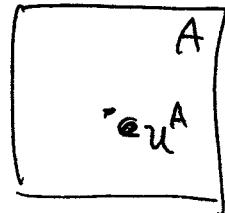
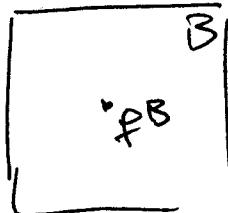


D Analytic Part of FMM

Improve the accuracy.

ε , fixed constant > 0 .



f^B : outgoing far field rep.

u^A : local field rep.

$$(2D) \quad G(x,y) = \ln|x-y| \\ = \operatorname{Re} [\ln(x-y)]$$

We will regard $G(x,y) = \ln(x-y)$ and at the end of the computation, discard the imaginary part

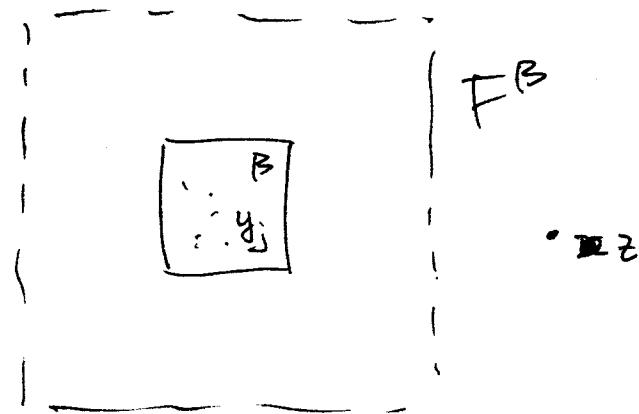
D Definition.

$$u^B(x) = \sum_{y_j \in B} G(x, y_j) \cdot f_j$$

f^B : is a representation that approximates $u^B(x) |_{x \in F^B}$

u_A : local rep. is a representation that approximates $u^{FA}(x)|_{x \in A}$

D. Far field rep.: Want to approx $u^B(x)|_{F^B}$.



$$G(x, y) = \ln(x-y) = \ln x + \ln\left(1 - \frac{y}{x}\right) \quad \left|\frac{y}{x}\right| \leq \frac{\sqrt{2}}{3}$$

$$= \ln x + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left(\frac{y}{x}\right)^k \quad (*)$$

$$u^B(z) = \sum_{y_j \in B} G(z, y_j) \cdot f_j$$

$$= \sum_{y_j \in B} \left[\ln z + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left(\frac{y_j}{z}\right)^k \right] \cdot f_j$$

$$= \left(\sum_{y_j \in B} f_j \right) \cdot \ln z + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left[\sum_{y_j \in B} \left(\frac{y_j}{z}\right)^k \cdot f_j \right] \quad (*)$$

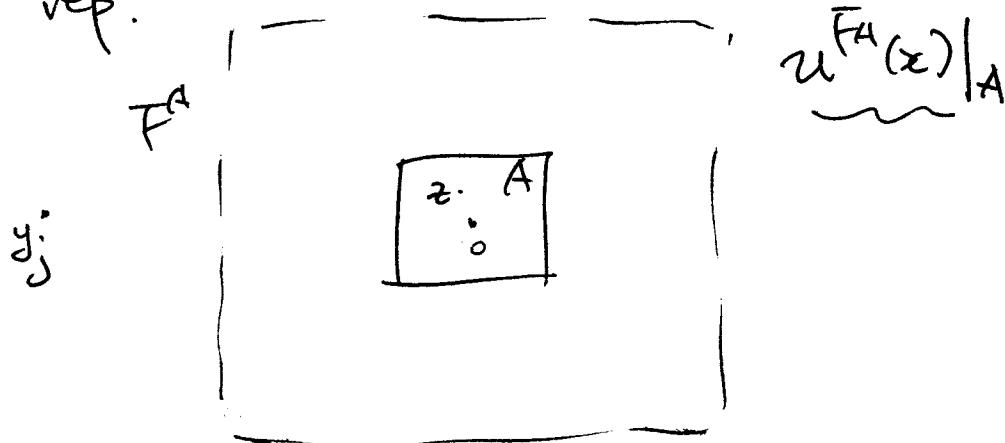
$$= \left(\sum_{y_j \in B} f_j \ln z \right) + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left[\sum_{y_j \in B} y_j^k \cdot f_j \right] \cdot \frac{1}{z^k}$$

