The 1st Scenario for Getting at Multi-Scale Modeling

Analytical descriptions for explaining mathematically the facts discovered by experiments (a macro-scale viewpoint)

CTRW

Waiting time function

 $w(t) \sim \left(\frac{t}{\tau_0}\right)^{-(\alpha+1)} \begin{array}{c} \text{Discovered by} \\ \text{experiments} \end{array}$

1. Description of CTRW with PDF

$$P(x,t) = \int_{0}^{t} \eta(x,t') \Phi(t-t') dt' \qquad \frac{PDF for particle}{\underbrace{existing}_{and space \ x}} in time \ t$$

1) PDF for particle $\underline{gathering}$ in time t and space x

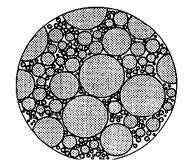
$$\eta(x,t) = \int_{-\infty}^{\infty} dx' \int_{0}^{t} dt' \eta(x',t') \phi(x-x',t-t') + \delta(t)a(x)$$

2) PDF for particle <u>staying</u> by the duration time t

$$\Phi(t) = 1 - \int_0^t w(t')dt' \qquad \qquad \phi(x,t) = \lambda(x)w(t)$$

Microstructure of porous media

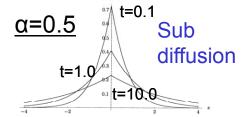
Mathematical reasoning behind the observed phenomena in macro-scale

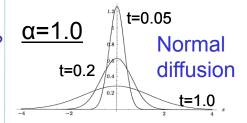


- •What are geometric invariants?
- •How do we combine the geometric invariants with FDE in a mathematical framework?

Fractional differential equation (FDE)







2. Determination of w(t)

Necessary condition

$$w(t) \sim \left(\frac{t}{\tau_0}\right)^{-(\alpha+1)}, \quad t \to \infty$$



3. Analytic procedures

- 1) Fourier transformation to \underline{x} and Laplace transformation to \underline{t}
- 2) Approximation taken for only the first term of the infinite series
- 3) Inverse Fourier transformation to x and Inverse Laplace transformation to t

$$\partial_t^{\alpha} P(x,t) = c \nabla^{\beta} P(x,t) - \gamma_0 \partial_x P(x,t)$$

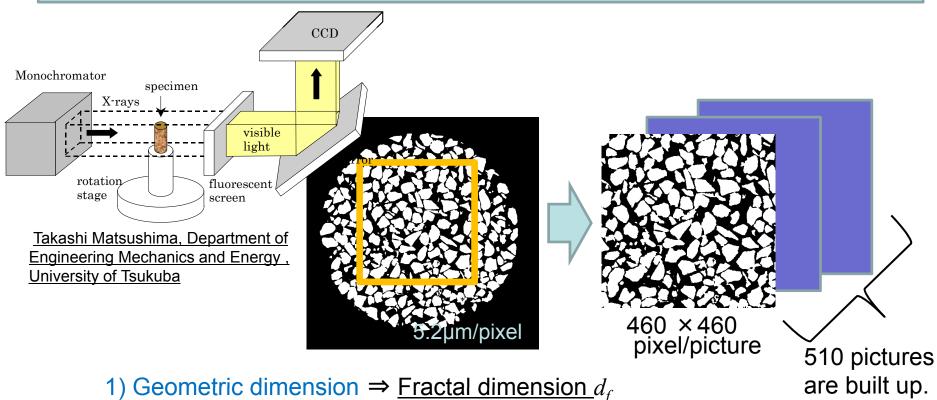
Fractional differential in terms of time

$$\partial_t^{\alpha} f(t) = \frac{1}{\Gamma(1-\alpha)} \int_0^t (t-\tau)^{-\alpha} f'(\tau) d\tau$$

The degree of history to be retained from the initial time (t = 0) to the current time (time t). The smaller α is, the more history will be retained.

The 2nd Scenario for Getting at Multi-Scale Modeling

Characterization of the geometric features of the specimens of 3D CT-image (a micro-scale viewpoint)



- 2) Analytic dimension \Rightarrow Spectral dimension d_s

How do we combine these geometric invariants with FDE(Fractional Differential Equation) in a mathematical framework?



Relationship between df, ds and the Fractional Differential Order α Conjectured by Physicists

Manabu Machida, Department of Mathematics, University of Michigan

Bouchaud, J. P., and A. Georges, 1990, Physics. Reports 195, 127

The behavior of the mean volume V occupies by a diffusion particles initially concentrated on a given site x is given by the mean squared displacement and the fractal dimension.

$$V(x,r) \sim r^{d_f} \sim \left(\sqrt{\langle x(t)^2 \rangle}\right)^{d_f} \quad \text{(1)} \quad \begin{array}{l} V(x,r) \coloneqq \mu(B(x,r)) & \text{V(x, r) is the} \\ B(x,r) \coloneqq \left\{ y \in V \middle| d(x,y) < r \right\} & \text{geodesic ball B(x,r)}. \end{array}$$

