

Bilinear Estimate $\sigma < h$.

$$\|\rho(n, \sigma D)u\|_{L^2(\Omega)} \lesssim h \|u\|_{L^2}. \quad \|\rho(n, \sigma D)v\|_{L^2(\Omega)} \lesssim \sigma \|v\|_{L^2(\Omega)}.$$

$$\|n - \sqrt{h}\|_p \lesssim \underbrace{G(h, \sigma)}_{?} \|u\|_2 \|v\|_2.$$

Th. $\{\mu_n, \nu_n, T\}$. p haploid-like.

$$G(h, \sigma) = \begin{cases} h^{-\frac{1}{2} + \frac{1}{p}} & \sigma < p \leq \infty \\ h^{-\frac{3}{4} + \frac{3}{p}} & 3 \leq p \leq 5 \\ h^{-\frac{1}{2} - \frac{1}{p}} & 2 \leq p \leq 3 \end{cases}$$

Think of $\|h \cdot v\|_p$, not estimate $\|u(n) v(n)\|_p$ set $\{n = m\}$.

$$L \rightarrow L^\infty : \sigma^{-n} h^{-m} (h + |t-s|)^{-\frac{1}{p_\infty}} (\sigma + |t-s|)^{-p_\infty}.$$

$$L^2 \rightarrow L^2 : \sigma^{-n} h^{-m} (h + |t-s|)^{-\frac{1}{2}} (\sigma + |t-s|)^{-p_2}.$$

$|t-s| < \sigma$: high p .

$\sigma < |t-s| < h$: mid p .

$h < |t-s|$: low p .

$T_h^\alpha = M_\alpha$ (at scale h).

$$\boxed{h^{-2\alpha}} \cdot h^{1-\alpha} \cdot |T_h^\alpha| \geq c h^{-\frac{1}{2} + \frac{\alpha}{p}}.$$

High p : $\alpha=0, \beta=0$.

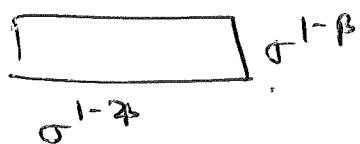


①

to on $|n| < \varepsilon^{\frac{1}{2}}$, ~~$|T_h^0| > c h^{-\frac{1}{2}}$~~ , $|T_0^0| > c \sigma^{-\frac{1}{2}}$.

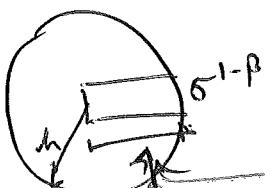
$$\|T_h^0 T_0^0\| > h^{-\frac{1}{2}} \sigma^{-\frac{1}{2} + \frac{1}{p}}.$$

T_σ^0 :



$$|T_\sigma^0| > c \sigma^{-\frac{1}{2} + \frac{p}{2}}$$

T_h^0 :



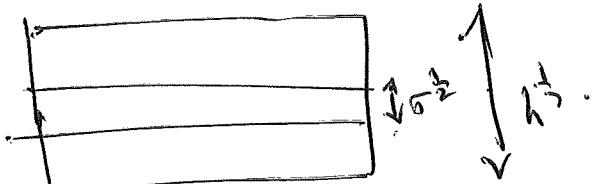
$$|T_h^0| > h^{-\frac{1}{2}} \quad |T_\sigma^{B_h}| > \sigma^{-\frac{1}{2} + \frac{p}{2}}$$

~~$\sigma^{1-2\beta_h} = h$ (defines $\sigma^{1-2\beta_h}$)~~

$$\begin{aligned} \|T_h^0 T_\sigma^{B_h}\|_p &\geq h^{-\frac{1}{2}} \sigma^{-\frac{1}{2} + \frac{p}{2}} (\sigma^{1-2\beta_h + 1-\beta_h})^{\frac{1}{p}} \\ &= h^{-\frac{3}{4} + \frac{3}{2p}} \sigma^{-\frac{1}{4} + \frac{1}{2p}} \end{aligned}$$

now P:

$$T_h^0 T_\sigma^{\frac{1}{2}}$$



$$\|T_h^0 T_\sigma^{\frac{1}{2}}\| > h^{-\frac{1}{4}} \sigma^{-\frac{1}{4}} \cdot \sigma^{\frac{1}{2p}}$$



$$|T_h^0|^{\frac{1}{2}} > h^{-\frac{1}{4}}, \quad |T_\sigma^{\frac{1}{2}}| > \sigma^{-\frac{1}{4}}.$$

Direction: $n = \sum_i x_i (n, \mu D) n, v = \sum_i x_i (n, \sigma D) v$.

$$n \cdot v = \sum_i x_i (n, \mu D) n \cdot x_i (n, \sigma D) v$$

Fadensatz: If ξ ; call it ξ_1 .

$$|\partial_{\xi_1} p(y, \xi)| > c > 0 \text{ on spt } X_i.$$

$$\text{Is } |\partial_{\xi_1} p(x, \xi)| > c > 0 \text{ on spt } X_j.$$

Case 1: Yes \rightarrow propagation in same direction.

$$p(x, \xi) = e_1(x, \xi)(\xi_1 - a_1(x, \xi')) \text{ on } X_i \text{ spt}.$$

$$p(x, \xi) = e_2(x, \xi)(\xi_2 - a_2(x, \xi')) \text{ on } X_j \text{ spt}.$$

Case 2: ~~prop~~ No: propagation in diff directions.

$$p(x, \xi) = e_1(x, \xi)(\xi_1 - a_1(x, \xi')) \text{ on spt } X_1.$$

$$p(x, \xi) = e_2(x, \xi)(\xi_2 - a_2(x, \xi')) \text{ on spt } X_2.$$

Case 3: $x_1 = t$, $w(t, n, y) := n(t, x_2) v(t, y_2)$.

$$(hD_t - a_1(t_1, x_2 hD_{x_2}) - a_2(t, y_2, hD_{y_2})) w = o_2(n).$$

Approximation at $BD \rightarrow 2D$. $\underline{\sigma \ll h}$ | Run or turn.

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$$\text{modest: } \|w_n(t) w_0(t)\|_{L_t^p L_{x_2}^p}$$

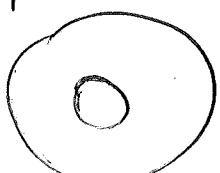
| more oscillation.
and straight
line along.

$\sigma \approx h$ is bad b/c of interactions.

$$x_1 = t, x_2 = t, t = (t_1, t_2), n = (e_1, e_2, \bar{n})$$

$$\|w(t_1) w(t_2)\|_{L_{t_1}^p L_{t_2}^p L_{x_2}^p}.$$

Example:



high P



mid P.

③