Since $\left(\frac{y_j}{z}\right) \leq \frac{\sqrt{2}}{3}$, we can truncate (4) after $P = \underbrace{\log_3\left(\frac{1}{\varepsilon}\right)}_{\sqrt{2}}$ terms to get an ε -accuracy approximation.

$$\Rightarrow u^B(z) \approx \underbrace{\left(\sum_{y_j \in B} f_j\right) \cdot \ln z + \sum_{k=1}^P \left(-\frac{1}{k} \sum_{y_j \in B} y_j^k f_j\right)}_{a_k} \frac{1}{z^k} + O(\varepsilon).$$

$(a_0, a_1, \dots, a_p) \Rightarrow$ compact rep for $u^B(z)|_{F^B}$.
 ↳ far field rep. ("multipole representation")

D. local rep.



$$\begin{aligned}
 G(z, y) &= \ln(z-y) \\
 &= \ln(-y) + \ln\left(1 - \frac{z}{y}\right) \\
 &= \ln(-y) + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left(\frac{z}{y}\right)^k \quad (*).
 \end{aligned}$$

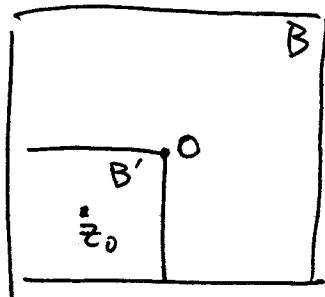
$$\begin{aligned}
 u^{FA}(z) &= \sum_{y_j \in FA} G(z, y_j) f_j \\
 &= \sum_{y_j \in FA} \left[\ln(-y_j) + \sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left(\frac{z}{y_j}\right)^k \right] \cdot f_j \\
 &= \underbrace{\sum_{y_j \in FA} (\ln(-y_j) \cdot f_j)}_{b_0} + \underbrace{\sum_{k=1}^{\infty} \left(-\frac{1}{k}\right) \left(\sum_{y_j \in FA} \frac{1}{y_j^k} \cdot f_j\right) \cdot z^k}_{b_k} \quad (*)
 \end{aligned}$$

As long as $P \geq \log_3 \left(\frac{1}{\epsilon}\right)$

$$u^{FA}(z) \approx \underbrace{\sum_{y_j \in FA} (\ln(-y_j) \cdot f_j)}_{b_0} + \underbrace{\sum_{k=1}^P \left(-\frac{1}{k}\right) \left(\sum_{y_j \in FA} \frac{1}{y_j^k} f_j\right) \cdot z^k}_{b_k}.$$

(b_0, b_1, \dots, b_P) . local ^{field} representation

D. Far 2 Far translation.



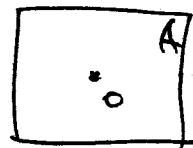
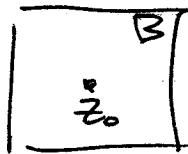
Given a_0, \dots, a_P far field rep of B' centered at z_0 , how to construct b_0, b_1, \dots, b_P : far field rep of the parent box B ?

$$u_{B'}(z) \approx a_0 \ln(z - z_0) + \sum_{k=1}^P \frac{a_k}{(z - z_0)^k} + O(\varepsilon)$$

$$\approx b_0 \ln z + \sum_{k=1}^P \frac{b_k z^k}{z^k} + O(\varepsilon)$$

$$\Rightarrow \begin{cases} b_0 = a_0 \\ b_\ell = -a_0 \frac{z_0^\ell}{\ell} + \sum_{k=1}^{\ell-1} a_k \binom{\ell-1}{k-1} \cdot z_0^{\ell-k}. \end{cases}$$

▷ far - 2 - local translation



Given $f_{B_0} \dots a_0 f$ far field rep of B
centered at z_0 , how to construct $f_{B_0} \dots b_0 f$
local field rep at A ?

