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Linear Frequency Principle.

§ 1. neural tangent kernel (NTK) & frequency principle (FP)

- 2-layer NN: $f(x; \theta) = f_\theta(x) = \sum_{k=1}^m a_k \sigma(w_k \cdot x + b_k)$

parameter: $\theta = \text{vec}\{a_k, w_k, b_k\}_{k=1}^m$

sample : $S = \{(x_i, y_i)\}_{i=1}^n$, $\{x_i\}_{i=1}^n \stackrel{\text{i.i.d.}}{\sim} D(\Omega)$, $y_i = f(x_i)$
 $\Omega \subset \mathbb{R}^d$

empirical risk $R_S(\theta) = \frac{1}{2} \sum_{i=1}^n (f_\theta(x_i) - y_i)^2$

gradient descent (GD) : $\begin{cases} \dot{\theta} = -\nabla_\theta R_S(\theta) \\ \theta(0) = \theta_0 \end{cases}$ (1)

- linearization of NN

$$f(x, \theta) = f(x, \theta_0) + \nabla_\theta f(x, \theta_0) \cdot (\theta - \theta_0)$$

$$R_S^{lin}(\theta) = \frac{1}{2} \sum_{i=1}^n (f^{lin}(x_i, \theta) - y_i)^2$$

GD : $\begin{cases} \dot{\theta} = -\nabla_\theta R_S^{lin}(\theta) \\ \theta(0) = \theta_0 \end{cases}$

still denote as

$$f(x, \theta) = f(x, \theta_0) + \nabla_\theta f(x, \theta_0) \cdot (\theta - \theta_0)$$

$$R_S(\theta) = \frac{1}{2} \sum_{i=1}^n (f(x_i, \theta) - y_i)^2$$

GD : $\begin{cases} \dot{\theta} = -\nabla_\theta R_S(\theta) \\ \theta(0) = \theta_0 \end{cases}$ (2)

Remark : $\nabla_\theta f(x, \theta) = \nabla_\theta f(x, \theta_0)$

- neutral tangent kernel (NTK)

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$$\begin{aligned}
 (2) \Rightarrow \frac{d}{dt} f(x, \theta) &= \nabla_{\theta} f(x, \theta) \cdot \dot{\theta} \\
 &= -\nabla_{\theta} f(x, \theta) \cdot \nabla_{\theta} R_S(\theta) \\
 &= -\nabla_{\theta} f(x, \theta) \cdot \sum_{i=1}^n \nabla_{\theta} f(x_i, \theta) (f(x_i, \theta) - y_i) \\
 &= -\sum_{i=1}^n K_m(x, x_i) (f(x_i, \theta) - y_i)
 \end{aligned}$$

$$\begin{aligned}
 \text{NTK : } K_m(x, x') &= \nabla_{\theta} f(x, \theta) \cdot \nabla_{\theta} f(x', \theta) \\
 &\quad \uparrow \quad \uparrow \\
 &= \nabla_{\theta} f(x, \theta_0) \cdot \nabla_{\theta} f(x', \theta_0) \\
 &\quad \xrightarrow{\text{only depends on } \theta_0} \\
 &\quad x, x' \in \mathcal{S}
 \end{aligned}$$

linear dynamics with NTK :

$$\frac{d}{dt} (f(x, \theta(t)) - f(x)) = -\sum_{i=1}^n K_m(x, x_i) (f(x_i, \theta(t)) - f(x_i)) \quad (3)$$

$$\text{Let } u(x, t) = f(x, \theta(t)) - f(x) \quad (u(x, t) \text{ depends on } \theta)$$

$$p(x) = \sum_{i=1}^n \delta(x_i)$$

$$u_p(x, t) = u(x, t) p(x)$$

$$\Rightarrow \frac{d}{dt} u(x, t) = - \int_{\mathbb{R}^d} K_m(x, x') u_p(x', t) dx' \quad (4)$$

