

# M285K - course #2

### I.3 Setting.

$$- \nabla \cdot \left( a^w(\cdot) \right) \nabla u_\varepsilon^w = \nabla \cdot f \quad \text{in } \mathbb{R}^d \quad (\text{avoid boundary issues})$$

with  $a^w: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$

measurable

& unif elliptic i.e.  
 $(\exists \alpha, \beta > 0)$

$$\begin{cases} e \cdot a^w(x) e \geq \alpha |e|^2 \\ |a^w(x) e| \leq \beta |e| \end{cases} \quad \forall x, e \in \mathbb{R}^d \\ w \in \Omega$$

Lemma (Lax-Milgram).

$$\forall w \quad \forall f \in L^2(\Omega)$$

$$\exists! \text{ solution } u_\varepsilon^w \text{ in } \dot{H}^2(\mathbb{R}^d) = \left\{ \varphi \in H_{loc}^1(\mathbb{R}^d) \right. \\ \left. \text{st } \nabla \varphi \in L^2 \right\}$$

$$\text{In addition: } \|\nabla u_\varepsilon^w\|_{L^2(\mathbb{R}^d)} \leq \frac{1}{\alpha} \|f\|_{L^2(\mathbb{R}^d)}.$$

Micro model: • periodic setting: a periodic  
• random setting: statistical ensemble of coeff fields.

Natural requirement: stationarity = "law of  $a$  is invariant under shifts"  
 $a(\cdot) \stackrel{\text{law}}{=} a(\cdot + y) \forall y.$

Definitions. Let  $(\Omega, \mathcal{F}, \mathbb{P})$   $\mu$  space.

• A random field on  $\mathbb{R}^d$  is a map  $\varphi: \mathbb{R}^d \times \Omega \rightarrow \mathbb{R}$   
st.  $\forall x \in \mathbb{R}^d: \varphi(x, \cdot): \Omega \rightarrow \mathbb{R}$  meas.

• It is stationary if the finite-dim law is shift-invariant  
i.e.  $\forall$  finite  $F \subset \mathbb{R}^d$ :

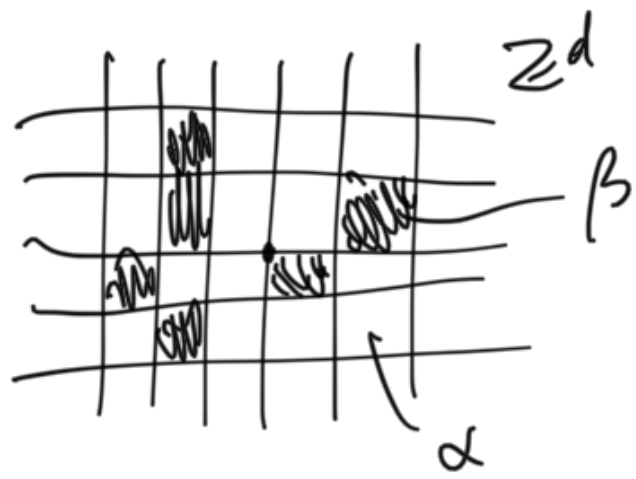
ic jointly measurable  
the law of  $\{ \vartheta(x+y, \cdot) \}_{x \in E}$  is independent of the shift.

- It is measurable if  $\vartheta: \mathbb{R}^d \times \Omega \rightarrow \mathbb{R}$  is jointly measurable.  
(In part:  $\forall \omega$ , realization  $\vartheta(\cdot, \omega): \mathbb{R}^d \rightarrow \mathbb{R}$  meas.)

Lemma / Exercise (von Neumann '32) - stochastic version of Fubini's thm.

A stationary random field is measurable  
 $\iff$  it is stochastically continuous  
ie  $\forall x \in \mathbb{R}^d, \forall \delta > 0: \mathbb{P} \left[ \lim_{|y| \rightarrow 0} |\vartheta(x+y, \cdot) - \vartheta(x, \cdot)| > \delta \right] = 0$

Examples: • random checkerboard



$\left\{ \begin{array}{l} \text{iid } \pi \nu \{V_z\}_{z \in \mathbb{Z}^d}, V_z \in \{\alpha, \beta\} \\ \text{if } U \sim \text{uniform on } Q = [-\frac{1}{2}, \frac{1}{2}]^d \end{array} \right.$

$$C_0 \tilde{a}(x) = \sum_{z \in \mathbb{Z}^d} V_z^w \mathbb{1}_{x \in U+z+Q}$$

• inclusion models



• Gaussian model:

$G$  Gaussian random field
   
 $\left\{ \begin{array}{l} \text{centered } \mathbb{E}[G(x)] = 0 \quad \forall x \\ \text{covariance } \mathbb{E}[G(x)G(y)] = k(x,y) \equiv C(x-y) \end{array} \right.$

covariance  
 set  
 $\downarrow$   
 $\Rightarrow$  stationary!

