

Stochastic Fields - Toward High k_{\perp} ; Random Conductivity

→ so far:

- reviewed theory of Hamiltonian chaos
- derived QL D_M
- derived $\chi_{\perp e}$ due stochastic fields in $k_{\perp} < 1$ regime = diffusion
- focused on interaction of scattering (flucts), collisions, coarse graining
- discussed transport in FC plasma, as ~~example~~ example of $\tau_{ac} \rightarrow \infty$ regime.

→ Observations

- idea of resonance (small denominator problem) and resonance overlap fundamental to Hamiltonian chaos.
- $k_{\perp} \sim \tau_{ac} / \tau_{coll}$
- might ask: unified treatment that combines $k_{\perp} < 1$, $k_{\perp} > 1$ regimes \Rightarrow renormalized response.
- in hydro treatment of $\chi_{\perp e}$, what of nominal 3rd order contribution vs $\chi_{\perp, coll}$. See (K+P) ↓
- in diffusive treatment of high k_{\perp} regime (as in Taylor + Mc Namara) valid? See ~~Reichenko~~ Reichenko, Gurbatov papers.
- $\tau_{ac} \rightarrow \infty$ can recover strong QL at modest k_{\perp} level

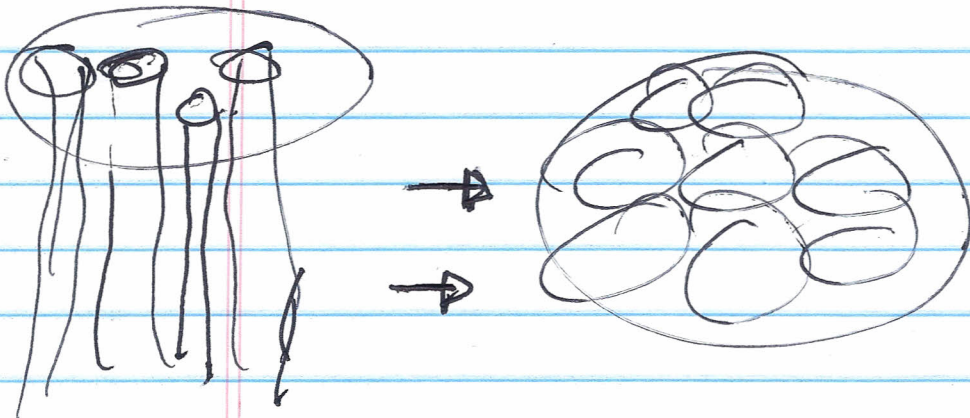
Here:

- general analysis of diffusion
- aspects of percolation, large Ku regime
- Dykhne method \rightarrow conduction in random media.

\rightarrow Recall, $Ku \sim \rho_0 \sigma B / R \Delta_L$

- have considered low Ku , with
 - finite ρ_0
 - \Rightarrow inhomogeneity in Z

- now, consider $Ku \rightarrow \infty$ limit, opposite
 - \Rightarrow random field, in x, y .
 - \Rightarrow homogeneous in Z



$$\text{d.e.} \left\{ \begin{array}{l} \frac{dx}{dz} = b_r = \frac{\partial A}{\partial y} \\ \frac{dy}{dz} = b_\theta = -\frac{\partial A}{\partial x} \end{array} \right.$$

$$\text{From: } \frac{dr}{dz} = b_r$$

$$r \frac{d\theta}{dz} = \frac{\langle B_\theta \rangle}{B_0} + b_y$$

equivalent, of $(L_s \rightarrow \infty)$ course, to G.C. plasma:

$$\left\{ \begin{array}{l} \frac{dx}{dt} = -\frac{c}{B} \partial_y \phi \\ \frac{dy}{dt} = \frac{c}{B} \partial_x \phi \end{array} \right.$$

\Rightarrow motivates study of random media transport!

