

# Butterfly as Linear Algebra

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# Motivation

We want to compute

$$\hat{f}_k = \sum_{j=0}^{N-1} e^{t_k s_j} f_j$$

for  $0 \leq k < N$  and  $t_k, s_j \in \mathbf{R}$ .

# Kernel Approximation

For  $z \in \mathbf{C}$ :

$$e^z = \sum_{\alpha=0}^m \frac{z^\alpha}{\alpha!} + E_m$$

with error  $|E_m| \leq \left(\frac{|z|e}{m}\right)^m$ .

# Basic Trick

We can show that

$$ts = \tau\sigma + \tau(s - \sigma) + \sigma(t - \tau) + (t - \tau)(s - \sigma)$$

for any  $\sigma, \tau$ .

# Basic Trick in Use

We can use it to say

$$\begin{aligned} e^{ts}f &= e^{\tau\sigma} e^{\tau(s-\sigma)} e^{\sigma(t-\tau)} e^{(t-\tau)(s-\sigma)} f \\ &= e^{\tau\sigma} e^{\tau(s-\sigma)} e^{\sigma(t-\tau)} \sum_{\alpha \geq 0} \frac{1}{\alpha!} (t-\tau)^\alpha (s-\sigma)^\alpha f \\ &= \sum_{\alpha \geq 0} \left\{ \left[ \frac{1}{\alpha!} e^{\tau\sigma} e^{\tau(s-\sigma)} (s-\sigma)^\alpha \right] e^{\sigma(t-\tau)} (t-\tau)^\alpha f \right\} \\ &= \sum_{\alpha \geq 0} K(s; \alpha, \tau, \sigma) e^{\sigma(t-\tau)} (t-\tau)^\alpha f. \end{aligned}$$

# Basic Trick Emphasis

We emphasize the definition:

$$K(s; \alpha, \tau, \sigma) = \frac{1}{\alpha!} e^{\tau\sigma} e^{\tau(s-\sigma)} (s - \sigma)^\alpha.$$



# Partitioning the Inputs

For a given level  $\ell$ , assume we have partitioned the target space  $T$  into  $2^\ell$  evenly spaced intervals and have partitioned  $S$  into  $2^{L-\ell}$  where  $L = \mathcal{O}(\log_2 N)$  is some maximum depth which we determine later.

For example if  $T \subset [0, 1]$  then

$$\begin{aligned} T\{1\} &= \left[0, \frac{1}{2}\right] && \cup \left[\frac{1}{2}, 1\right] \\ T\{2\} &= \left[0, \frac{1}{4}\right] \cup \left[\frac{1}{4}, \frac{1}{2}\right] && \cup \left[\frac{1}{2}, \frac{3}{4}\right] \cup \left[\frac{3}{4}, 1\right]. \end{aligned}$$

# Partitioning the Inputs

In general

$$T\{\ell\} = \bigcup_{i=0}^{2^{\ell}-1} T(i; \ell)$$

and

$$S\{L - \ell\} = \bigcup_{j=0}^{2^{L-\ell}-1} S(j; L - \ell).$$

Define the centers of these boxes (intervals here, but feel free to get more generic) as

$$\tau(i; \ell) \text{ and } \sigma(m; L - \ell).$$

# Return to Motivation

Recall

$$\widehat{f}_k = \sum_{j=0}^{N-1} e^{t_k s_j} f_j$$

and notice that for a fixed  $k$  and fixed  $\ell$ , there is a unique  $i$  such that  $t_k \in T(i; \ell)$ , then invoking  $K(s; \alpha, \tau(i; \ell), \sigma)$  as  $\sigma$  varies over

$$\sigma(0; L - \ell), \dots, \sigma(2^{L-\ell} - 1; L - \ell)$$

will turn our sum into  $2^{L-\ell}$  sums (each with another sum over  $\alpha$ ).

## Return to Motivation

Thus, letting  $\tau = \tau(i; \ell)$ :

$$\begin{aligned}\widehat{f}_k &= \sum_{j=0}^{N-1} e^{t_k s_j} f_j \\ &= \sum_m \sum_{s \in S(m; L-\ell)} e^{t_k s} f(s) \\ &= \sum_m \sum_{s \in S(m)} \sum_{\alpha \geq 0} K(s; \alpha, \tau, \sigma(m)) e^{\sigma(m)(t_k - \tau)} (t_k - \tau)^\alpha f(s).\end{aligned}$$

Above we write  $\sigma(m)$  instead of  $\sigma(m; L - \ell)$  since the refinement level is clear from context.