$\mathbb{E}[G(x+z)G(y+z)] = \dots$

$a_0: \mathbb{R} \xrightarrow{\text{smooth}} \mathbb{R}^{d \times d}$  with <sup>conv</sup> elliptic values  $[\alpha, \beta]$  | + assume that  $C$  continuous

Set  $a(x, w) = a_0(G(x, w))$

Rem:  $\mathbb{E} |G(x) - G(y)|^2 = \mathbb{E} [G(x)^2 + G(y)^2 - 2G(x)G(y)]$   
 $= 2(C(0) - C(x \cdot y)) \xrightarrow{|x \cdot y| \rightarrow 0} 0$

$\Rightarrow G$  stoch ds  $\Rightarrow G$  meas stat r.f.

In the sequel:  $(\Omega, \mathcal{F}, \mathbb{P})$  generated by  $a$  (coeff field)

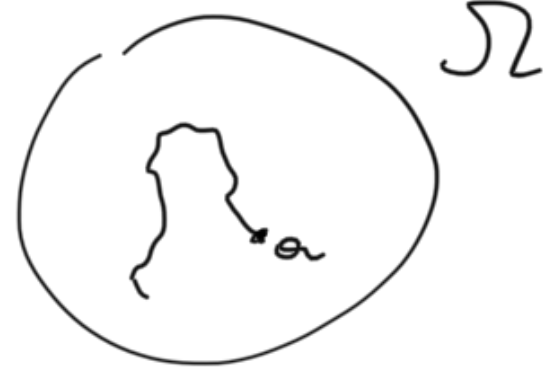
More precisely, concrete choice:

- $\Omega = \{ \text{coeff fields } a: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d} \text{ meas } \& \text{ unif all } [\alpha, \beta] \}$
- $\mathcal{F} = \sigma\text{-algebra generated by } \{ a \mapsto \int_{\mathbb{R}^d} \varphi : \varphi \in C_c^\infty(\mathbb{R}^d) \}$
- ~~examp~~  $(\mathbb{R}^d, +)$  acts on  $\Omega$ :  $(\tau_z a)(x) = a(x+z)$

- stationarity:  $\mathbb{P}[\tau_z A] = \mathbb{P}[A] \quad \forall z \in \mathbb{R}^d \quad \forall A \in \mathcal{F}$ .

Rem.  $\left\{ \begin{array}{l} \tau \text{ measure-preserving action of } (\mathbb{R}^d, +) \text{ on } (\Omega, \mathcal{F}, \mathbb{P}) \\ \hookrightarrow \tau_0 = \text{id}, \tau_x \tau_y = \tau_{x+y} \end{array} \right. \mid \text{dynamical system}$

$\mathbb{R}^d \times \Omega \rightarrow \Omega$   
 $(z, \omega) \mapsto \tau_z \omega$  meas.



- $\varphi: \mathbb{R}^d \times \Omega \rightarrow \mathbb{R}$  is  $\tau$ -stationary  
 if  $\varphi(x, \omega) = \varphi(0, \underbrace{\omega(+x)}_{\tau_x \omega})$  ( $\Rightarrow \varphi$  is meas. stat random field in the previous sense)

- ergodicity:  $\left[ \begin{array}{l} \text{if } A \in \mathcal{F} \text{ satisfies } \tau_z A = A \quad \forall z \\ \text{then } \mathbb{P}A \text{ is } 0 \text{ or } 1. \end{array} \right.$

$\hookrightarrow$   $\Omega$  (minimal etc. etc.)

$\mathbb{T}$   (invariant condition)

Ergodic theorem (pointwise version - dt setting) (Birkhoff, Khinchin).

If  $\tau$  ergodic on  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  
Then  $\forall \varphi \in L^1(\Omega)$  we have  $\forall \omega \in \Omega$ :

$$\frac{1}{R^d} \int_{\mathbb{R}^d} \varphi(\tau_z \omega) dz \xrightarrow{R \rightarrow \infty} \mathbb{E}[\varphi].$$

$\sim$  generalization of LLN for iid random variables.