Formally can extend D_M calculation to include resonance broadening

c.e $\frac{\partial \tilde{F}}{\partial z} + b \cdot \nabla \tilde{F} = -b \frac{\partial \langle F \rangle}{\partial r}$

\Rightarrow

$$D_M = \sum_n |\tilde{b}_{r_n}|^2 \frac{\epsilon}{k_{rn} + i k_{\perp}^2 D_M}$$

where

$$k_{\perp}^2 D_M / k_{rn} \sim k_{rn}^2$$

For $k_{rn} \ll 1 \Rightarrow$

$$D_M \approx \sum_n |\tilde{b}_{r_n}|^2 \delta(k_{rn}), \text{ aka RSTZ}$$

For $k_{rn} \gg 1$

$$D_M \approx \sum_n |\tilde{b}_{r_n}|^2 / k_{\perp}^2 D_M$$

aka Taylor, McNamara.

$$\Rightarrow D_M \approx \left(\sum_H |\tilde{A}_H|^2 \right)^{1/2}$$

$$\sim \tilde{b} \Delta$$

So $\kappa_H < 1 \Rightarrow D_M \sim \tilde{b}^2 \ell_{av}$

$\kappa_H > 1 \Rightarrow D_M \sim \tilde{b} \Delta$

and transport $\sim \langle A^2 \rangle^{1/2}$

But, is $\kappa_H > 1$ regime really diffusive?

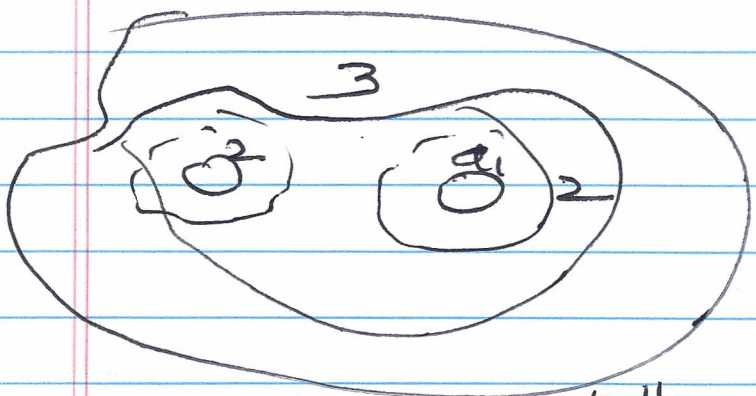
\rightarrow recall: $\frac{dx}{dz} = \frac{\nabla A}{\perp} \times \vec{z}$

{akin 2D random media, for
A indep z.

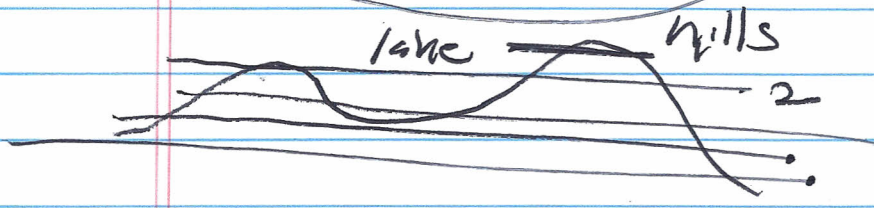
\rightarrow can view physically as:

topographical map

c.e



Map



levels

now, as $\frac{dx}{dyA} = \frac{-dy}{-dxA} = \frac{dz}{1}$

$\Rightarrow \frac{dy}{dx} = -\frac{\partial xA}{\partial yA}$

$\Rightarrow \boxed{\nabla A \cdot dx = 0}$

\Rightarrow - Linear traverse const A_0 contours, as on map

- $\langle A \rangle = 0$, $\langle A^2 \rangle = A_0^2$

- avg. depth, height of "lakes", "hills" set by A_0

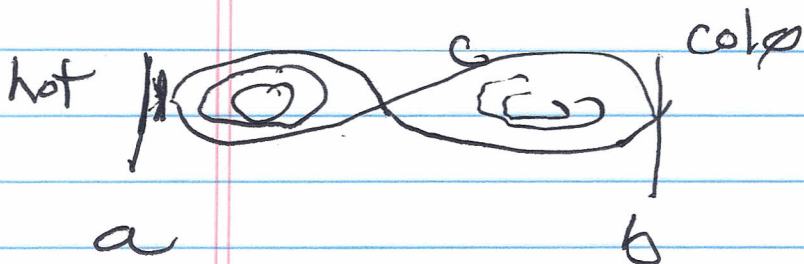
- most contours closed, isolated
 \Rightarrow little contribution to transport

- but contours along "passes"

, i.e. 3, can take on long path lengths.