- NTK for 2-layer NN

$$f_{\theta}(x) = \frac{1}{\sqrt{m}} \sum_{k=1}^m a_k \sigma(w_k \cdot x + b_k) = \frac{1}{\sqrt{m}} \sum_{k=1}^m \sigma^*(x, g_k)$$

generalized coordinates $g_k = (a_k, w_k, b_k)$

$$\begin{aligned}
K_m(x, x') &= \frac{1}{m} \nabla_{\theta} f(x, \theta_0) \cdot \nabla_{\theta} f(x', \theta_0) = \frac{1}{m} \sum_{k=1}^m \nabla_{g_k} \sigma^*(x, g_k) \cdot \nabla_{g_k} \sigma^*(x', g_k) \quad (3) \\
&= \frac{1}{m} \sum_{k=1}^m \left(\nabla_{a_k} \sigma^*(x, g_k) \cdot \nabla_{a_k} \sigma^*(x', g_k) \right. \\
&\quad + \nabla_{w_k} \sigma^*(x, g_k) \cdot \nabla_{w_k} \sigma^*(x', g_k) \\
&\quad \left. + \nabla_{b_k} \sigma^*(x, g_k) \cdot \nabla_{b_k} \sigma^*(x', g_k) \right) \\
&= \frac{1}{m} \sum_{k=1}^m \left(\sigma(w_k^T x + b_k) \sigma(w_k^T x' + b_k) \right. \\
&\quad + a_k^2 \sigma'(w_k^T x + b_k) a_k^2 \sigma'(w_k^T x' + b_k) (x \cdot x') \\
&\quad \left. + a_k^2 \sigma'(w_k^T x + b_k) a_k^2 \sigma'(w_k^T x' + b_k) \right)
\end{aligned}$$

as $m \rightarrow \infty$, $K_m(x, x') = K(x, x') + O(\frac{1}{\sqrt{m}})$

$$\begin{aligned}
K(x, x') &:= \mathbb{E}_{g_k} \nabla_{g_k} \sigma^*(x, g_k) \cdot \nabla_{g_k} \sigma^*(x', g_k) \\
&= \mathbb{E}_g \left(\sigma(w \cdot x + b) \sigma(w \cdot x' + b) + a^2 \sigma'(w \cdot x + b) \sigma'(w \cdot x' + b) (x \cdot x') \right. \\
&\quad \left. + a^2 \sigma'(w \cdot x + b) \sigma'(w \cdot x' + b) \right)
\end{aligned}$$

$$g = (a, w, b), \text{ usually } a \sim N(0, \sigma_a^2)$$

$$w \sim N(0, \sigma_w^2 I_d)$$

$$b \sim N(0, \sigma_b^2)$$

- frequency principle (FP)

first observed by experiments, later some rigorous results for general DNN (or even continuous parametric models) are available.

FP phenomenon: DNNs often fit target functions from low to high frequencies in (GD) training dynamics

Question: for GD of DNN, can we write down its dependence on different frequencies?

Answer:

✓

This can be done for the GD of a very wide ($m \rightarrow +\infty$) linearized 2-layer NN.

- informal derivation of LFP (linear frequency principle)

recall $\frac{d}{dt} u(x) = - \int_{\mathbb{R}^d} K(u, x') u_p(x') dx' + O(\frac{1}{m})$

drop lower order term $O(\frac{1}{m})$

regard $u(\cdot)$ as a column vector, $K(\cdot, \cdot)$ as a square matrix

$$\Rightarrow " \frac{d}{dt} u = -K u_p "$$

regard the Fourier transform as a matrix

$$F : u \mapsto F[u] = \hat{u}$$

$$= u \mapsto Fu''$$

$$F^{-1} : u \mapsto F^{-1}[u]$$

$$= u \mapsto F^{-1}u''$$

$$\Rightarrow " \frac{d}{dt} Fu = -F K u_p "$$

$$= -(FKF^{-1})Fu_p$$

$$\frac{d}{dt} \hat{u} = -\hat{K} \hat{u}_p$$

$$" \hat{K} := FKF^{-1} "$$

goal: calculate \hat{K} rigorously.