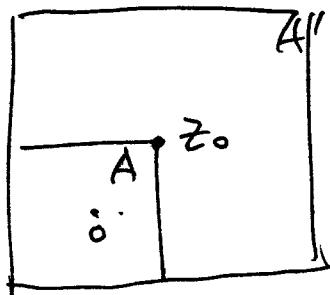
$$u^B(z) \approx a_0 \ln(z - z_0) + \sum_{k=1}^P \frac{a_k}{(z - z_0)^k}$$

$$\approx b_0 \cdot \ln z + \sum_{k=1}^P b_k z^k \quad \text{— Taylor expand}$$

$$b_0 = a_0 \cdot \ln(-z_0) + \sum_{k=1}^P a_0 (-z_0)^{-k}$$

$$b_k = \left(-\frac{a_0}{k}\right) z_0^{-k} + \sum_{l=1}^P a_k (-z_0)^{-k} \binom{k+l-1}{k-1} z_0^{-k}$$

▷ Local \Rightarrow local translation



Given local rep $\{a_0 - \alpha_p\}$ at A' , how to const. local rep $\{b_0 - b_p\}$ at A .

$$\forall z \in A, \quad u^{F^A}(z) \asymp \sum_{k=0}^P a_k \cdot (z - z_0)^k$$

\uparrow local rep at z_0 .

$$= \sum_{k=0}^P b_k(z)^k$$

\uparrow local rep at 0 .

$$\Rightarrow b_k = \underbrace{\sum_{l=0}^P (-z_0)^{k-l} \binom{k}{l} \cdot a_l}_{\text{local rep at } 0}.$$

▷ Far field rep $\Phi + I = \mathcal{O}(\log(\frac{1}{\varepsilon}))$

local field rep. $P+I = \mathcal{O}(\log(\frac{1}{\varepsilon}))$

$$\left. \begin{array}{l} F2F \\ F2L \\ L2F \end{array} \right\} \rightarrow \text{matrix of size } (P+1) \times (P+1)$$

$$= \mathcal{O}(P^2) = \mathcal{O}(\log^2(\frac{1}{\varepsilon}))$$

\Rightarrow FMM is still a linear algo with ~~constant~~ constant depending on $\log(\frac{1}{\varepsilon})$.

▷ Other alternatives for far field representations
 local field

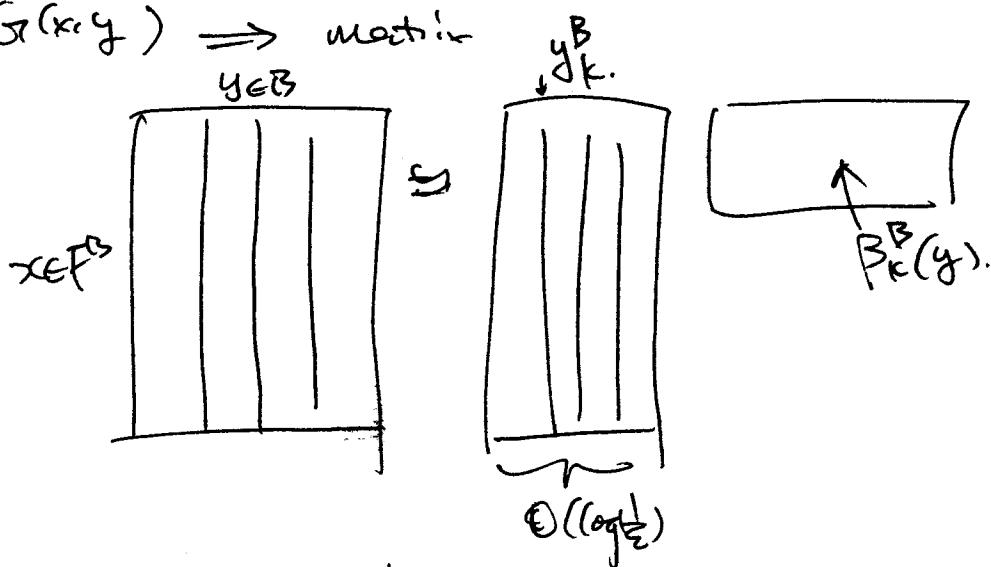
▷ far field rep.