(Proof: difficult - SLLN, much easier:  $L^2$  conv. result, Von Neumann mean erg. thm)

Interpretation ( $d=1$ ):  $\varphi = \mathbb{1}_E$

$$\frac{1}{R} \int_{\mathbb{R}} \mathbb{1}_E(\tau_z \omega) dz \rightarrow \mathbb{P}[E]$$



$$\underbrace{\mathbb{R} \cup \{\infty\}}_{\text{overaged amount of time spent in } E}$$

overaged amount of time spent in  $E$

Time average

$\rightarrow$  statistical average

Corollary.  $\left( \begin{array}{l} \forall \varphi: \mathbb{R}^d \times \Omega \rightarrow \mathbb{R} \text{ } \tau\text{-stat.}, \mathbb{E}|\varphi| < \infty \\ \text{Then } \forall' a: \varphi(\frac{\cdot}{\varepsilon}, a) \xrightarrow{*L^\infty} \mathbb{E}[\varphi] \end{array} \right.$

Proof. Show that  $\forall g \in L^1(\mathbb{R}^d): \int g \varphi(\frac{\cdot}{\varepsilon}, a) \rightarrow \mathbb{E}[\varphi] \int g$ .

$$\textcircled{1} g = \mathbb{1}_Q : \int g \varphi(\frac{\cdot}{\varepsilon}, a) = \int_Q \varphi(\frac{\cdot}{\varepsilon}, a) = \varepsilon^d \int_{\frac{1}{\varepsilon}Q} \underbrace{\varphi(z, a)}_{\varphi(0, \tau_2 a)} dz$$

$$(\forall' a) \xrightarrow{\text{erg thm}} \mathbb{E}[\varphi] \int \mathbb{1}_Q.$$

(2)  $g = \sum \alpha_n \uparrow \downarrow \theta_n$   
 the opposite!

△

Stronger assumption:  $\tau$ -mixing if  $\mathbb{P}[A \cap \tau B] \xrightarrow{|\tau| \rightarrow \infty} \mathbb{P}A \mathbb{P}B$ .



decondition.

Quantitative: e.g. Rosenblatt  $\alpha$ -mixing

•  $G_1, G_2 \subset \mathcal{F}$  sub- $\sigma$ -alg.

$$\alpha_0(G_1, G_2) = \sup \left\{ | \mathbb{P}(G_1 \cap G_2) - \mathbb{P}G_1 \mathbb{P}G_2 | : \begin{array}{l} G_1 \in G_1 \\ G_2 \in G_2 \end{array} \right\}$$

•  $d(\mathcal{R}, \mathcal{D}) = \sup \left\{ \alpha_0(\sigma(a|_{S_1}), \sigma(a|_{S_2})) : S_1, S_2 \in \mathcal{R}^d \right\}$

average diameters

$\text{dist}(S_1, S_2) \geq R, \text{diam}(S_i) \leq D$

•  $\alpha$ -mixing if  $\alpha(R, D) \xrightarrow{R \rightarrow \infty} 0$  ( $\forall D$ )

Example: finite range of dependence  $\Leftrightarrow \alpha(R, D) = 0 \forall R \geq R_0$   
 $\Leftrightarrow a|_{S_1} \perp a|_{S_2} \forall \text{dist}(S_1, S_2) \geq R_0$

### I.4 Explicit 1D case ( $d=1$ )

$$- \left( a\left(\frac{\cdot}{2}\right) u'_\varepsilon \right)' = f' \text{ in } \mathbb{R}, \quad u'_\varepsilon \in L^2(\mathbb{R})$$
$$(f \in L^2(\mathbb{R}))$$

$$- a\left(\frac{\cdot}{2}\right) u'_\varepsilon = f \quad \text{~~is~~}$$

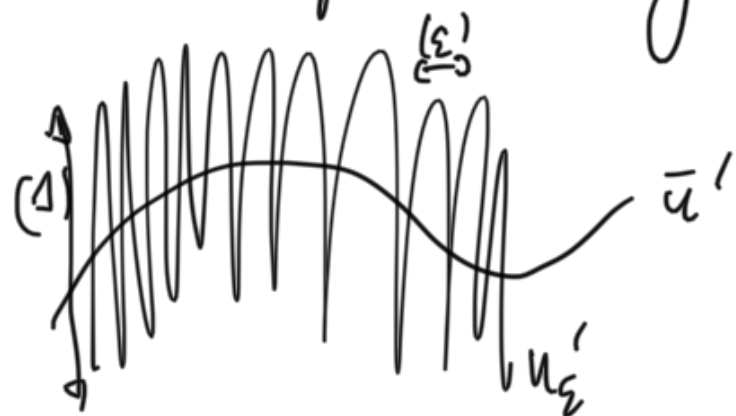
$$u'_\varepsilon = - \frac{1}{a\left(\frac{\cdot}{2}\right)} f$$

trivial in 1D: sol-operator = mult by  $\frac{1}{a(\frac{\cdot}{2})}$

1D 1D 1D-1D 1D

(cf in 1D:  $V \Delta V = Id$ )

Lemma (1D quad. homog.)



Proof: algorithm  $\Delta$

If a stat. ergodic,  
 Then  $u_\epsilon' \xrightarrow{L^2(\mathbb{R})} \underbrace{\mathbb{E}\left[\frac{1}{a}\right]}_f = \bar{u}'$   
 where  $-(\bar{a} \bar{u}')' = f'$  in  $\mathbb{R}$   
 with  $\bar{a} = \mathbb{E}\left[\frac{1}{a}\right]^{-1}$   
 geom average.