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§2 derivation of LFP dynamics

- Fourier transform on \mathbb{R}^d

for $g \in L^1(\mathbb{R}^d)$

$$\mathcal{F}[g](\xi) := \mathcal{F}_{x \rightarrow \xi}[g](\xi) := \int_{\mathbb{R}^d} g(x) e^{-2\pi i x \cdot \xi} dx$$

inverse Fourier transform

$$\mathcal{F}^{-1}[g](x) := \mathcal{F}_{\xi \rightarrow x}^{-1}[g](x) := \int_{\mathbb{R}^d} g(\xi) e^{2\pi i x \cdot \xi} d\xi$$

Recall: $\sigma = \text{ReLU}, \tanh, \dots$

not smooth, not in $L^2 \dots$

L^1/L^2 theory of Fourier transform is not suitable
distribution / generalized function is required

Schwartz space $S(\mathbb{R}^d)$

space of tempered distributions $S'(\mathbb{R}^d)$

$$\psi \in S'(\mathbb{R}^d), \phi \in S(\mathbb{R}^d), \langle \psi, \phi \rangle := \langle \psi, \phi \rangle_{S'(\mathbb{R}^d), S(\mathbb{R}^d)} := \psi(\phi)$$

Fourier transform of $\psi \in S'(\mathbb{R}^d)$

$$\langle \mathcal{F}[\psi], \phi \rangle = \langle \psi, \mathcal{F}[\phi] \rangle \quad \forall \phi \in S(\mathbb{R}^d)$$

- Prop. 1. $\langle 2\pi \mathcal{F}[u], \phi \rangle = \langle L[\mathcal{F}[u_p]], \phi \rangle + O\left(\frac{1}{\sqrt{m}}\right)$

$L[\cdot]$ is called LFP operator

$$L[\mathcal{F}[u_p]] = - \int_{\mathbb{R}^d} \hat{K}(\xi, \xi') \mathcal{F}[u_p](\xi') d\xi'$$

$$\hat{K}(\xi, \xi') := \mathbb{E}_g K_g(\xi, \xi') \quad (6)$$

$$:= \mathbb{E}_g \mathbb{F}_{x \rightarrow \xi} [\nabla_g \sigma^*(x, g)] \cdot \overline{\mathbb{F}_{x' \rightarrow \xi'} [\nabla_g \sigma^*(x', g)]}$$

Proof: Recall

$$\begin{aligned} \partial_t u(x, t) &= - \int_{\mathbb{R}^d} K(x, x') u_p(x', t) dx' + O(\frac{1}{m}) \\ &\approx - \int_{\mathbb{R}^d} K(x, x') u_p(x', t) dx'. \end{aligned}$$

use " \approx " to keep leading order

$$\forall \phi \in \mathcal{S}(\mathbb{R}^d)$$

$$\langle \mathbb{F}[\partial_t u], \phi \rangle = \langle \partial_t u, \mathbb{F}[\phi] \rangle$$

$$= \int_{\mathbb{R}^d} \partial_t u(x, t) \int_{\mathbb{R}^d} \phi(\xi) e^{-i 2\pi x \cdot \xi} d\xi dx \quad (\because \partial_t u \text{ locally integrable})$$

$$\approx - \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} K(x, x') u_p(x', t) dx' \int_{\mathbb{R}^d} \phi(\xi) e^{-i 2\pi x \cdot \xi} d\xi dx$$

$$= - \int_{\mathbb{R}^{3d}} K(x, x') u_p(x', t) dx' \phi(\xi) e^{-i 2\pi x \cdot \xi} d\xi dx$$

$$= - \int_{\mathbb{R}^{3d}} \mathbb{E}_g \nabla_g \sigma^*(x, g) \cdot \nabla_g \sigma^*(x', g) u_p(x', t) dx' \phi(\xi) e^{-i 2\pi x \cdot \xi} d\xi dx$$

$$= - \mathbb{E}_g \int_{\mathbb{R}^d} \nabla_g \sigma^*(x, g) u_p(x', t) dx' \cdot \int_{\mathbb{R}^{2d}} \nabla_g \sigma^*(x, g) e^{-i 2\pi x \cdot \xi} \phi(\xi) d\xi dx$$

$$= - \mathbb{E}_g \int_{\mathbb{R}^d} \nabla_g \sigma^*(x, g) u_p(x', t) dx' \cdot \langle \mathbb{F}_{x \rightarrow \cdot} [\nabla_g \sigma^*(x, g)](\cdot), \phi(\cdot) \rangle$$

$$\text{Since } \int_{\mathbb{R}^d} D_g \sigma^*(x', g) u_p(x') dx'$$

$$= \int_{\mathbb{R}^d} \overline{F_{x' \rightarrow \xi'} [D_g \sigma^*(x', g)](\xi')} F_{x' \rightarrow \xi'} [u_p](\xi') d\xi'$$