\Rightarrow transport occurs primarily along there

\Rightarrow percolation, not diffusion



$a \rightarrow b$ transport isolated along contour c .

\sim more like 'lightning bolt' than diffusion

\sim signature would be sharply localized strike mark (if $b \rightarrow P(F)$) and not periodic

~ percolation \rightarrow extension of \mathbb{P}
 mean length as $A \rightarrow 0$

$$l_A \sim A^{-\gamma}$$

$$\gamma \sim 2.4.$$

Message:

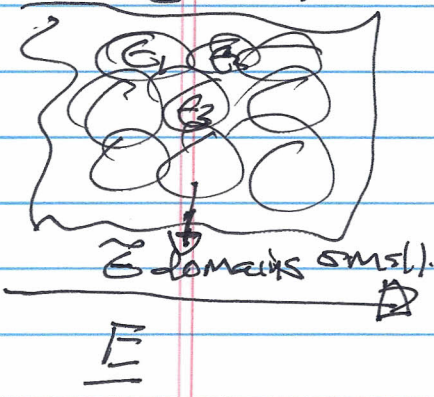
- to understand $ku > 1$ regime,
 useful to examine

\rightarrow transport in random media

\rightarrow percolation theory

- Random Media \Rightarrow {characterize media with
 {static ensemble scatterers.
 \rightarrow 2D

(i) Effective Permittivity = Mean Field Theory (M/CM)



- what is effective \underline{D} ?

= given local fluctuations

$$\underline{E} = \underline{\bar{E}} + \underline{\tilde{E}}$$

what is effective \underline{G} ?

Now!

$$\underline{\bar{D}} = \underline{\bar{E}} \underline{\bar{G}}$$

\bar{D} mean, \bar{E} mean, \bar{G} effective G

Mean field question

$$\underline{\bar{E}} = \frac{1}{V} \int d^3x \underline{E}$$

Now.

$$\nabla \cdot (\underline{G} \underline{E}) = 0$$

Proceed by QL/MFT

$$\Rightarrow \nabla \cdot (\underline{\bar{E}} + \underline{\tilde{E}}) (\underline{\bar{E}} + \underline{\tilde{E}}) = 0$$

where: $\underline{\bar{D}} = \overline{(\underline{\bar{E}} + \underline{\tilde{E}}) (\underline{\bar{E}} + \underline{\tilde{E}})} = \underline{\bar{E}} \underline{\bar{E}} + \overline{\underline{\tilde{E}} \underline{\tilde{E}}}$

\uparrow
QL

so, need $\tilde{\underline{E}}$:

$$\nabla \cdot (\underline{\underline{E}} \underline{\underline{E}}) + \tilde{\underline{E}} \nabla \cdot \underline{\underline{E}} + \underline{\underline{E}} \cdot \nabla \tilde{\underline{E}} + \cancel{\lambda \mathbf{1}} \implies \nabla \cdot \underline{\underline{D}} = 0$$

to first order:

$$\underline{\underline{E}} \nabla \cdot \underline{\underline{E}} = -\underline{\underline{E}} \cdot \nabla \tilde{\underline{E}}$$

relates
 $\underline{\underline{E}}$ to $\tilde{\underline{E}}$

domains

Now, avg. over volume of P particles $\langle \rangle$
 \implies exploits
 (isotropy)

$\langle \rangle$ in principle can
be different from $\underline{\underline{E}}$.

$$\begin{aligned} \nabla \cdot \langle \tilde{\underline{E}} \rangle &= \partial_x \langle \tilde{E}_x \rangle = \partial_y \langle \tilde{E}_y \rangle = \partial_z \langle \tilde{E}_z \rangle \\ &= \frac{1}{3} \langle \nabla \cdot \tilde{\underline{E}} \rangle \end{aligned}$$

so, taking $\underline{\underline{E}}$ to be: $\underline{\underline{E}} = \underline{\underline{E}} \mathbf{1}$ (general)

$$\langle \nabla \cdot \tilde{\underline{E}} \rangle = 3 \partial_x \langle \tilde{E}_x \rangle$$

so

$$3 \underline{\underline{E}} \partial_x \langle \tilde{E}_x \rangle = -\underline{\underline{E}} \cdot \nabla \tilde{\underline{E}}$$