M. Barlow and E. Perkins, Brownian motion on the Sierpinski gasket, Probab. Th. Rel. Fields, 79 (1988), showed that the following heat kernel takes place for a large variety of fractal sets;

$$p(x,y,t) \sim t^{-\frac{d_s}{2}} \exp\left(-\left(\frac{d(x,y)^{d_w}}{ct}\right)^{\frac{1}{d_w-1}}\right) \qquad \text{(2)} \qquad \text{in the case of } y=x, \\ p(x,x,t) \sim t^{-\frac{d_s}{2}} \qquad \text{(3)}$$

We assume that

$$p(x,x,t) \sim V(x,r)^{-1} \tag{4}$$

With Eq. (1), Eq. (4) and Eq. (3), we obtain that

 $\frac{1 \text{ Eq. (1), Eq. (2)}}{\left\langle x(t)^2 \right\rangle \sim V(x,r)^{\frac{2}{d_f}}} \sim p(x,x,t)^{-\frac{2}{d_f}} \sim \left(t^{-\frac{d_s}{2}}\right)^{-\frac{1}{d_f}} \sim t^{\frac{d_s}{d_f}}$ (5) $\alpha = \frac{d_s}{d_f}$ (7) Numerical experiments using CTRW say that

$$\langle x(t)^2 \rangle \sim t^{\alpha}$$
 (6)

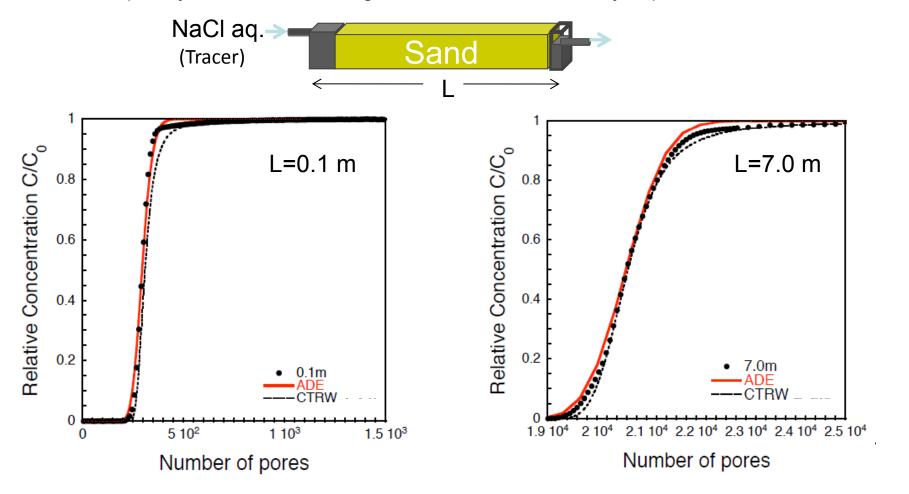
By comparing Eq. (5) and Eq. (6), we have

$$\alpha = \frac{d_s}{d_f} \qquad (7)$$

Comparison of Diffusion Behavior between Small Scale Experiment and Large Scale One

Yuko Hatano, Department of Risk Engineering, University of Tsukuba

- √ The result of the small scale experiment shows that the diffusion follows ADE (normal diffusion).
- ✓ The result of the large scale one not only differs from ADE, but CTRW also can't explain the behavior completely. We need a scaling law to combine laboratory experiments with field scale.



The 3rd Scenario for Getting at Multi-Scale Modeling

Deductive reasoning to derive the fractional differential equation using the homogenization method (a multi-scale viewpoint)

Masaaki Uesaka, Graduate School of Mathematical Science, The University of Tokyo

J.L. Auriault and J. Lewandowska, Non-Gaussian Diffusion Modeling in Composite Porus Media by Homogenization: Tail Effect, Transport in Porous Media, 21:47-70,1995

$$\begin{cases} \partial_t c^{\varepsilon} - \nabla \cdot \left(D(x/\varepsilon) \nabla c^{\varepsilon} \right) = 0 & in \ \Omega \times (0,T) \\ c^{\varepsilon} (x,0) = c_0(x) & in \ \Omega \\ \partial_v c^{\varepsilon} = 0 & on \ \partial \Omega \times (0,T) \end{cases}$$
 (1)

$$c^{\varepsilon}(x,0)=c_0(x)$$

$$\partial c^{\varepsilon} = 0$$

$$in \Omega$$

on
$$\partial \Omega \times (0,T)$$
 (3)

(2)

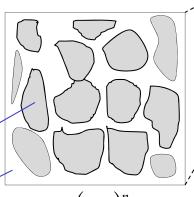
 Ω is composed of periodic components of a microcell.