$$\langle \partial_t F[u], \phi \rangle$$

$$\approx - E_g \int_{\mathbb{R}^d} \overline{F_{x' \rightarrow \xi'} [D_g \sigma^*(x', g)](\xi')} F_{x' \rightarrow \xi'} [u_p](\xi') d\xi'$$

$$\cdot \langle F_{x' \rightarrow \cdot} [D_g \sigma^*(x', g)](\cdot), \phi(\cdot) \rangle$$

$$= - \int_{\mathbb{R}^{2d}} E_g \overline{F_{x' \rightarrow \xi'} [D_g \sigma^*(x', g)](\xi')} F_{x' \rightarrow \xi} [D_g \sigma^*(x', g)](\xi)$$

$$F_{x' \rightarrow \xi} [u_p](\xi') d\xi' \phi(\xi) d\xi$$

$$= - \int_{\mathbb{R}^{2d}} \hat{f}(\xi, \xi') F[u_p](\xi') d\xi' \phi(\xi) d\xi$$

$$= \langle L[F[u_p]], \phi \rangle$$

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§3 explicit expression for LFP dynamics with ReLU activation

- Def: delta-like function $\delta_w: S(\mathbb{R}^d) \rightarrow \mathbb{R}$

$$\langle \delta_w, \phi \rangle = \int_{\mathbb{R}} \phi(yw) dy$$

- Lem 1: \forall nonzero $w \in \mathbb{R}^d$, let $\hat{w} = \frac{w}{\|w\|}$

$$\text{then } \frac{1}{\|w\|^d} \delta_{\hat{w}} \left(\frac{x}{\|w\|} \right) = \delta_w(x)$$

Proof: $\forall \phi \in S(\mathbb{R}^d)$

$$\left\langle \frac{1}{\|w\|^d} \delta_{\hat{w}} \left(\frac{\cdot}{\|w\|} \right), \phi(\cdot) \right\rangle = \langle \delta_{\hat{w}}(\cdot), \phi(\|w\|\cdot) \rangle$$

$$= \int_{\mathbb{R}} \phi(\|w\|y \hat{w}) dy$$

$$= \int_{\mathbb{R}} \phi(yw) dy$$

$$= \langle \delta_w(\cdot), \phi(\cdot) \rangle$$

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Lem 2: \forall unit vector $v \in \mathbb{R}^d$ \forall nonzero $w \in \mathbb{R}^d$

$$\text{let } \hat{w} := \frac{w}{\|w\|} \quad \forall g \in S'(\mathbb{R}^d)$$

then in the sense of distribution

$$(a) \quad \mathcal{F}_{x \rightarrow \xi}[g(v \cdot x)](\xi) = \delta_v(\xi) \mathcal{F}[g](\xi \cdot v)$$

$$(b) \quad \mathcal{F}_{x \rightarrow \xi}[g(w \cdot x + b)](\xi) = \delta_w(\xi) \mathcal{F}[g]\left(\frac{\xi \cdot \hat{w}}{\|w\|}\right) e^{2\pi i \frac{b}{\|w\|} \xi \cdot \hat{w}}$$

$$(c) \quad \mathcal{F}_{x \rightarrow \xi}[x g(w \cdot x + b)](\xi) = \frac{i}{2\pi} \nabla_\xi \left[\delta_w(\xi) \mathcal{F}[g]\left(\frac{\xi \cdot \hat{w}}{\|w\|}\right) e^{2\pi i \frac{b}{\|w\|} \xi \cdot \hat{w}} \right]$$

Proof: $\forall \phi \in S(\mathbb{R}^d)$

$$(a) \quad \langle \mathcal{F}_{x \rightarrow \cdot}[g(v \cdot x)](\cdot), \phi(\cdot) \rangle_{S'(\mathbb{R}^d), S(\mathbb{R}^d)}$$

$$= \langle g(v^\top \cdot), \mathcal{F}_{x \rightarrow \cdot}[\phi(x)](\cdot) \rangle_{S'(\mathbb{R}^d), S(\mathbb{R}^d)}$$

$$= \langle g(\cdot), \mathcal{F}_{y \rightarrow \cdot}[\phi(yv)](\cdot) \rangle_{S'(\mathbb{R}), S(\mathbb{R})}$$

$$= \langle \mathcal{F}_{y \rightarrow \cdot}[g(y)](\cdot), \phi(\cdot \cdot v) \rangle_{S'(\mathbb{R}), S(\mathbb{R})}$$

