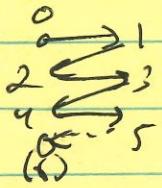


EARLY JUNE,  
2019

Toy Field  $\rightarrow$  1D (ALCs of CPT) of numbered CHAIN

In 1D, 6 ( $N$ ) numbered chain  $\rightarrow$  RIM (L, MRC).



$$P_1(x) = \frac{1}{\sqrt{2\pi \cdot 26^2}} e^{-(x-0)^2/26^2}$$

$$P_2(x) = \frac{1}{\sqrt{2\pi \cdot 26^2}} e^{-\frac{x^2}{2 \cdot 26^2}}$$

$$P_3(x) = \frac{1}{\sqrt{2\pi \cdot 36^2}} e^{-\frac{x^2}{2 \cdot (36^2)}}$$

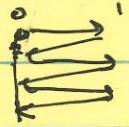
=

$$P_5(x) = \frac{1}{\sqrt{2\pi \cdot (56^2)}} e^{-\frac{(x-0)^2}{2 \cdot (56^2)}}$$

$$P_6(x) = \frac{1}{\sqrt{2\pi \cdot (66^2)}} e^{-\frac{x^2}{2 \cdot (66^2)}}$$

$$\int_{-\infty}^{\infty} P_6(x) dx = P_6(0) = \frac{1}{\sqrt{2\pi \cdot (66^2)}}$$

Now calculate properties after Ring closure



BASED ON  
 $P(x) \propto P_1(x) P_2(x)$   
 FINALLY even PATH PATH

To CREATE SAMPLES,

Put  $x_0 \rightarrow$  ORIGIN

Select  $x_1$  from  $P(x)$ .

Then  $x_2$  select from:

$$-\frac{[(x - (x_1 - 0)]^2}{26^2}$$

$$-\frac{x^2}{2 \cdot (46^2)}$$

$$\left\{ \begin{array}{l} P(x) = \delta(x) \Leftarrow \text{TRANSLATE TO ORIGIN} \\ P_1(x) = \left[ \frac{1}{\sqrt{2\pi \cdot 26^2}} e^{-(x-0)^2/26^2} \right] \left[ \frac{1}{\sqrt{2\pi \cdot (56^2)}} e^{-\frac{(x-0)^2}{2 \cdot (56^2)}} \right] \frac{1}{N} \\ = \frac{1}{\sqrt{2\pi \cdot 56^2}} e^{-\frac{(x-0)^2}{2 \cdot (56^2)}} \end{array} \right.$$

$\approx$  Just based on normalization requirement.

$$\begin{aligned} P_2(x) &= N \exp \left[ -\frac{x^2}{26^2} \left[ \frac{1}{2} + \frac{1}{4} \right] \right] \\ &= \frac{1}{\sqrt{2\pi \cdot (46^2)}} \exp \left[ -\frac{x^2}{2 \cdot (46^2)} \right] \end{aligned}$$

$$\begin{aligned} P_3(x) &= N \exp \left[ -\frac{(x-0)^2}{26^2} \left[ \frac{1}{2} + \frac{1}{4} \right] \right] \\ &= \frac{1}{\sqrt{2\pi \cdot 26^2}} \exp \left[ -\frac{(x-0)^2}{26^2} \cdot \frac{3}{2} \right] \end{aligned}$$

$$P_4(x) = P_2(x)$$

$$P_5(x) = P_1(x)$$

(2)

CRT-checks to run on  $P_1(x), \dots, P_5(x)$

Code up BACKTRACK SAMPLE,

- calculate  $P_{f(1)}, P_{f(2)} \dots P_{f(5)}$

Compare to analytical calculation.

Compare to numerical calculations

- based on MARKOV CHAIN MonteCarlo:

$\sum$  { lecture  $x_1, \dots, x_5$ ,  
accept/reject  
based on  
metropolis-Hastings }

- based on  $P_f(x), P_h(y)$

from simple sample - then  
take products and  
check.

[This is analogous to Ramanujan's  
calculation of Ramanujan's  
tau function.]

BACKTRACK  
SIMPLIFY + numeric.

How about also calculate & visualize  $P(x_1, x_5)$  ?