**NOTE:** We write  $f(s)$  to denote  $f_j$  in the case that  $s = s_j$ .

# Rearrangement

Rearranging

$$\begin{aligned}\hat{f}_k &= \sum_m \sum_{s \in S(m)} \sum_{\alpha \geq 0} K(s; \alpha, \tau, \sigma(m)) e^{\sigma(m)(t_k - \tau)} (t_k - \tau)^\alpha f(s) \\ &= \sum_m \sum_{\alpha \geq 0} e^{\sigma(m)(t_k - \tau)} (t_k - \tau)^\alpha \sum_{s \in S(m)} K(s; \alpha, \tau, \sigma(m)) f(s) \\ &= \sum_m \sum_{\alpha \geq 0} C(\alpha, \ell, i, m) e^{\sigma(m)(t_k - \tau)} (t_k - \tau)^\alpha.\end{aligned}$$

# Rearrangement

We emphasize the definition:

$$C(\alpha, \ell, i, m) = \sum_{s \in S(m)} K(s; \alpha, \tau(i), \sigma(m)) f(s)$$

or more explicitly

$$C(\alpha, \ell, i, m) = \frac{1}{\alpha!} \sum_{s \in S(m)} e^{\tau(i)\sigma(m)} e^{\tau(i)(s-\sigma(m))} (s - \sigma(m))^\alpha f(s).$$

# Approximation

For any level  $\ell$  with  $t_k \in T(i; \ell)$  we can approximate the exact value

$$\hat{f}_k = \sum_m \sum_{\alpha \geq 0} C(\alpha, \ell, i, m) e^{\sigma(m)(t_k - \tau(i))} (t_k - \tau(i))^\alpha.$$

by

$$\hat{f}_k \approx \sum_m \sum_{\alpha=0}^{M-1} C(\alpha, \ell, i, m) e^{\sigma(m)(t_k - \tau(i))} (t_k - \tau(i))^\alpha$$

for some max number of terms  $M$ .

# Approximation

Notice that  $C(\alpha, \ell, i, m)$  does not depend on  $k$ , so we only need to compute this for  $M$  choices of  $\alpha$ ,  $2^\ell$  choices of  $i$  and  $2^{L-\ell}$  choices of  $m$ . In total

$$M \cdot 2^L = \mathcal{O}(M \cdot N) = \mathcal{O}(N)$$

independent of  $\ell$ .



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- ▶ Since at level  $L$ , we know  $|S(m; L)| = \mathcal{O}(1)$  hence computing each coefficient is  $\mathcal{O}(1)$ .
- ▶ However, for each of these boxes, we need to compute  $\exp\{\sigma(m; L)(t_k - \tau(0; 0))\}$  which is  $\mathcal{O}(N^2)$  since we have  $2^L = \mathcal{O}(N)$  choices for  $m$  and  $N = \mathcal{O}(N)$  choices for  $t_k$ .

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- ▶ Since at level 0, we know  $|S(0; 0)| = N = \mathcal{O}(N)$  hence computing each coefficient (for fixed  $i$ ) is  $\mathcal{O}(N)$ .
- ▶ Since we have  $\mathcal{O}(N)$  choices for  $i$ , there are  $\mathcal{O}(N^2)$  coefficients to compute.



## Here's the Catch

The the  $\ell = L$  approach sinks us immediately by asking us to compute  $\mathcal{O}(N^2)$  coefficients, we have a **positive takeaway**.

Since there is only one value  $\sigma = \sigma(0; 0)$ , we can compute each value as

$$\hat{f}_k \approx \sum_{\alpha=0}^{M-1} C(\alpha, L, i, m=0) e^{\sigma(t_k - \tau(i; L))} (t_k - \tau(i; L))^{\alpha}.$$

This is happily  $\mathcal{O}(M) = \mathcal{O}(1)$  to compute.

## Now What

- ▶ We've seen that no matter what  $\ell$  is, we have  $M \cdot 2^L = \mathcal{O}(N)$  coefficients  $C(\alpha, \ell, i, m)$  to compute.

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- ▶ When  $\ell = L$  we can compute all the values  $\hat{f}_k$  in  $\mathcal{O}(1)$  provided we already have the  $C(\alpha, \ell, i, m)$ .

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- ▶ When  $\ell = L$  we can compute all the values  $\hat{f}_k$  in  $\mathcal{O}(1)$  provided we already have the  $C(\alpha, \ell, i, m)$ .
- ▶ This means that the computation of  $\{\hat{f}_k\}_k$  is  $\mathcal{O}(N)$ .

