRW framework allows of study asymptotic solutions, FPT for processes on graphs.

- Grebenkov, Tupikina, "Heterogeneous continuous time random walks", PRE 012148 (2018)

Random walks: discrete space and time

Will we get Gaussian for any hopping process, where $\langle x^2 \rangle$ is finite?



Random walks: discrete space and time

Will we get Gaussian for any hopping process, where $\langle x^2 \rangle$ is finite?

Yes! Probability distribution is independent from details of single-step hopping [Gnedenko, Kolmogorov 1954]



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