$x_5$        $x_1$       { there should be  
a weak correlation -  
WHAT IS IT? }

$$P(x_5|x_1) \propto e^{-\frac{(x_5 - x_1)^2}{2}} = e^{-\frac{(x_5 - x_1)^2}{2} \cdot \frac{1}{2}}$$

$$P(x_1, x_5) = P(x_1) P(x_5|x_1) \propto \left[ e^{-\frac{(x_1 - 0)^2}{2}} e^{-\frac{(x_5 - x_1)^2}{2} \cdot \frac{1}{2}} \right]$$

$$\stackrel{?}{\rightarrow} e^{-\frac{(x_1 - 0)^2}{2} \cdot \frac{1}{2} \cdot \frac{1}{2}}$$

$$\propto \exp\left[-\frac{1}{2}\right] \left[ \frac{(x_1 - 0)^2}{2} + \frac{(x_5 - x_1)^2}{2} + \frac{(x_1 - 0)^2}{2} \right]$$

$$\propto \exp\left(-\frac{1}{2}\right) \left[ \frac{(x_1 - 0)^2}{2} + \frac{1}{4}((x_5 - x_1)^2 + (x_1 - 0)^2) \right] \quad \text{where } x_1 = x_1 - 0, \quad \Delta x_5 = x_5 - 0,$$

$$\text{Check: } P(x) = \int p(x_1, x_5) dx_5 = \int \exp\left[-\frac{1}{2}\right] \left[ \frac{5}{4} \Delta x_5^2 - \frac{1}{2} (x_1 - 0)^2 + \frac{5}{4} x_1^2 \right] dx_5$$

$$= \int \exp\left[-\frac{1}{2}\right] \left[ \frac{5}{4} (\Delta x_5 - \frac{1}{5} \Delta x_1)^2 - \frac{1}{20} \Delta x_1^2 + \frac{5}{4} x_1^2 \right] dx_5$$

$$\propto \exp\left(-\frac{1}{2}\right) \left[ \frac{25}{20} \Delta x_1^2 \right]$$

OK

(3)

Calculate  $P$  for  $x_1 - x_0, x_2 - x_0$

$$P(x) \sim \exp\left[-\frac{1}{2\sigma^2}\right] \left[ \Delta x_5^2 + \frac{1}{4}(\Delta x_5 - \Delta x_1)^2 + \Delta t_1^2 \right]$$

$$\langle \Delta x_5^2 \rangle = \sqrt{\frac{4}{5}\sigma^2} \quad \leftarrow P(x, \rho_{\text{fit}}) \sim \exp\left(-\frac{1}{2\sigma^2}\right) \left[ \frac{1}{4}(\Delta x_5 - \frac{1}{5}\Delta x_1)^2 + \frac{5}{8}\sigma^2 \right]$$

$$\begin{aligned} \langle \Delta x_5^2 \rangle &= \int \Delta x_5^2 \exp\left(-\frac{1}{2\sigma^2}\right) (\Delta x_5^2 + \frac{1}{4}(\Delta x_5 - \Delta x_1)^2 + \Delta t_1^2) d\Delta x_5 / \int \\ &= \int \left[ \Delta x_5^2 + \frac{1}{4}(\Delta x_5 - \Delta x_1)^2 + \Delta t_1^2 \right] \exp\left(-\frac{1}{2\sigma^2}\right) \left( \Delta x_5^2 + \frac{1}{4}(\Delta x_5 - \Delta x_1)^2 + \frac{5}{8}\sigma^2 \right) d\Delta x_5 \\ &\stackrel{?}{=} \int \left[ \frac{4}{5}\sigma^2 + \frac{1}{25}\Delta x_1^2 \right] \exp\left(-\frac{1}{2\sigma^2}\right) d\Delta x_5 \quad A = \Delta x_5 - \frac{1}{5}\Delta x_1 \\ &= \left[ \frac{4}{5}\sigma^2 + \frac{1}{25}\Delta x_1^2 \right] \quad \Delta x_5 = A + \frac{1}{5}\Delta x_1 \\ &= \left[ \frac{4}{5}\sigma^2 + \frac{1}{36}\Delta x_1^2 \right] \quad \Delta x_5 = A + \frac{1}{5}\Delta x_1 \\ &= \frac{5}{6}\sigma^2. \quad (\text{Should be } \frac{5}{6}\sigma^2) \end{aligned}$$

REVISIT  
ON  
NEXT  
PAGE →  $\sigma_L$   
found  
 $\frac{5}{6}\sigma^2$

Alternatively, do integral over  $x_1$  first.