## Now What

- To summarize, we know we can compute the  $\mathcal{O}(N)$  values  $C(\alpha, 0, i = 0, m)$  in  $\mathcal{O}(N)$  and we know we can use the  $\mathcal{O}(N)$  values  $C(\alpha, L, i, m = 0)$  to compute  $\{\hat{f}_k\}_k$  in  $\mathcal{O}(N)$ .

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- ▶ (Let's assume that) Butterfly gives a process to convert

$$\{C(\alpha, 0, i, m)\} \rightarrow \{C(\alpha, 1, i, m)\} \rightarrow \cdots \rightarrow \{C(\alpha, L, i, m)\}$$

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$$\mathcal{O}(N) + \mathcal{O}(N \log N) + \mathcal{O}(N) = \mathcal{O}(N \log N).$$

- ▶ **In other words, the process of converting the coefficients will dominate.**

# High Level Approach

In order to facilitate the process

$$\{C(\alpha, \ell, i, m)\}_{\alpha, i, m} \rightarrow \{C(\alpha, \ell + 1, i, m)\}_{\alpha, i, m}$$

we split into two parts:

- ▶ Refine  $T$  and compute interaction coefficients for  $T \{\ell + 1\}$  and  $S \{L - \ell\}$ .
- ▶ This first step will use the existing interaction coefficients for  $T \{\ell\}$  and  $S \{L - \ell\}$ . These are exactly  $\{C(\alpha, \ell, i, m)\}_{\alpha, i, m}$ .
- ▶ Coarsen  $S$  and compute interaction coefficients for  $T \{\ell + 1\}$  and  $S \{L - \ell - 1\}$  using the intermediate coefficients computed in the first step. This will produce  $\{C(\alpha, \ell + 1, i, m)\}_{\alpha, i, m}$ .

# Memories...

Before we get started, recall for  $\tau = \tau(i; \ell)$  and  $\sigma = \sigma(m; L - \ell)$ :

$$C(\alpha, \ell, i, m) = \frac{1}{\alpha!} \sum_{s \in S(m; L - \ell)} e^{\tau\sigma} e^{\tau(s - \sigma)} (s - \sigma)^\alpha f(s).$$

## Refining $T$

When refining  $T(i; \ell)$ , we get two new intervals. Assuming they are ordered by their indices, we have

$$T(i; \ell) = T(2i; \ell + 1) \cup T(2i + 1; \ell + 1).$$

For shorthand, we'll use the notation

$$\tau(i) = \tau(i; \ell) \text{ and } \tau^+(i') = \tau(i'; \ell + 1)$$

when the value of  $\ell$  is clear from context.

In addition, we define the intermediate interaction coefficients:

$$C^+(\alpha, \ell, i', m) = \frac{1}{\alpha!} \sum_{s \in S(m)} e^{\tau^+(i')\sigma(m)} e^{\tau^+(i')(s - \sigma(m))} (s - \sigma(m))^\alpha f(s).$$

# Refining $T$

One can show

$$C^+(\alpha, \ell, i', m) \approx e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{M-1} (\tau^+ - \tau)^\beta \binom{\alpha + \beta}{\beta} C(\alpha + \beta, \ell, i, m)$$

by truncating the sum

$$e^{(\tau^+ - \tau)(s - \sigma)} = \sum_{\beta \geq 0} \frac{1}{\beta!} (\tau^+ - \tau)^\beta (s - \sigma)^\beta$$

and rewriting

$$e^{\tau^+ \sigma} e^{\tau^+(s - \sigma)} = e^{(\tau^+ - \tau)\sigma} e^{\tau\sigma} e^{\tau(s - \sigma)} e^{(\tau^+ - \tau)(s - \sigma)}.$$

# Refining $T$

In reality,  $C(\alpha + \beta, \ell, i, m)$  isn't defined for all values of  $\beta$  since we only consider inputs to  $C(\cdot, \ell, i, m)$  from  $[0, M)$ .

Thus

$$C^+(\alpha, \ell, i', m) \approx e^{(\tau^+ - \tau)\sigma} \sum_{\gamma=\alpha}^{M-1} (\tau^+ - \tau)^{\gamma-\alpha} \binom{\gamma}{\alpha} C(\gamma, \ell, i, m)$$

is the actual approximation we use.