Idea: $\left\| G(x, y) \right\|_{\substack{x \in F^B \\ y \in B}}$ is approx low rank.

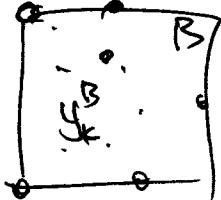
rank $\approx O(\log \frac{1}{\epsilon})$.

$G(x, y) \Rightarrow$ matrix



$$G(x, y) \approx \sum_{k=1}^{\log \frac{1}{\epsilon}} G(x, y_k^B) \cdot B_k^B(y)$$

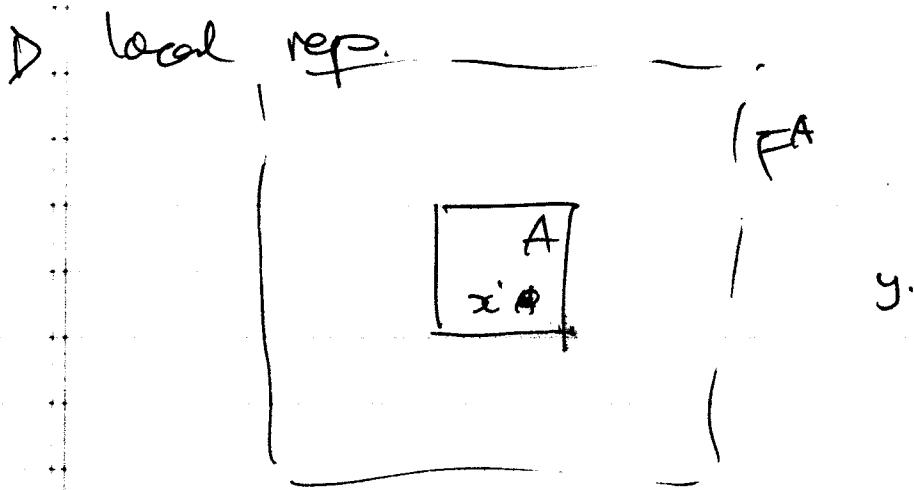
$$\begin{aligned}
 u^B(z) &= \sum_{y_j \in B} G(z, y_j) \cdot f_j \\
 &\approx \sum_{y_j \in B} \left[\sum_k G(z, y_k^B) \cdot \beta_k^B(y_j) \right] \cdot f_j \\
 &= \sum_k G(z, y_k^B) \cdot \underbrace{\left[\sum_{y_j \in B} \beta_k^B(y_j) \cdot f_j \right]}_{\text{representative columns}}
 \end{aligned}$$



We can reproduce $\hat{u}^B(z)$ by putting mass

$$\left[\sum_{y_j \in B} \beta_k^B(y_j) f_j \right] \text{ at } \{y_k^B\}.$$

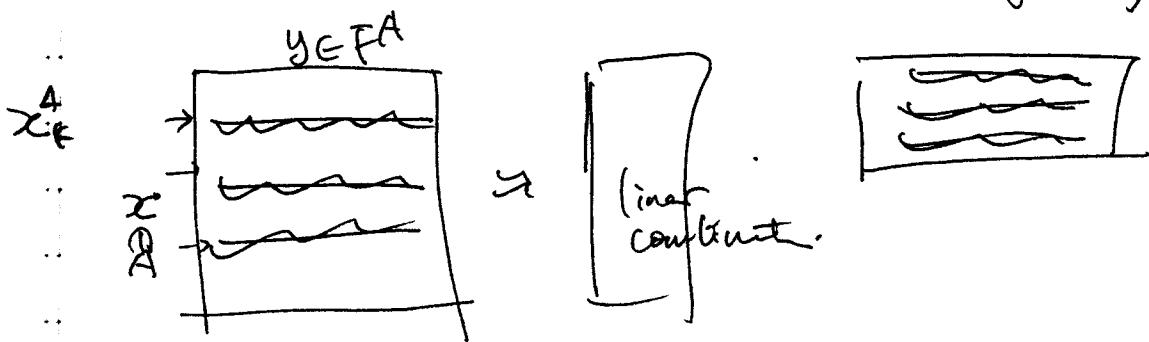
$$\left| \{y_k^B\} \right| = \log(\Sigma).$$