$$\langle \Delta x_5^2 \rangle = \int \Delta x_5^2 \exp\left[-\frac{1}{2}\left(\frac{1}{4}\Delta x_5^2 + \frac{1}{5}\Delta x_1^2\right)\right] d\Delta x_5$$

$$= \frac{5}{6}\sigma^2$$

$$\langle \Delta x_1^2 \rangle = \frac{5}{6}\sigma^2$$

$$\langle \Delta x_5^2 \rangle = \frac{\int dx_1 dx_2 \Delta x_5^2 \exp\left[-\frac{1}{20}\left[\Delta x_1^2 + \frac{1}{4}(\Delta x_2 - \Delta x_5)^2 + \Delta x_5^2\right]\right]}{\int dx_1 dx_2 \exp\left[-\frac{1}{20}\left[\Delta x_1^2 + \frac{1}{4}(\Delta x_2 - \Delta x_5)^2 + \Delta x_5^2\right]\right]}$$

complete the square

$\langle \Delta x_5^2 \rangle$  deviating  $\overset{N}{=}$

$$\int dx_1 dx_2 \exp\left[-\frac{1}{20}\left[\frac{5}{4}\left(\Delta x_1 - \frac{1}{5}\Delta x_5\right)^2 + \frac{6}{5}\Delta x_5^2\right]\right]$$

$\frac{1}{20} + \frac{6}{5} = \frac{26}{20}$

$$= \int_{-\infty}^{\infty} dx_5 \exp\left(-\frac{1}{2}\left[\frac{6}{5}\Delta x_5^2\right]\right) \quad \begin{matrix} \text{suggests} \\ r^2 = \frac{1}{5} \end{matrix}$$

$$= \sqrt{\frac{4}{5}} \sqrt{2\pi} \sigma$$

$$= \sqrt{\frac{4}{5}} \sqrt{\frac{2\pi}{5}} \sigma^2.$$

Now compute  $\langle \Delta x_1^2 \rangle$ ,  $\langle \Delta x_2^2 \rangle$ , similarly complete the square.

$$\langle \Delta x_1^2 \rangle = \int dx_1 dx_2 \Delta x_1^2 \exp\left(-\frac{1}{20}\right) \left[ \frac{5}{4}\left(\Delta x_1 - \frac{1}{5}\Delta x_5\right)^2 + \frac{6}{5}\Delta x_5^2 \right]$$

$$= \sqrt{\frac{4}{5}} \sqrt{2\pi} \int d(\Delta x_1) \Delta x_1^2 \exp\left(-\frac{1}{20}\right) \frac{6}{5}\Delta x_5^2 \quad \begin{matrix} a = \Delta x_1 - \frac{1}{5}\Delta x_5 \\ da = d\Delta x_1 \end{matrix}$$

$$= \sqrt{\frac{4}{5}} \sqrt{\frac{6}{5}} \sqrt{2\pi} \sigma^2 \times \left(\frac{6}{5}\sigma^2\right)$$

$$\langle \Delta x_1^2 \rangle = \frac{6}{5}\sigma^2$$

Now try other integers first.

$$\langle \Delta x_5^2 \rangle = \frac{1}{20} \int dx_1 dx_2 \Delta x_5^2 \exp\left(-\frac{1}{20}\right) \left[ \frac{5}{4}\left(\Delta x_1 - \frac{1}{5}\Delta x_5\right)^2 + \frac{6}{5}\Delta x_5^2 \right]$$

$$= \frac{1}{20} \int da db \left( a + \frac{1}{5}b \right)^2 \exp\left(-\frac{1}{20}\right) \left[ \frac{5}{4}a^2 + \frac{6}{5}b^2 \right] \quad \begin{matrix} a = \Delta x_1 - \frac{1}{5}\Delta x_5 \\ da db = d\Delta x_1 d\Delta x_5 \end{matrix}$$

$$= \frac{1}{20} \sqrt{\frac{4}{5}} \sqrt{\frac{6}{5}} \sqrt{2\pi} \sigma^2 \left[ \frac{4}{5}\sigma^2 + \frac{1}{25}\left(\frac{6}{5}\sigma^2\right)\sigma^2 \right]$$

$$= \sqrt{\frac{4}{5}} \sqrt{\frac{6}{5}} \left(\frac{5}{2}\sigma^2\right) \frac{1}{20} \quad \begin{matrix} \frac{29}{20} + \frac{1}{30} \rightarrow \frac{25}{20} \\ (0,1) \end{matrix}$$

$$= \frac{5}{2}\sigma^2$$

(5)

$$\langle \delta x_i \delta x_j \rangle = \frac{1}{N} \int_{-\infty}^{+\infty} \exp\left[-\frac{1}{2\sigma^2}\right] \left[ \frac{5}{4}(\alpha_i - \frac{1}{5}\alpha_j)^2 + \frac{6}{5}\alpha_j^2 \right] d\alpha_i d\alpha_j$$