# Coarsening $S$

Now we have written the  $C^+(\cdot, \ell, i', m)$  values in terms of the  $C(\cdot, \ell, i, m)$  values.

From here, we want to use the  $C^+(\cdot, \ell, i', m)$  values to express the  $C(\cdot, \ell + 1, i', m')$  values in the same fashion, but here by coarsening  $S$ :

$$S(m'; L - \ell - 1) = S(2m'; L - \ell) \cup S(2m' + 1; L - \ell).$$

The definition  $m'$  here requires that  $m \in \{2m', 2m' + 1\}$ . In either case  $m' = \lfloor \frac{m}{2} \rfloor$ .

Similarly, for shorthand, we'll use the notation

$$\sigma(m) = \sigma(m; L - \ell) \text{ and } \sigma^-(m') = \sigma(m'; L - \ell - 1).$$

# Coarsening $S$

When computing

$$C(\alpha, \ell + 1, i', m') = \frac{1}{\alpha!} \sum_{s \in S(m'; L - \ell - 1)} e^{\tau^+ \sigma^-} e^{\tau^+(s - \sigma^-)} (s - \sigma^-)^\alpha f(s)$$

we need to split our sum into

$$\sum_{s \in S(m'; L - \ell - 1)} = \sum_{s \in S(2m'; L - \ell)} + \sum_{s \in S(2m' + 1; L - \ell)} .$$



# Coarsening $S$

After splitting the into two sums, we can show

$$\begin{aligned} C(\alpha, \ell + 1, i', m') &= \sum_{\beta=0}^{\alpha} \frac{(\sigma(2m') - \sigma^-)^{\alpha-\beta}}{(\alpha - \beta)!} C^+(\beta, \ell, i', 2m') \\ &\quad + \sum_{\beta=0}^{\alpha} \frac{(\sigma(2m' + 1) - \sigma^-)^{\alpha-\beta}}{(\alpha - \beta)!} C^+(\beta, \ell, i', 2m' + 1). \end{aligned}$$

by using binomial expansion

$$(s - \sigma^-)^{\alpha} = \sum_{\beta=0}^{\alpha} \binom{\alpha}{\beta} (s - \sigma)^{\beta} (\sigma - \sigma^-)^{\alpha-\beta}.$$

# Combining Transformations

For each fixed  $\ell, i, m$ , the coefficient vector  $C(\cdot, \ell, i, m) \in \mathbf{R}^M$  and the system we've described by refining and coarsening gives a block linear transformation:

$$A: \begin{bmatrix} C(\cdot, \ell, i, 2m') \\ C(\cdot, \ell, i, 2m' + 1) \end{bmatrix} \mapsto \begin{bmatrix} C(\cdot, \ell + 1, 2i, m') \\ C(\cdot, \ell + 1, 2i + 1, m') \end{bmatrix}.$$

In the case of higher dimensions or for inputs in  $\mathbf{C}$  the transformation goes from operating on vectors of size  $2M$  to size  $2^d M$  where  $d$  is the dimension (over  $\mathbf{R}$ ) of the space containing the  $t_k, s_j$ .

# Determine Matrix $A$

To determine each element  $A_{pq}$  we need to find the coefficient of the  $q^{\text{th}}$  element of the input vector in the  $p^{\text{th}}$  element of the output vector.

This splits into four distinct cases, depending on whether each of  $p, q$  correspond to the top or bottom half. Thus we can split  $A$  into a block  $2 \times 2$  matrix

$$A = \begin{bmatrix} E & F \\ G & H \end{bmatrix}.$$

# Determine Matrix $A$

For  $0 \leq p, q < M$ .

$$C(p, \ell + 1, 2i, m') = \cdots + E_{pq} C(q, \ell, i, 2m') + \cdots$$

$$C(p, \ell + 1, 2i, m') = \cdots + F_{pq} C(q, \ell, i, 2m' + 1) + \cdots$$

$$C(p, \ell + 1, 2i + 1, m') = \cdots + G_{pq} C(q, \ell, i, 2m') + \cdots$$

$$C(p, \ell + 1, 2i + 1, m') = \cdots + H_{pq} C(q, \ell, i, 2m' + 1) + \cdots$$

# Determine General Block Submatrix $X$

With  $i' \in \{2i, 2i + 1\}$  and  $m \in \{2m', 2m' + 1\}$

$$C(p, \ell + 1, i', m') = \cdots + X_{pq} C(q, \ell, i, m) + \cdots$$

we use  $\sigma = \sigma(m)$ ,  $\sigma^- = \sigma(m')$ ,

$$C(p, \ell + 1, i', m') = \cdots + \sum_{\beta=0}^p \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} C^+ (\beta, \ell, i', m) + \cdots$$

and recall with  $\tau = \tau(i)$ ,  $\tau^+ = \tau(i')$ ,

$$C^+ (\beta, \ell, i', m) = e^{(\tau^+ - \tau)\sigma} \sum_{\gamma=\beta}^{M-1} (\tau^+ - \tau)^{\gamma-\beta} \binom{\gamma}{\beta} C(\gamma, \ell, i, m).$$