$$u^F_A|_{x \in A}$$

$$G(x, y) \begin{cases} x \in A \\ y \in F_A \end{cases} \quad \text{approx low rank}$$

rank $\leq O(\log(\frac{1}{\epsilon}))$



$$G(x, y) \leq \sum_k \underbrace{\alpha_k^A(x)}_{\text{linear comb. coeffs}} \cdot \underbrace{G(x_k^A, y)}_{\text{rows}}$$

$$u^F_A(z) = \sum_{y_j \in F_A} G(z, y_j) f_j$$

$$\approx \sum_{y_j \in F_A} \sum_k \alpha_k^A(z) \cdot G(x_k^A, y_j) \cdot f_j$$

$$= \sum_k \alpha_{ik}^A(x) \cdot \underbrace{\left(\sum_{y_j \in F^A} G(x_k^A, y_j) \beta_j \right)}_{\text{potenzis } x_k^A}$$

▷ Butterfly algorithm

▷ Let $x_0, x_1, \dots, x_{N-1} \in [0, N]$
 $\xi_0, \dots, \xi_{N-1} \in [0, N]$

f_0, \dots, f_{N-1} sources.

Want to compute $u_i = \sum_j G(x_i, \xi_j) \cdot f_j$



Ex: $G(x, \xi) = e^{2\pi i \frac{x \cdot \xi}{N}}$

When $x_i = i, \in [0, N]$

$\xi_j = j, \in [0, N]$

$$G(x_i, \xi_j) = e^{2\pi i \frac{x_i \cdot \xi_j}{N}} \Rightarrow \text{Fourier transform}$$

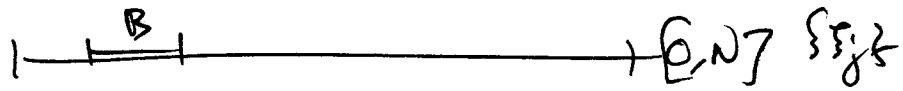
When $x_i \rightarrow \text{nonunif}$

$\xi_j = j,$

$\Rightarrow \text{nonunif Fourier transform.}$

Ex: $G(x, \xi) = \frac{1}{|x - \xi|}$
 $\Rightarrow \text{Laplace transform.}$

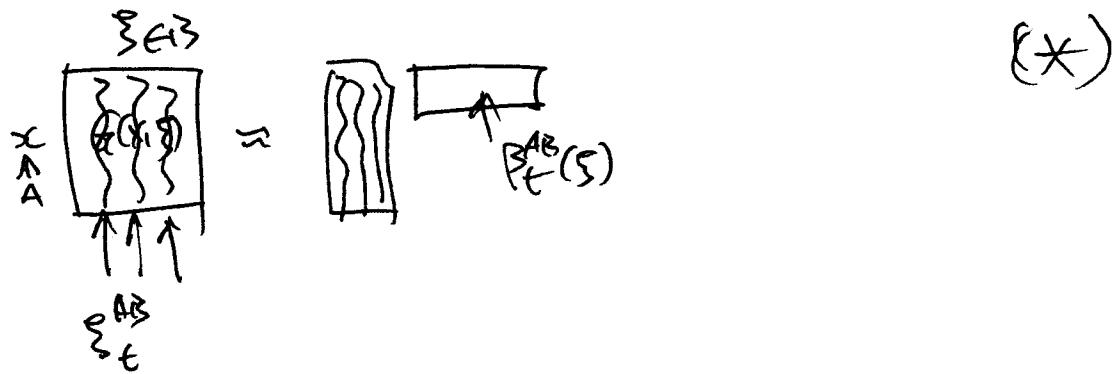
D. Assumption.