$$= \langle \mathcal{F}[g](\cdot \cdot v^\top v), \phi(\cdot \cdot v) \rangle_{S'(\mathbb{R}), S(\mathbb{R})}$$

$$= \langle \delta_v(\cdot) \mathcal{F}[g](\cdot \cdot v^\top v), \phi(\cdot) \rangle_{S'(\mathbb{R}^d), S(\mathbb{R}^d)}$$

$$(b) \quad \mathcal{F}_{x \rightarrow \xi}[g(\hat{w} \cdot x)](\xi) = \delta_{\hat{w}}(\xi) \mathcal{F}[g](\xi \cdot \hat{w})$$

note that

$$\mathcal{F}_{x \rightarrow \xi}[g(x - x_0)](\xi) = \mathcal{F}_{x \rightarrow \xi}[g](\xi) e^{-2\pi i x_0 \cdot \xi}$$

then

$$\mathcal{F}_{x \rightarrow \xi}[g(\hat{w} \cdot x + b)](\xi) = \mathcal{F}_{x \rightarrow \xi}[g(\hat{w} \cdot (x + b\hat{w}))](\xi)$$

$$= \delta_w(\xi) F[g](\xi \cdot \hat{w}) e^{2\pi i b \hat{w} \cdot \xi} \quad (9)$$

$$\begin{aligned} & \therefore F_{x \rightarrow \xi} [g(w \cdot x + b)](\xi) \\ &= F_{x \rightarrow \xi} [g(\hat{w} \cdot \|w\| x + b)](\xi) \\ &= \frac{1}{\|w\|^d} F_{x \rightarrow \xi} [g(\hat{w} \cdot x + b)]\left(\frac{\xi}{\|w\|}\right) \\ &= \frac{1}{\|w\|^d} \delta_{\hat{w}}\left(\frac{\xi}{\|w\|}\right) F[g]\left(\frac{\xi \cdot \hat{w}}{\|w\|}\right) e^{2\pi i \frac{b}{\|w\|} \hat{w} \cdot \xi} \\ &= \delta_w(\xi) F[g]\left(\frac{\xi \cdot \hat{w}}{\|w\|}\right) e^{2\pi i \frac{b}{\|w\|} \hat{w} \cdot \xi} \end{aligned}$$

(c) & $\forall \tilde{g} \in S'(\mathbb{R}^d)$

$$\begin{aligned} F_{x \rightarrow \xi} [x \tilde{g}(x)](\xi) &= \frac{1}{2\pi} \nabla_\xi [F[\tilde{g}](\xi)] \\ \Rightarrow (c) & \quad \# \end{aligned}$$

• apply the lemma with

$$g_1(w \cdot x + b) = \begin{pmatrix} \partial_a [a \sigma(w \cdot x + b)] \\ \partial_b [a \sigma(w \cdot x + b)] \end{pmatrix}$$

$$\text{and } x g_2(w \cdot x + b) = \nabla_w [a \sigma(w \cdot x + b)]$$

$$\cdot \text{Thm } \langle \partial_t F[u] , \phi \rangle = \langle L[F[u_p]] , \phi \rangle + O(m^{-k}) + O(\sigma_b^{-3})$$

$$\begin{aligned} & L[F[u_p]](\xi) \\ &= -\frac{1}{(2\pi)^{d/2} \sigma_b \|\xi\|^{d+1}} \mathbb{E}_{a,r} \left[\frac{1}{r} F[g_1] \left(\frac{\|\xi\|}{r} \right) \cdot F[g_2] \left(-\frac{\|\xi\|}{r} \right) \right] F[u_p](\xi) \\ &+ \frac{1}{(2\pi)^{d/2} \sigma_b} \nabla \cdot \left\{ \mathbb{E}_{a,r} \left[\frac{1}{r \|\xi\|^{d+1}} F[g_1] \left(\frac{\|\xi\|}{r} \right) F[g_2] \left(-\frac{\|\xi\|}{r} \right) \right] \nabla F[u_p](\xi) \right\} \end{aligned}$$

$$\text{where } r = \|w\|$$

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Remark: if $\Gamma = \text{ReLU}$, then

$$\begin{aligned} & L[\mathcal{F}[u_p](\xi)] \\ &= -\frac{1}{(2\pi)^{5/2}\sigma_b} \mathbb{E}_{a,r} \left[\frac{r^3}{(6\pi^4/\|\xi\|^{d+3})} + \frac{a^2 r}{4\pi^2/\|\xi\|^{d+1}} \right] \mathcal{F}[u_p](\xi) \\ &\quad + \frac{1}{(2\pi)^{5/2}\sigma_b} \nabla \cdot \left\{ \mathbb{E}_{a,r} \left[\frac{a^2 r}{4\pi^2/\|\xi\|^{d+1}} \right] \nabla \mathcal{F}[u_p](\xi) \right\} \end{aligned}$$

