## Mixing and Reaction in Highly Heterogeneous Porous Media

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## Thanks to my collaborators





## Heterogeneity Spreading vs Mixing



Heterogeneity => Typically Superdiffusive spreading

## Topics I'll Talk About

- Incomplete Mixing and Slowdown on Chemical Reactions - Fickian and Non-Fickian
- Incomplete Mixing - When Might Tails not be due to fractional type behavior?
- How fractional dispersion can make reactions happen in places where Fickian models say it cannot - And I don't mean tails.


## Topic 1

## -

Incomplete Mixing and Slowdown on Chemical
Reactions - Fickian and NonFickian Transport

## What does any of this mean for reactive transport?

## Consider the following example:



Instantaneous? Reversible? Equilibrium?

## Let's start easy - forget heterogeneity

- Kinetic, irreversible

$$
\begin{aligned}
& \mathrm{d}[\mathrm{~A}] / \mathrm{dt}=\mathrm{k}[\mathrm{~A}][\mathrm{B}] \\
& \mathrm{d}[\mathrm{~B}] / \mathrm{dt}=\mathrm{k}[\mathrm{~A}][\mathrm{B}] \\
& \mathrm{d}[\mathrm{C}] / \mathrm{dt}=-\mathrm{k}[\mathrm{~A}][\mathrm{B}]
\end{aligned}
$$

- Analytical Solution if
[A]=[B] (assume initially equal $->$ always equal)
- $A=A_{0} /\left(1+k A_{0} t\right)$




## To study this let's use a numerical

## model....



- Move Particles with a random walk
- Based on the distance between two particles calculate probability that they will collocate
- Then based on the reaction multiply probability that reaction will occur


# Step 1 - Move Particles by Random Motion 

Update Particle Positions by $\mathbf{x}(\mathrm{t}+\mathrm{dt})=\mathbf{x}(\mathrm{t})+\boldsymbol{\xi}$
Random Jump Reflecting Dispersion

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## Step 2 - Search for Neighbors of Opposite Particle

Particle 1

Gives distances
s1
s2

## Step 3 - Calculate Probability of RXN

Probability of Reaction

Particle 1-1

$=\quad$ function of distance<br>Convolution of position<br>Probability of Collocation densities<br>Fickian=> Gaussian<br>Fractional=>Stable<br>e.g.<br>X<br>$$
v(\mathrm{~s}, \Delta t)=\frac{1}{(8 \pi D \Delta t)^{d / 2}} e^{-\frac{s^{2}}{8 D \Delta t}}
$$



Probability of Reaction Given
function of reaction Kinetics
Collocation

## Step 4 - Die or Survive

Particle 1-1
Generate a random number $0<P<1$

If $\mathrm{P}>$ Probability of Reaction
Kill both particles

If less move to next blue particle

## Step 4 - Die or Survive

Particle 1-2

Kill both particles
If less move to next blue particle

## Step 4 - Die or Survive

Particle 1-2

Kill both particles
If less move to next blue particle

And so on Cycling through all blues

## Step 4 - Die or Survive

Particle 1-2

Kill both particles
If less move to next blue particle

And so on Cycling through all blues

## Repeat for Each red Particle

Particle 2

And so on Cycling through all reds
Then back to Step One (Move Particles)

Non Dimensional Time $\left(C_{0} K_{f} t\right)$


## What do we observe? For 1d Brownian Motion



Analytical Solution

## Other Observations of the Same (different methods of study)



Benson \& Meerschaert 2008, WRR

Countless other examples:

Astrophysics Particle Physics
Biochemical Processes
Turbulent Environmental Flows
Population Dynamics
Warfare Simulation