Suppose A, B two intervals of $[0, N]$

with $w^A \cdot w^B = N$, then

$$\left| G(x, \xi) - \sum_{\ell=1}^{r_\varepsilon} G(x, \xi_\ell^{AB}) \cdot \beta_\ell^{AB}(\xi) \right| \leq \varepsilon. \quad \begin{matrix} \forall x \in A \\ \forall \xi \in B \end{matrix}$$



H.W.: $G(x, \xi) = e^{2\pi i \frac{x \cdot \xi}{n}}$ F.T.
satisfies this assumption

Define. $u^B(x) = \sum_{\xi_j \in B} G(x, \xi_j) f_j$
potential generated by $\xi_j \in B$.

$$u^B(x) \Big|_{x \in A}$$

We are interested in. $u^B(x) \Big|_{x \in A}$.

Apply (*) to $\sum_{\xi_j \in B}$

$$\left| \sum_{t=1}^{r_\varepsilon} G(x, \xi_j) f_j - \sum_{t=1}^{r_\varepsilon} E(x, \xi_t^{AB}) \cdot \beta_t^{AB}(\xi_j) f_j \right| = O(\varepsilon)$$

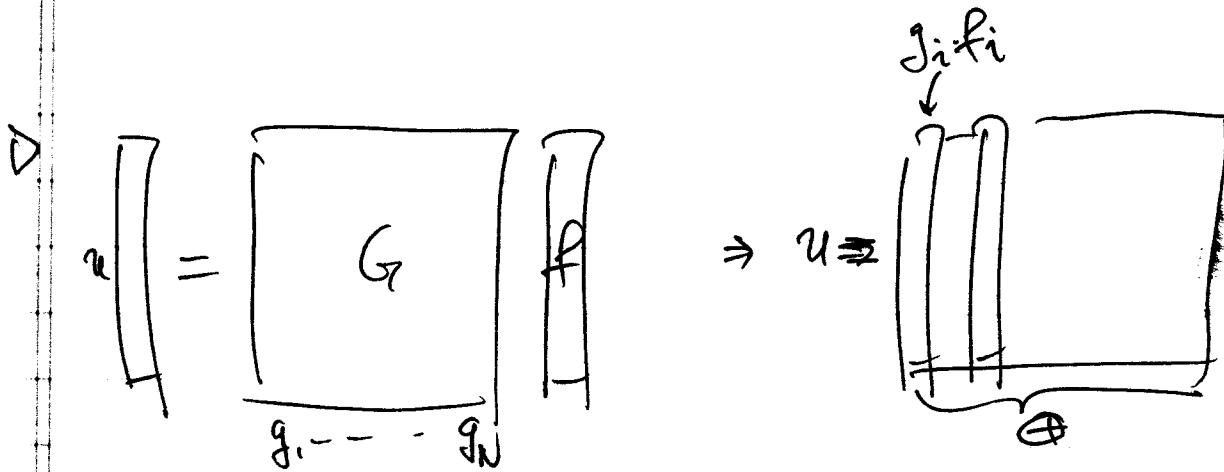
$$\left| \underbrace{\sum_{\xi_j \in B} G(x, \xi_j) f_j}_{u^B(x)} - \underbrace{\sum_j \sum_{t=1}^{r_\varepsilon} G(x, \xi_t^{AB}) \cdot \beta_t^{AB}(\xi_j) f_j}_{(u^B(x) - \sum_{t=1}^{r_\varepsilon} G(x, \xi_t^{AB}) \cdot \beta_t^{AB}(\xi_j) f_j)} \right| = O(\varepsilon)$$

$$\left| u^B(x) - \sum_{t=1}^{r_\varepsilon} G(x, \xi_t^{AB}) \cdot \left(\sum_j \beta_t^{AB}(\xi_j) f_j \right) \right| = O(\varepsilon)$$

"Effective rep".