1D: no geometry  
 (11 resistances in series)

Lemma (corrector result).  
 = description of oscillation

If a stat. erg.  
 Then  $u_\epsilon' - \bar{u}' \left(1 + \underbrace{\left(\frac{\bar{a}}{a(x)}\right)^{-1}}_{\rightarrow 0}\right) = 0$

Rem. Less trivial for Dirichlet BC on  $(0,1)$ :  $\begin{cases} -(a(x) v_x')' = f, [0,1] \\ v_x(0) = v_x(1) = 0. \end{cases}$

Again  $v_x' \rightarrow \bar{v}'$ :  $\begin{cases} -(\bar{a} \bar{v}')' = f \\ \bar{v}(0) = \bar{v}(1) = 0. \end{cases}$

$$\underbrace{(v_a' - \bar{v}')}_{\text{relevant oscillation}} \left( 1 + \underbrace{\left( \frac{\bar{a}}{a(x)} - 1 \right)}_{\text{oscillation}} \right) = \underbrace{\frac{1}{a(x)} \left( -\int_0^1 f + \frac{\int_0^1 f \frac{1}{a(x)}}{\int_0^1 \frac{1}{a(x)}} \right)}_{\rightarrow 0 \forall a \text{ by avg. thm.}} \triangleleft$$

Convergence rate: need quant error estimate for  $\int_0^1 f \frac{1}{a(x)} \xrightarrow{\varepsilon \rightarrow 0} \mathbb{E}\left[\frac{1}{a}\right] \int_0^1 f$

$\rightarrow$  need quant. avg., e.g. mixing.

Here explicit:  $\mathbb{E} \left| \int_0^1 f \frac{1}{a(z)} - \mathbb{E} \left[ \frac{1}{a} \right] \int_0^1 f \right|^2 = \text{Var} \left[ \int_0^1 f \frac{1}{a(z)} \right]$

$$= \int_0^1 dx \int_0^1 dy f(x) f(y) \underbrace{\text{Cov} \left[ \frac{1}{a(x)}, \frac{1}{a(y)} \right]}$$

stat:  $\text{Cov} \left[ \frac{1}{a(x-\frac{y}{2})}, \frac{1}{a(y)} \right]$

$\begin{matrix} x \rightarrow y + \varepsilon x \\ y \rightarrow y \end{matrix}$

$$\leq \varepsilon \int_0^{\frac{1}{\varepsilon}} dx \int_0^1 dy \underbrace{f(y + \varepsilon x) f(y)}_{\leq \frac{1}{2} (|f(y + \varepsilon x)|^2 + |f(y)|^2)} \text{Cov} \left[ \frac{1}{a(x)}, \frac{1}{a(y)} \right].$$

$$\leq \varepsilon \left( \int_0^1 |f|^2 \right) \cdot \underbrace{\int_{-\infty}^{\infty} dx \left| \text{Cov} \left[ \frac{1}{a(x)}, \frac{1}{a(y)} \right] \right|}.$$

$\rightarrow$  rate  $O(\sqrt{\varepsilon})$  in  $L^2(\Omega)$  if dens of coned of  $\frac{1}{a}$

e.g. a  $\alpha$ -mixing with integrable  $\alpha$ -mixing coeff

$$\Rightarrow |Cov(\frac{1}{a(x)}, \frac{1}{a(y)})| \leq a(|x-y|, D=0)$$

Exercise:  $\begin{cases} E[|\vartheta_n - \bar{\vartheta}|^2(x)]^{1/2} \leq \sqrt{\varepsilon} \\ E[|\vartheta'_n - \bar{\vartheta}'(1 + \underbrace{(\frac{\bar{a}}{a(x)} - 1)})|^2(x)]^{1/2} \leq \sqrt{\varepsilon}. \end{cases}$

Fluctuations: macro observable  $\int g u_n$ ,  $g \in C_c^\infty(\mathbb{R}^d)$ .

$$= \int g g \Theta_n$$

a) convergence of variances:  $Var \left[ \frac{1}{\sqrt{\varepsilon}} \int g u_n \right] \rightarrow \int |g|^2 \left( \int Cov(\frac{1}{a(x)}, \frac{1}{a(y)}) dx \right)$

$u_i$   
scaling

b) asymptotic normality: stronger mixing, use favorite proof...

$$d_{TV} \left( \frac{\varepsilon^{-1/2} \int g u_\varepsilon' - \mathbb{E}[g]}{\text{Var}[g]^{1/2}}, \mathcal{N} \right) \leq \sqrt{\varepsilon}.$$

↑  
standard  
Gauss

---

No lecture on Friday 8/1.

Next lecture: Monday 11/1!