↑
replaced

\tilde{E}_x with $\langle \tilde{E}_x \rangle$

$$\langle \tilde{E}_x \rangle = - \frac{\tilde{E}_x \tilde{E}}{3\epsilon}$$

then
mean
D

$$\tilde{E} \tilde{E} = - \tilde{E}^2 \left(\frac{\tilde{E}_x \tilde{E}}{3\epsilon} \right)$$

$$= - \frac{\tilde{E}^2}{3\epsilon} \tilde{E}_x$$

$$D = \epsilon \tilde{E} + \tilde{E} \tilde{E}$$

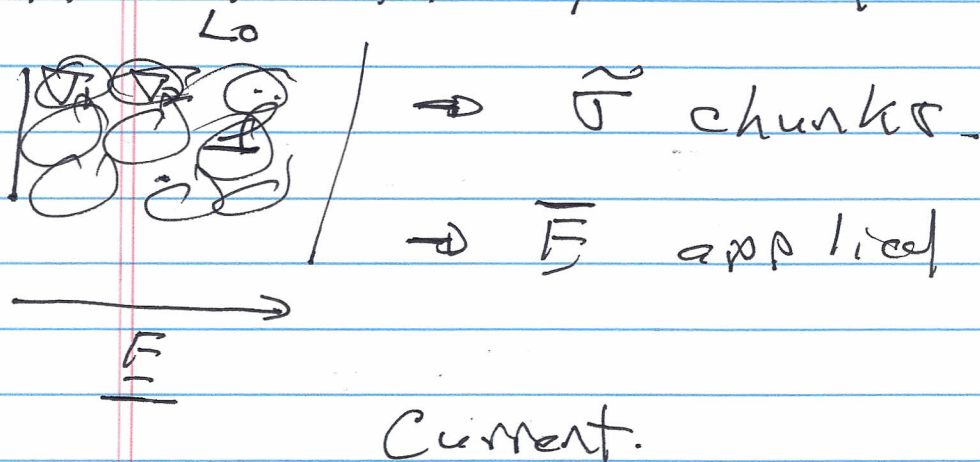
$$= \left(\epsilon - \frac{1}{3} \frac{\tilde{E}^2}{\epsilon} \right) \tilde{E}$$

$$\epsilon_{\text{eff}} = \epsilon - \left(\frac{1}{3} \right) \frac{\tilde{E}^2}{\epsilon}$$

⇒ effective permittivity of mixture
as calculated in MFT.

(2) Effective Conductivity → See Dykhne paper.

Could ask similar problem:



Have: $\underline{J} = \underline{\sigma} \underline{E}$

$\underline{\nabla} \cdot \underline{J} = 0$

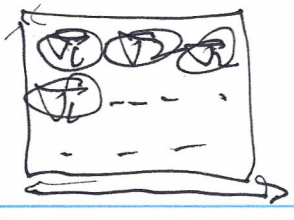
Could approach by P.T. and QLT but do better by considering symmetry.

In particular consider:

→ 2D random medium

→ consists of random assortment of two arbitrary conductivity elements $\underline{\sigma}_1, \underline{\sigma}_2$. Note $\underline{\sigma}$ need not be small.

→ consider $\underline{\sigma}_1, \underline{\sigma}_2$ as "phases", with domains $l_{1,2} \ll L_0$



conduction path convoluted for β .
 $\sigma_2 \neq \sigma_1$

→ Defines Pblm: Conductivity of 2-phase system

Assume: - 1:1 mixture
 - avg. geometrical config. equal

c.e. - equal # domains
 - no spatial dist.

$$\sigma = \sigma(x, y)$$

$$\sigma = \begin{cases} \sigma_1 \\ \sigma_2 \end{cases} \text{ have equal areas}$$

of course:

$$\begin{cases} \underline{J} = \sigma \underline{E} \\ \nabla \times \underline{E} = 0 \\ \nabla \cdot \underline{J} = 0 \end{cases}$$

Seek: σ_{eff} , s/t

$$\langle \underline{J} \rangle = \sigma_{eff} \langle \underline{E} \rangle$$

where $\langle \rangle$ represents a volume avg.

$$\langle \underline{J} \rangle = \frac{1}{V} \int d^3x \underline{J}$$

$$\langle \underline{E} \rangle = \frac{1}{V} \int d^3x \underline{E}$$

Goal is effective proportionality between mean \underline{J} , $\langle \underline{J} \rangle$; and mean \underline{E} , $\langle \underline{E} \rangle$.