## What's going on... Let's take a look at concentrations in 1d



## What's going on... Let's take a look at concentrations in 1d





## But what does this have to with fractional transport?

- Consider the following problem

$$
\begin{gathered}
\frac{\partial C_{i}}{\partial t}=D \frac{\partial^{\alpha} C_{i}}{\partial|x|^{\alpha}}-k C_{A} C_{B}, \quad i=A, B \quad-\infty<x<\infty \\
C_{i}(x, t)=\overline{C_{i}}(t)+C_{i}^{\prime}(x, t)
\end{gathered}
$$

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$\frac{\partial C_{i}}{\partial t}=D \frac{\partial^{\alpha} C_{i}}{\partial|x|^{\alpha}}-k C_{A} C_{B}, \quad i=A, B \quad-\infty<x<\infty$
Average $C_{i}(x, t)=\overline{C_{i}}(t)+C_{i}^{\prime}(x, t)$ Remainder

$$
\frac{\partial \overline{C_{i}}}{\partial t}=-k \overline{C_{A}} \overline{C_{B}}-k \overline{C_{A}^{\prime} C_{B}^{\prime}} \quad \frac{\partial C_{i}^{\prime}}{\partial t}=D \frac{\partial^{\alpha} C_{i}^{\prime}}{\partial|x|^{\alpha}}-k \overline{C_{A}} C_{B}^{\prime}-k C_{A}^{\prime} \overline{C_{B}}-k C_{A}^{\prime} C_{B}^{\prime}+k \overline{C_{A}^{\prime} C_{B}^{\prime}}
$$

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& \frac{\partial C_{i}}{\partial t}=D \frac{\partial^{\alpha} C_{i}}{\partial|x|^{\alpha}}-k C_{A} C_{B}, \quad i=A, B \quad-\infty<x<\infty \\
& \text { Average } \quad C_{i}(x, t)=\overline{C_{i}}(t)+C_{i}^{\prime}(x, t) \quad \text { Remainder } \\
& =-k \overline{C_{A}} \frac{C_{B}}{C_{B}}-k \overline{C_{A}^{\prime} C_{B}^{\prime}}=D \frac{\partial^{\alpha} C_{i}^{\prime}}{\partial|x|^{\alpha}}-k \overline{C_{A}} C_{B}^{\prime}-k C_{A}^{\prime} \overline{C_{B}}-k C_{A}^{\prime} C_{B}^{\prime}+k \overline{C_{A}^{\prime} C_{B}^{\prime}} \\
& f(x, y, t)=\overline{C_{A}^{\prime}(x, t) C_{B}^{\prime}(y, t)}
\end{aligned}
$$

## But what does this have to with fractional transport?

- Consider the following problem

$$
\frac{\partial C_{i}}{\partial t}=D \frac{\partial^{\alpha} C_{i}}{\partial|x|^{\alpha}}-k C_{A} C_{B}, \quad i=A, B \quad-\infty<x<\infty
$$



$$
C_{i}(x, t)=\overline{C_{i}}(t)+C_{i}^{\prime}(x, t) \quad \text { Remainder }
$$

Perturbation closure

$$
\frac{\partial \overline{C_{i}}}{\partial t}=-k \overline{C_{A}} \overline{C_{B}}-k \overline{C_{A}^{\prime} C_{B}^{\prime}} \quad \frac{\partial C_{i}^{\prime}}{\partial t}=D
$$

$$
\frac{\partial C_{i}^{\prime}}{\partial t}=D \frac{\partial^{\alpha} C_{i}^{\prime}}{\partial|x|^{\alpha}}-k \overline{C_{A}} C_{B}^{\prime}-k C_{A}^{\prime} \overline{C_{B}}-k C_{A}^{\prime} C_{B}^{\prime}+k \overline{C_{A}^{\prime} C_{B}^{\prime}}
$$

$$
f(x, y, t)=\overline{C_{A}^{\prime}(x, t) C_{B}^{\prime}(y, t)}
$$

$$
f(x, y, t)=\int_{-\infty}^{\infty} R(\xi, y) G(x, \xi, t) d \xi
$$

$$
\begin{gathered}
\overline{C_{A}^{\prime}(x, 0) C_{B}^{\prime}(y, 0)}=R(x, y) \\
G(x, \xi, t)=\frac{1}{2 \pi} \int_{-\infty}^{\infty} e^{-2 D_{\dagger}|k|^{\alpha} t} e^{i k(x-\xi)} d k
\end{gathered}
$$