Proof. $\hat{K} = \mathbb{E}_g \hat{K}_g$, $\hat{K}_g = \hat{K}_{a,b} + K_w$

$$\begin{aligned} K_w(\xi, \xi') &= \mathcal{F}_{x \rightarrow \xi} [x g_2(w \cdot x + b)](\xi) \cdot \overline{\mathcal{F}_{x' \rightarrow \xi'} [x' g_2(w \cdot x' + b)](\xi')} \\ \hat{K}_{a,b}(\xi, \xi') &= \mathcal{F}_{x \rightarrow \xi} [g_1(w \cdot x + b)](\xi) \cdot \overline{\mathcal{F}_{x' \rightarrow \xi'} [g_1(w \cdot x' + b)](\xi')} \\ &= \delta_w(\xi) \delta_w(\xi') \mathcal{F}[g_1]\left(\frac{\xi \cdot \hat{w}}{\|w\|}\right) \cdot \overline{\mathcal{F}[g_1]\left(\frac{\xi' \cdot \hat{w}}{\|w\|}\right)} e^{2\pi i b(\xi - \xi') \cdot \frac{\hat{w}}{\|w\|}} \end{aligned}$$

$$\begin{aligned} \langle \phi, \psi \rangle_{K_{a,b}} &:= \int_{\mathbb{R}^d} \phi(\xi) \int_{\mathbb{R}^d} \hat{K}_{a,b}(\xi, \xi') \psi(\xi') d\xi' d\xi \\ &= \int_{\mathbb{R} \times \mathbb{R}} \phi(\eta w) \psi(\eta' w) \mathcal{F}[g_1](\eta) \cdot \overline{\mathcal{F}[g_1](\eta')} e^{2\pi i b(\eta - \eta')} d\eta d\eta' \end{aligned}$$

Recall that $b \sim \mathcal{N}(0, \sigma_b^2)$

$$\rho_b(b) = \frac{1}{\sqrt{2\pi}\sigma_b} e^{-\frac{b^2}{2\sigma_b^2}}$$

$$\mathcal{F}[\rho_b](\eta) = e^{-2\pi^2\sigma_b^2\eta^2}$$

$$\begin{aligned} \mathbb{E}_b(e^{2\pi i b(\eta - \eta')}) &= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}\sigma_b} e^{-b^2/2\sigma_b^2} e^{2\pi i b(\eta - \eta')} db \\ &= \mathcal{F}[\rho_b](-(\eta - \eta')) \\ &= e^{-2\pi^2\sigma_b^2(\eta - \eta')^2} \end{aligned}$$

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$$\begin{aligned}
\mathbb{E}_b \langle \phi, \psi \rangle_{K_{ab}} &= \int_{R \times R} \phi(\eta w) \psi(\eta' w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta')} \mathbb{E}_b [e^{2\pi i b(\eta - \eta')}] d\eta d\eta' \\
&= \int_{R \times R} \phi(\eta w) \psi(\eta' w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta')} e^{-2\pi^2 \sigma_b^2 (\eta - \eta')^2} d\eta d\eta' \\
&= \int_R \phi(\eta w) \mathcal{F}[g_J](\eta) \left[\int_R \psi(\eta' w) \overline{\mathcal{F}[g_J](\eta')} e^{-2\pi^2 \sigma_b^2 (\eta - \eta')^2} d\eta' \right] d\eta \\
\text{Laplace method} &= \int_R \phi(\eta w) \mathcal{F}[g_J](\eta) \left[\psi(\eta w) \overline{\mathcal{F}[g_J](\eta)} \frac{1}{\sqrt{2\pi} \sigma_b} + O(\sigma_b^{-3}) \right] d\eta \\
&= \frac{1}{\sqrt{2\pi} \sigma_b} \int_R \phi(\eta w) \psi(\eta w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta)} d\eta + O(\sigma_b^{-3})
\end{aligned}$$