N.B: - σ_{eff} is macroscopic in uniform medium.

- dispersion in σ , i.e. $\langle (\sigma - \bar{\sigma})^2 \rangle$ will increase with volume (i.e. # scats \uparrow)

Now can exploit the observation:
can write:

$\hat{n} \perp \text{plane}$

$$\left. \begin{aligned} \underline{J}' &= (\underline{\sigma}_1, \underline{\sigma}_2)^{1/2} \hat{n} \times \underline{E} \\ \underline{E}' &= (\underline{\sigma}_1, \underline{\sigma}_2)^{-1/2} \hat{n} \times \underline{J} \end{aligned} \right\} \begin{array}{l} \text{rotate} \\ \text{rescale} \end{array}$$

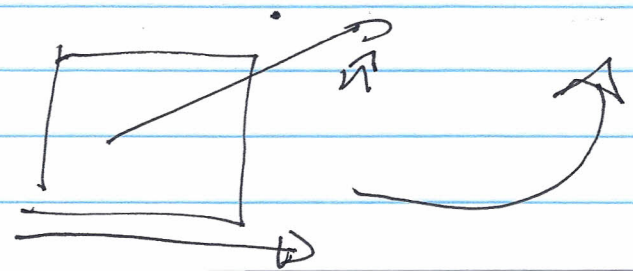
and plug into:

$$\left\{ \begin{aligned} \underline{J} &= \underline{\sigma} \underline{E} \\ \underline{\nabla} \times \underline{E} &= 0 \\ \underline{\nabla} \cdot \underline{J} &= 0 \end{aligned} \right.$$

Now

$$\begin{aligned} \underline{J}' &= (\underline{\sigma}_1, \underline{\sigma}_2)^{1/2} \hat{n} \times \underline{E} && \underline{\sigma}' \text{ tbd} \\ &= \underline{\sigma}' \underline{E}' = ((\underline{\sigma}_1, \underline{\sigma}_2)^{-1/2} \hat{n} \times \underline{J}') \underline{\sigma}' \end{aligned}$$

so, simply:



$$\underline{\underline{J'}} = \underline{\underline{V}} \times \underline{\underline{J}} = (\underline{\underline{V}}_1 \underline{\underline{V}}_2)^{1/2} \underline{\underline{V}} \times \underline{\underline{E}}$$

and

$$\underline{\underline{J}} = \underline{\underline{V}} \underline{\underline{E}}$$

$$\Rightarrow \underline{\underline{J'}} = (\underline{\underline{V}}_1 \underline{\underline{V}}_2)^{1/2} \underline{\underline{J}}$$

but since system statistically homogeneous,

$$\left. \begin{aligned} \langle \underline{\underline{J'}} \rangle &= \underline{\underline{V}}_{\text{eff}} \langle \underline{\underline{E}} \rangle \\ \langle \underline{\underline{J}} \rangle &= \underline{\underline{V}}_{\text{eff}} \langle \underline{\underline{E}} \rangle \end{aligned} \right\} (**)$$

i.e. effective conductivity same!

Now simply averaging (*)

$$\langle \underline{\underline{J'}} \rangle = (\underline{\underline{V}}_1 \underline{\underline{V}}_2)^{1/2} \underline{\underline{V}} \times \langle \underline{\underline{E}} \rangle$$

$$\langle \underline{\underline{E}} \rangle = (\underline{\underline{V}}_1 \underline{\underline{V}}_2)^{-1/2} \underline{\underline{V}} \times \langle \underline{\underline{J}} \rangle$$

and noting (**), indeed

$$\underline{\underline{V}}_{\text{eff}} = (\underline{\underline{V}}_1 \underline{\underline{V}}_2)^{1/2} \underline{\underline{V}}$$

Note:

= effective conductivity is geometric
mean of σ_1, σ_2

= exact result \Rightarrow $\left\{ \begin{array}{l} 2D \\ l_{1,2} \ll L_0 \\ \text{stat. homogeneity} \end{array} \right.$

$\left\{ \begin{array}{l} \text{explicit rotation,} \\ \text{rescaling} \\ \text{statistical symmetry} \end{array} \right.$

Now can generalize to more general
 conductivity $\sigma(x, y)$.