## So my equation becomes

$$
\frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \int_{-\infty}^{\infty} R(\xi, x) G(x, \xi, t) d \xi .
$$

## So

$$
\begin{aligned}
\frac{\partial \overline{C_{i}}}{\partial t} & =-k{\overline{C_{i}}}^{2}+k \int_{-\infty}^{\infty} R(\xi, x) G(x, \xi, t) d \xi \\
R(x, y) & =\sigma^{2} l \delta(x-y) \\
& \square \frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \chi t^{-\frac{1}{\alpha}}
\end{aligned}
$$

## So

$$
\frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \int_{-\infty}^{\infty} R(\xi, x) G(x, \xi, t) d \xi
$$

$$
R(x, y)=\sigma^{2} l \delta(x-y) \longmapsto \frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \chi t^{-\frac{1}{\alpha}}
$$

$$
\overline{C_{i}}(t)=\frac{\sqrt{\chi^{*}}}{t^{\frac{1}{2 \alpha}}} \frac{\left(I_{-\frac{\alpha-1}{2 \alpha-1}}(z)-\kappa K_{\frac{\alpha-1}{2 \alpha-1}}(z)\right)}{\left(I_{\frac{\alpha}{2 \alpha-1}}(z)+\kappa K_{\frac{\alpha}{2 \alpha-1}}(z)\right)}
$$

$$
z=\frac{2 \alpha \sqrt{\chi^{*}}}{2 \alpha-1} t^{\frac{2 \alpha-1}{2 \alpha}}
$$

## So

$$
\begin{aligned}
& \frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \int_{-\infty}^{\infty} R(\xi, x) G(x, \xi, t) d \xi \\
& R(x, y)=\sigma^{2} l \delta(x-y) \Longrightarrow \frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \chi t^{-\frac{1}{\alpha}} \\
& \overline{C_{i}}(t)=\frac{\sqrt{\chi^{*}}}{t^{\frac{1}{2 \alpha}}} \frac{\left.\left(I_{-\frac{\alpha-1}{2 \alpha-1}}^{2 \alpha}(z)-\kappa K_{\frac{\alpha-1}{2 \alpha-1}}^{2 \alpha-1}\right)\right)}{\left(I_{\frac{\alpha}{2 \alpha-1}}(z)+\kappa K_{\frac{\alpha}{2 \alpha-1}}^{2(z)}\right)} \\
& z=\frac{2 \alpha \sqrt{\chi^{*}}}{2 \alpha-1} t^{\frac{2 \alpha-1}{2 \alpha}}
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## So

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& \frac{\partial \overline{C_{i}}}{\partial t}=-k{\overline{C_{i}}}^{2}+k \int_{-\infty}^{\infty} R(\xi, x) G(x, \xi, t) d \xi . \\
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& z=\frac{2 \alpha \sqrt{\chi_{A}^{*}}}{2 \alpha-1} t^{\frac{2 \alpha-1}{2 \alpha}}(t)=\frac{1}{1+\left(t-t_{0}\right)} \\
&
\end{aligned}
$$

## What does Solution Look like



## Validation



## Topic 2

- 

Incomplete Mixing - When
Might Tails Exist, but not be due to fractional type behavior?
Or are they and I'm just wrong

## Let's Look at Some Experiments



Famous Experiment by Gramling et al

## Gramling's Measurements vs Predictions





## Gramling's Measurements vs Predictions




## Anomalous Transport

## We Don't Think So



There is no evidence of anomalous transport in non reactive flow experiments through the same column

## Looking closely at Gramling data



Looks a lot like what we called islands from our numerical models.....