where

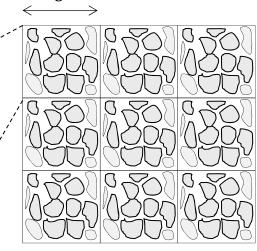
$$D(y) = D_f(y) 1_F(y) + D_m(y) 1_M(y)$$
 (4)

$$D_m/D_f \cong \varepsilon^2 \tag{5}$$

The ratio of the two effective diffusion coefficients is defined in terms of the power of the homogenization parameter.



$$Y = (0,1)^n$$



$$\Omega = \Omega_f^{\varepsilon} \cup \Omega_m^{\varepsilon}$$

Masaaki Uesaka, Graduate School of Mathematical Science, The University of Tokyo

J.L. Auriault and J. Lewandowska, Non-Gaussian Diffusion Modeling in Composite Porus Media by Homogenization: Tail Effect, Transport in Porous Media, 21:47-70,1995

When ε goes to zero, c^{ε} converges to the solution of the following homogenized equations.

$$\begin{cases}
\partial_{t} c - \int_{0}^{t} K(t - \tau) \partial_{\tau}^{2} c d\tau = \nabla \cdot \left(D^{eff} \nabla c\right) & in \Omega \times (0, T) \\
c(x, 0) = c_{0}(x) & in \Omega \\
\partial_{v} c = 0 & on \partial \Omega \times (0, T)
\end{cases} \tag{6}$$

Where D^{eff} and K are determined by D_m , D_f , and the shape of M;

The memory function K(t) is an inverse Laplace transform of the function (9), where k(y,p) is a solution of Eq. (10) and (11)

Here
$$k(y,p)$$
 is a solution of Eq. (10) and (11)
$$(\mathcal{L} K)(p) = \frac{1}{p} \int_{M} k(y,p) dy$$

$$\begin{cases} \nabla_{y} \cdot (D_{m} \nabla_{y} k(\cdot,p)) = p(k(\cdot,p)-1) & in M \\ k(\cdot,p)=0 & in \partial M. \end{cases}$$
 (10)

What shape of *M* leads to a similar effect like the fractional differential?

Summary and Discussion

- <u>The 1st scenario</u>: Analytical descriptions for explaining mathematically the facts discovered by experiments (a macro-scale viewpoint)
 - \rightarrow The α fractional differential in terms of time
 - \rightarrow Levy flights \rightarrow The β fractional differential in terms of space
- <u>The 2nd scenario</u>: Characterization of the geometric features of the specimens of 3D-CT image (a micro-scale viewpoint)
 - The equality $\alpha = d_s/d_f$ based on the hypothesis: $p(x,x,t) \sim V(x,r)^{-1}$ exists.
 - $ightharpoonup \operatorname{If} d_s$ and d_f can be determined by the geometric features, we will be able to combine the fractional differential equation with the geometric features in the observed region.
 - Experimental results say that we need a scaling law to connect laboratory experiments with field scale's observations.
- <u>The 3rd scenario</u>: Deductive reasoning to derive the fractional differential equation using the homogenization method (a multi-scale viewpoint)
 - Geometric conditions M in a microcell
 - → The memory function generates the long-tail effect
 - \rightarrow Our concern is 'what shape of M leads to a similar effect like the fractional differential'?

Relationship among CTRW, FDE and d_s/d_f in terms of α , between Levy flight, FDE in terms of β

CTRW

$$\langle x(t)^2 \rangle \sim t^{\alpha}$$

Geometric index

$$\alpha = \frac{d_s}{d_f}$$

Levy symmetric alpha stable distribution

$$(\mathcal{F}\lambda)(\mathbf{k}) = \exp(-c|\mathbf{k}|^{\beta} + i\gamma_0)$$

FDE (Fractional Differential Equation)

The α fractional differential in terms of time

The β fractional differential in terms of space

$$\partial_t^{\alpha} P(x,t) \neq c \nabla^{\beta} P(x,t) - \gamma_0 \partial_x P(x,t)$$

The sub-Gaussian heat kernel on graphs

$$p(x,y,t) \sim t^{-\frac{d_s}{2}} \exp\left(-\left(\frac{d(x,y)^{d_w}}{ct}\right)^{\frac{1}{d_w-1}}\right)$$

Let E(x, R) be the mean exit time from the ball B(x,R). Consider the hypothesis

$$E(x,R) \sim R^{d_w}$$
 walk dimension $d_w \ge 2$
 $d_w = 2 \rightarrow \text{normal}$
 $d_w > 2 \rightarrow \text{sub-diffusion}$

Thank you for your attention!

異分野連携の場「数学イノベーションプラットフォーム(仮称)」へ、ようこそ!

数学・数理科学により、諸科学・工学・産業界で得られている一連の実験的事実または経験的事実を一貫性のある論理で統合できる理論を構築し、同理論に基づく数理モデルを介し、諸科学・工学・産業における未解決問題のブレークスルーと当該分野の研究・開発工期の驚異的短縮ができたという実績を示してゆくことを目指します。

数学イノベーションを目指す有志の皆様の 参加を歓迎します。