# Determine General Block Submatrix $X$

Since

$$\sum_{\beta=0}^p \sum_{\gamma=\beta}^{M-1} = \sum_{\gamma=0}^{M-1} \sum_{\beta=0}^{\min(p,\gamma)},$$

the coefficient of  $C(\gamma, \dots)$  when  $\gamma = q$  is

$$X_{pq} = e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{\min(p,q)} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}.$$

## Explicit Blocks: $m = 2m'$

- ▶ In both cases  $\sigma = \sigma(2m')$ ,  $\sigma^- = \sigma(m')$  and  $\tau = \tau(i)$ .
- ▶ When  $\tau^+ = \tau(2i)$ ,

$$E_{pq} = e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{\min(p,q)} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}.$$

- ▶ When  $\tau^+ = \tau(2i+1)$ ,

$$G_{pq} = e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{\min(p,q)} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}.$$

## Explicit Blocks: $m = 2m' + 1$

- ▶ In both cases  $\sigma = \sigma(2m' + 1)$ ,  $\sigma^- = \sigma(m')$  and  $\tau = \tau(i)$ .
- ▶ When  $\tau^+ = \tau(2i)$ ,

$$F_{pq} = e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{\min(p,q)} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}.$$

- ▶ When  $\tau^+ = \tau(2i + 1)$ ,

$$H_{pq} = e^{(\tau^+ - \tau)\sigma} \sum_{\beta=0}^{\min(p,q)} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}.$$



# Potential Optimization

We encounter

$$\frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} (\tau^+ - \tau)^{q-\beta} \binom{q}{\beta}$$

for various values of  $\sigma, \sigma^-, \tau, \tau^+$  but for a fixed set of choices of  $p, q$  (throughout the life of the code).

We can write this instead as

$$\frac{q!}{\beta!} \frac{(\sigma - \sigma^-)^{p-\beta}}{(p-\beta)!} \frac{(\tau^+ - \tau)^{q-\beta}}{(q-\beta)!}.$$

# Operation Count

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- ▶ Each  $p, q$  entry in each of the  $2 \times 2$  block submatrices can be constructed in  $\min(p, q)$  hence can be computed in  $\mathcal{O}(M)$ .
- ▶ Each  $A$  is applied to 2 of the  $2^L$  coefficient sets and this application requires  $\mathcal{O}((2M)^3)$  operations.
- ▶ Thus the total work to convert two of the coefficient sets is  $(2M)^2 \mathcal{O}(M) + \mathcal{O}((2M)^3) = \mathcal{O}(M^3)$ .

# Operation Count

- ▶ Since we have  $2^L/2 = 2^{L-1} = \mathcal{O}(N/2)$  such pairs of coefficient sets, the total work is  $\mathcal{O}(M^3 \cdot N)$ .

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- ▶ Since we only have  $L$  steps

$$\ell = 0 \rightarrow \ell = 1 \rightarrow \dots \rightarrow \ell = L - 1 \rightarrow \ell = L$$

the total work is  $\mathcal{O}(L \cdot N)$ . As we already mentioned  $L = \mathcal{O}(\log_2 N)$  hence the total work is  $\boxed{\mathcal{O}(N \log N)}$ .



# Error Analysis

We have truncated the Taylor series for the kernel in two primary places:

$$\hat{f}_k \approx \sum_m \sum_{\alpha=0}^{M-1} C(\alpha, \ell, i, m) e^{\sigma(m)(t_k - \tau(i))} (t_k - \tau(i))^\alpha$$

and

$$C^+(\alpha, \ell, i', m) \approx e^{(\tau^+ - \tau)\sigma} \sum_{\gamma=\alpha}^{M-1} (\tau^+ - \tau)^{\gamma-\alpha} \binom{\gamma}{\alpha} C(\gamma, \ell, i, m).$$

We need to understand how these errors propagate through our solution.