Could it be incomplete mixing only?

## Our Model

Set up an initial condition with all $A$ on one side and $B$ on the other

## Our Model



Move Every Particle - Jump by dispersion

## Our Model

I!


Now kill some particles probabilistically for reaction following rules from before

## When we use our Methods



Pretty Good Agreement - And we can Explain Why

## And the Tails....



We still use conventional transport models - but incorporate incomplete mixing effects!
To quote Dave Benson- in our model particles can 'advance further into enemy territory before reacting'

## Our Model



SO - Tail does not need to mean anomalous/fractional Or
Can it be interpreted that way?

## Topic 3

If not in the tails, how fractional dispersion can make reactions happen in places where Fickian dispersion cannot

# Let's consider another common, but very distinct chemical reaction 

## Instantaneous Equilibrium Reactions



$$
\begin{gathered}
\frac{\partial C_{i}}{\partial t}=D \nabla^{2} C_{i}+r \quad i=1,2 \\
\frac{\partial C_{3}}{\partial t}=-r \\
C_{1} C_{2}=K
\end{gathered}
$$

Equilibrium

## Instantaneous Equilibrium Reaction

$$
\text { (a) } \frac{\partial C_{i}}{\partial t}=D \nabla^{2} C_{i}+r \quad i=1,2
$$



$$
\begin{gathered}
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\frac{\partial C_{3}}{\partial t}=-r \\
C_{1} C_{2}=K
\end{gathered}
$$

- Local Measure of Mixing - Drives many Reactions

$$
r=\underbrace{(D \nabla U . \nabla U)}_{\text {MIXING }} \underbrace{\left(\frac{d^{2} C_{1}}{d U^{2}}\right)}_{\text {Chemistry }}
$$

- Local Measure of Mixing - Drives many Reactions

$$
r=(D \nabla U . \nabla U)
$$

MIXING

???=\$\$\$


Chemistry

Interesting, but let's worry about this later on as heterogeneity plays little role on this

## How do we Quanitify Mixing?

- Local Measure of Mixing - Drives many Reactions

$$
r=(D \nabla U . \nabla U)
$$



MIXING

- Global Measure of Mixing (integrate r over whole domain)

$$
M=\int_{\Omega}(D \nabla U . \nabla U) d \Omega=-\frac{1}{2} \frac{d}{d t} \int_{\Omega} U^{2} d \Omega
$$

Scalar Dissipation Rate

## Homogeneous Mixing



A little bit boring, no?

## Replace Fickian with Fractional Dispersion

$$
\begin{gathered}
\frac{\partial C}{\partial t}=D \frac{\partial^{\alpha} C}{\partial|x|^{\alpha}}+r \\
1<\alpha \leq 2
\end{gathered}
$$

## Take a step back



Recall we can think of this reaction in Terms of a conservative component $u$. Consider a system with u=constant initially and then we inject a pulse of different u at position $\mathrm{x}=0$.

$$
x=0
$$



## When $\alpha=2$

## Classical Fickian diffusion

Concentration at a fixed time



$$
r=(D \nabla U . \nabla U)\left(\frac{d^{2} C_{1}}{d U^{2}}\right)
$$

## What happens as $\alpha$ changes?



## Let's take a closer look at this



## What does this mean for precipitation

 reactions?$$
r(x, t)=D_{\alpha} \sum_{k=1}^{\infty}\binom{\alpha-1}{k} \frac{\partial^{\alpha-k} u}{\partial|x|^{\alpha-k}} \frac{\partial^{k}}{\partial x^{k}} \frac{d c}{d u}
$$



# What does this mean for precipitation reactions? 



## Global Measures - Mixing (Scalar Dissipation)



At early times more anomalous (smaller $\alpha$ )=more mixing At late times - less mixing

## Global Measures - Total Reaction Rate



More anomalous - always less reactions (a mixing scale effect???)

## Questions