keep leading order

$$\begin{aligned}
\mathbb{E}_{w,b} \langle \phi, \psi \rangle_{K_{ab}} &\approx \mathbb{E}_w \frac{1}{\sqrt{2\pi} \sigma_b} \int_R \phi(\eta w) \psi(\eta w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta)} d\eta \\
&= \int_{R^d} \frac{1}{\sqrt{2\pi} \sigma_b} \phi(\eta w) \psi(\eta w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta)} p_w(w) dw d\eta
\end{aligned}$$

assume p_w is radially symmetric, i.e., $p_w(w) = p_w(r) = p_w(\|w\|)$

$$p_r(r) := \int_{S^{d-1}} p_w(r) r^{d-1} d\hat{\omega} = 4\pi^2 p_w(r) r^{d-1}$$

change of variables

$$\begin{cases} \zeta = \eta w \\ r = \|w\| \end{cases}$$

$$\Rightarrow \det \left(\frac{\partial(\zeta, r)}{\partial(\eta, \zeta)} \right) = - \frac{r^{d-1}}{r \|\zeta\|^{d-1}} \quad (\text{Jacobish determinant})$$

 $\therefore \mathbb{E}_{w,b} \langle \phi, \psi \rangle_{K_{ab}}$

$$\approx \int_{R^d} \frac{1}{\sqrt{2\pi} \sigma_b} \phi(\eta w) \psi(\eta w) \mathcal{F}[g_J](\eta) \cdot \overline{\mathcal{F}[g_J](\eta)} p_w(r) dw d\eta$$

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$$\begin{aligned}
&= \int_{R^d \times R^+} \frac{1}{(2\pi)^d \sigma_b} \phi(\zeta) \psi(\zeta) F[g_1](\frac{\|\zeta\|}{r}) \overline{F[g_1](\frac{\|\zeta\|}{r})} \frac{r^{dn}}{r \|\zeta\|^{dn+1}} p_w(r) d\zeta dr \\
&= \int_{R^d \times R^+} \frac{1}{(2\pi)^d \sigma_b} \phi(\zeta) \psi(\zeta) F[g_1](\frac{\|\zeta\|}{r}) \overline{F[g_1](\frac{\|\zeta\|}{r})} \frac{r^{dn}}{r \|\zeta\|^{dn+1}} \frac{p_w(r)}{4\pi r^2} d\zeta dr \\
&= \frac{1}{(2\pi)^{d+2} \sigma_b} \int_{R^d} \phi(\zeta) \int_{R^+} \left[\frac{1}{r \|\zeta\|^{dn+1}} F[g_1](\frac{\|\zeta\|}{r}) \cdot \overline{F[g_1](\frac{\|\zeta\|}{r})} \right] \psi(\zeta) \\
&\quad p_w(r) dr d\zeta
\end{aligned}$$

taking $\psi = F[u_p]$, we have

$$\begin{aligned}
L_{a,b} [F[u_p]] &= - \frac{1}{(2\pi)^{d+2} \sigma_b \|\zeta\|^{dn+1}} \mathbb{E}_{a,r} \left[\frac{1}{r} F[g_1](\frac{\|\zeta\|}{r}) \cdot \overline{F[g_1](\frac{\|\zeta\|}{r})} \right] F[u_p](\zeta) \\
&= - \frac{1}{(2\pi)^{d+2} \sigma_b \|\zeta\|^{dn+1}} \mathbb{E}_{a,r} \left[\frac{1}{r} F[g_1](\frac{\|\zeta\|}{r}) \cdot \overline{F[g_1](\frac{\|\zeta\|}{r})} \right] F[u_p](\zeta)
\end{aligned}$$

similarly, from R_w we can derive

$$\begin{aligned}
L_w [F[u_p]] &= \frac{1}{(2\pi)^{d+2} \sigma_b} \nabla \cdot \left\{ \mathbb{E}_{a,r} \left[\frac{1}{r \|\zeta\|^{dn+1}} F[g_2](\frac{\|\zeta\|}{r}) \cdot \overline{F[g_2](\frac{\|\zeta\|}{r})} \right] \nabla F[u_p](\zeta) \right\} \#
\end{aligned}$$