Take: $\chi(x, y) = \ln \sigma - \langle \ln \sigma \rangle$

\downarrow
 Fluctuation in conductivity

$\left\{ \begin{array}{l} F(1, 2, \dots | \chi) \quad \text{even} \\ \text{c.e. } \langle \chi(1) \chi(2) \rangle \text{ etc} \end{array} \right.$

similar arguments \Rightarrow

$$\sigma_{\text{eff}} = \exp \langle \ln \sigma \rangle$$

$$= \left(\frac{\langle \sigma \rangle}{\langle 1/\sigma \rangle} \right)^{1/2}$$

For Gaussian distribution of fctns.

$$\langle \ln T \rangle = \ln T - \chi$$

$$T_{\text{eff}} = \langle \exp[\ln T - \chi] \rangle$$

$$= \langle T e^{-\chi} \rangle$$

$$= \langle T (1 - \chi - \frac{\chi^2}{2} + \dots) \rangle$$

$$\approx \langle T \rangle e^{-\overline{\chi^2}/2}$$

$$\left\{ T_{\text{eff}} = \langle T \rangle e^{-A^2/2}, \quad A^2 = \langle \chi^2 \rangle \right\}$$

→ What of Distribution of Currents and Fields?

- Can calculate distribution of currents fields over phases

⇒ i.e. what are statistical properties of currents, phases?

⇒ what is avg. over a phase?

observe:

$$A \equiv \langle (\sigma - \sigma_1) \underline{E} \rangle$$

$$= \langle \underline{V} \rangle - \sigma_1 \underline{E}$$

but also, $A \neq 0$ on σ_2 phase

$$A = \langle (\sigma - \sigma_1) \underline{E} \rangle$$

$$= \langle \begin{pmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{pmatrix} \underline{E} - \sigma_1 \underline{E} \rangle$$

cancels at phase 1!

$$\equiv \frac{1}{2} (\sigma_2 - \sigma_1) \langle \underline{E}_2 \rangle$$

$$\underline{E}_2 = \frac{1}{V_2} \int_{V_2} d^3x \underline{E} \quad \rightarrow \text{avg over phase 2}$$

equating: \hookrightarrow note

$$\langle \underline{V} \rangle - \sigma_1 \underline{E} = \frac{1}{2} (\sigma_2 - \sigma_1) \langle \underline{E}_2 \rangle$$

and re-deriving using:

$$\langle J \rangle = V_{eff} \langle E \rangle$$

$$V_{eff} = (\langle V_1 V_2 \rangle)^{1/2} \langle E \rangle$$

Then,

$$\langle E_2 \rangle = \frac{2\sqrt{V_1}}{\sqrt{V_1} + \sqrt{V_2}} E$$

$$\langle E_1 \rangle = \frac{2\sqrt{V_2}}{\sqrt{V_1} + \sqrt{V_2}} E$$

and $\langle J \rangle_{1,2} = \frac{2\sqrt{V_{1,2}}}{\sqrt{V_1} + \sqrt{V_2}} \langle J \rangle$

Similarly can calculate power dissipated in each phase (equal)!

- energy dissipated same in both 1, 2
- Joule heat diff same.

and cranking results:

$$\langle E^2 \rangle = \frac{1}{2} (\sqrt{V_1/V_2} + \sqrt{V_2/V_1}) \langle E \rangle^2$$

$$\langle J^2 \rangle = \frac{1}{2} (\sqrt{V_1/V_2} + \sqrt{V_2/V_1}) \langle J \rangle^2$$

$$\frac{\langle J^2 \rangle - \langle J \rangle^2}{\langle J \rangle^2} = \frac{1}{2} \left[\left(\frac{V_1}{V_2} \right)^{1/2} - \left(\frac{V_2}{V_1} \right)^{1/2} \right]^2$$

same for E.