$$\{f_t^{AB}\}$$

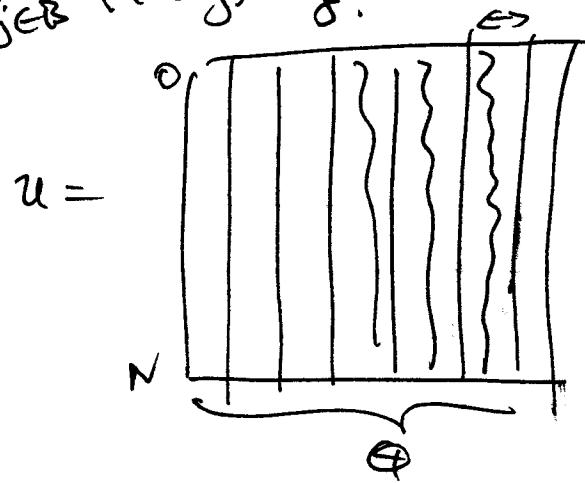
Equivalent sources.



Δ Step 0: Construct $\{f_t^{AB}\}$ for
 $A = [0, N]$
 $B = [j, j+1]$.

$$f_t^{AB} = \sum_{j \in S_j \in B} \beta_t^{AB}(g_j) \cdot f_j.$$

unit (length)



Δ Step 1: Construct $\{f_t^{AB}\}$ for
 $A = \sum_i [i, i+1] \quad i=0, 1, \dots, N-1$
 $B = 2[j, j+1] \quad j=0, 1, \dots, N_2 - 1$.

P — A 's parent
 B, B_2 — B 's children

$$W^P = N$$

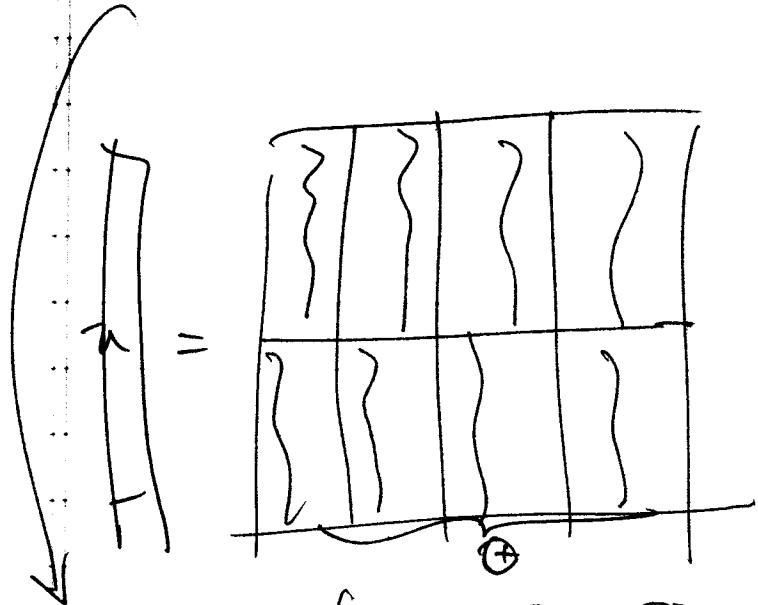
$$W^{B_1} = W^{B_2} = 1$$

$$\left| u^{B_1}(x) - \sum_t G(x, \xi_t^{PB}) \cdot f_t^{PB_1} \right| = O(\varepsilon) \quad x \in P$$

$$\left| u^{B_2}(x) - \sum_t G(x, \xi_t^{PB}) \cdot f_t^{PB_2} \right| = O(\varepsilon) \quad x \in P,$$

+

$$\left(u^B(x) - \left(\sum_t G(x, \xi_t^{PB_1}) \cdot f_t^{PB_1} + \sum_t G(x, \xi_t^{PB_2}) \cdot f_t^{PB_2} \right) \right) = O(\varepsilon). \quad x \in P$$



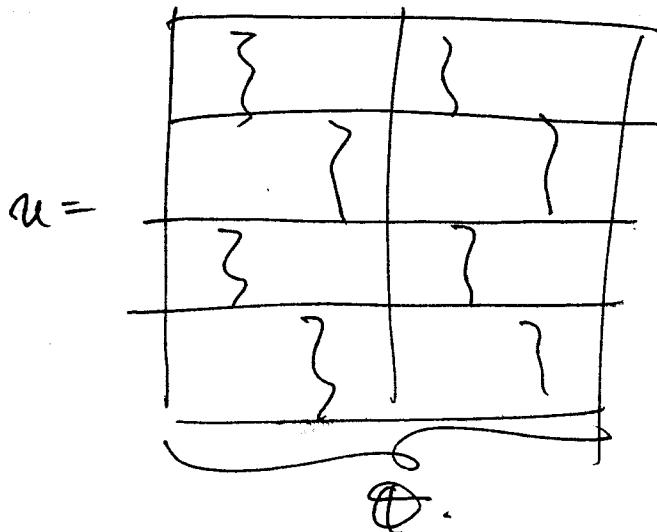
$$\left| u^B(x) - \left(\sum_t G(x, \xi_t^{PB_1}) \cdot f_t^{PB_1} + \sum_t G(x, \xi_t^{PB_2}) \cdot f_t^{PB_2} \right) \right| = O(\varepsilon) \quad x \in A$$

This says that $u^B(x) \Big|_{x \in A}$ can be approximated by $\{f_t^{PB_1}\}$ $\{f_t^{PB_2}\}$.

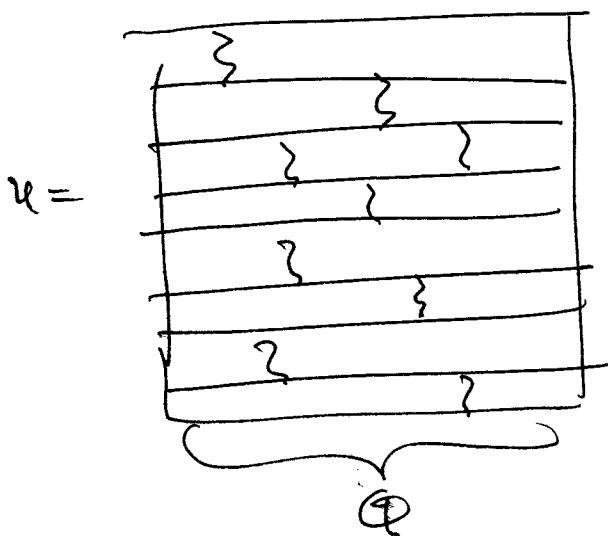
"[⊕]
Treat these eqv. sources as true sources"

" $f_t^{AB} = \sum_s \beta_t^{AB} (\xi_s^{PB_1}) \cdot f_s^{PB_1} + \sum_s \beta_t^{AB} (\xi_s^{PB_2}) \cdot f_s^{PB_2}$ "

At step 2.



At the final step



D Complexity analysis

Step 0: Construct $\{f_t^{AB}\}$ for
 $|A|=N$ $|B|=1$,
 $w^A=N$ $w^B=1$.

$$\#\{A\} = 1$$

$$\#\{B\} = N$$

Const of $\{f_t^{AB}\}$ for each pair is $O(1)$

$$\begin{aligned} \text{Total complexity} &= \#\{A\} \cdot \#\{B\} \cdot O(1) = 1 \cdot N \cdot 1 \\ &= O(N). \end{aligned}$$

Step 1: Const. $\{f_t^{AB}\}$ for
 $w^A = \frac{N}{2}$ $w^B = 2$

$$\#\{A\} = 2$$

$$\#\{B\} = \frac{N}{2}$$

Const of $\{f_t^{AB}\}$ for each pair (A, B) is $O(1)$

$$\begin{aligned} \text{Total complexity} &= \#\{A\} \cdot \#\{B\} \cdot O(1) = 2 \cdot \frac{N}{2} \cdot 1 \\ &= O(N) \end{aligned}$$

Step $\log_2 N$

$O(N)$

Each step takes $O(N)$ operations
 $\log_2 N$ steps

total complexity of the Butterfly algo
is $O(N \log N)$