$\alpha = \delta x_i - \frac{1}{5}\delta x_j$

$$= \frac{1}{N} \int d\alpha_i d\alpha_j \left[ \exp\left(-\frac{1}{2\sigma^2}\right) \alpha_i^2 + \frac{6}{5}\alpha_j^2 \right]$$

$$= \frac{1}{5} \times \frac{5\sigma^2}{6} = \frac{1}{6}\sigma^2$$

$$\frac{\langle \delta x_i \delta x_j \rangle}{\sqrt{\langle \delta x_i^2 \rangle} \sqrt{\langle \delta x_j^2 \rangle}} = \frac{\frac{1}{6}\sigma^2}{\frac{5}{6}\sigma^2} = \frac{1}{5} \quad \boxed{\text{O.K.}}$$

Quicker way + get there?  $(\Delta x, \delta x) \begin{pmatrix} \frac{5}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{5}{4} \end{pmatrix} (\delta x)$

$$\delta(\Delta x, \delta x) \approx$$

$$\begin{pmatrix} \frac{5}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{5}{4} \end{pmatrix} \begin{pmatrix} \quad \\ \quad \end{pmatrix}$$

⑥

Now the question is ... how do we do the Fourier Transform  
that includes orientation as well  
as translation?

Suppose we have

$$P(s, x)$$

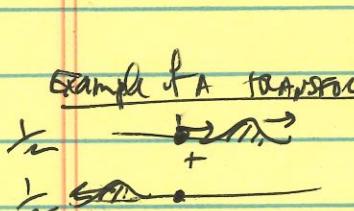
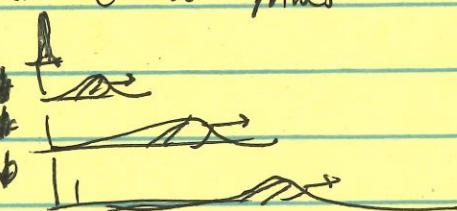
$\uparrow$        $\uparrow$   
position

expand in BASIS SET =  $\sum_{q=1}^{\infty} \tilde{P}(r, q) e^{2\pi i q s}$  wavevect.  
symmetry  
antisymmetry?

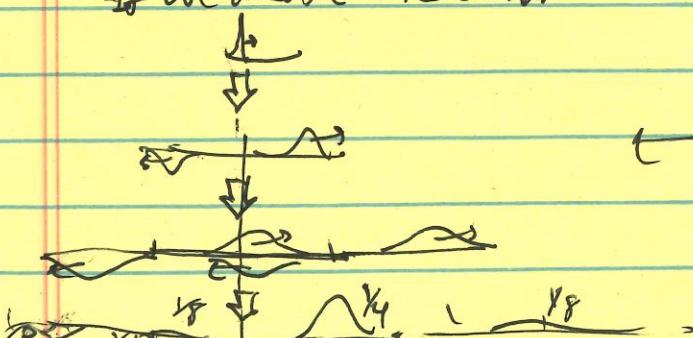
$$\begin{aligned} P(s, x) &= \sum_{r=1}^{\infty} \tilde{P}(r, q) e^{2\pi i q s} \\ &= \underbrace{\sum_{r=1}^{\infty} \sum_{q=1}^{\infty} \tilde{P}(r, q) e^{2\pi i q s}}_{\text{dq}} \\ &= \begin{cases} \int dq [ \tilde{p}(+, q) + \tilde{p}(-, q) ] e^{2\pi i q s} & s=1 \\ \int dq [ \tilde{p}(+, q) - \tilde{p}(-, q) ] e^{2\pi i q s} & s=-1 \end{cases} \\ &= \int dq [ \tilde{p}(+, q) + s \tilde{p}(-, q) ] e^{2\pi i q s} \end{aligned}$$

$$\sum r \tilde{P}(r, q)$$

$$e^{i\pi r}$$

Example of a transform:  

 If we just convolve  $P1, P2$ :  


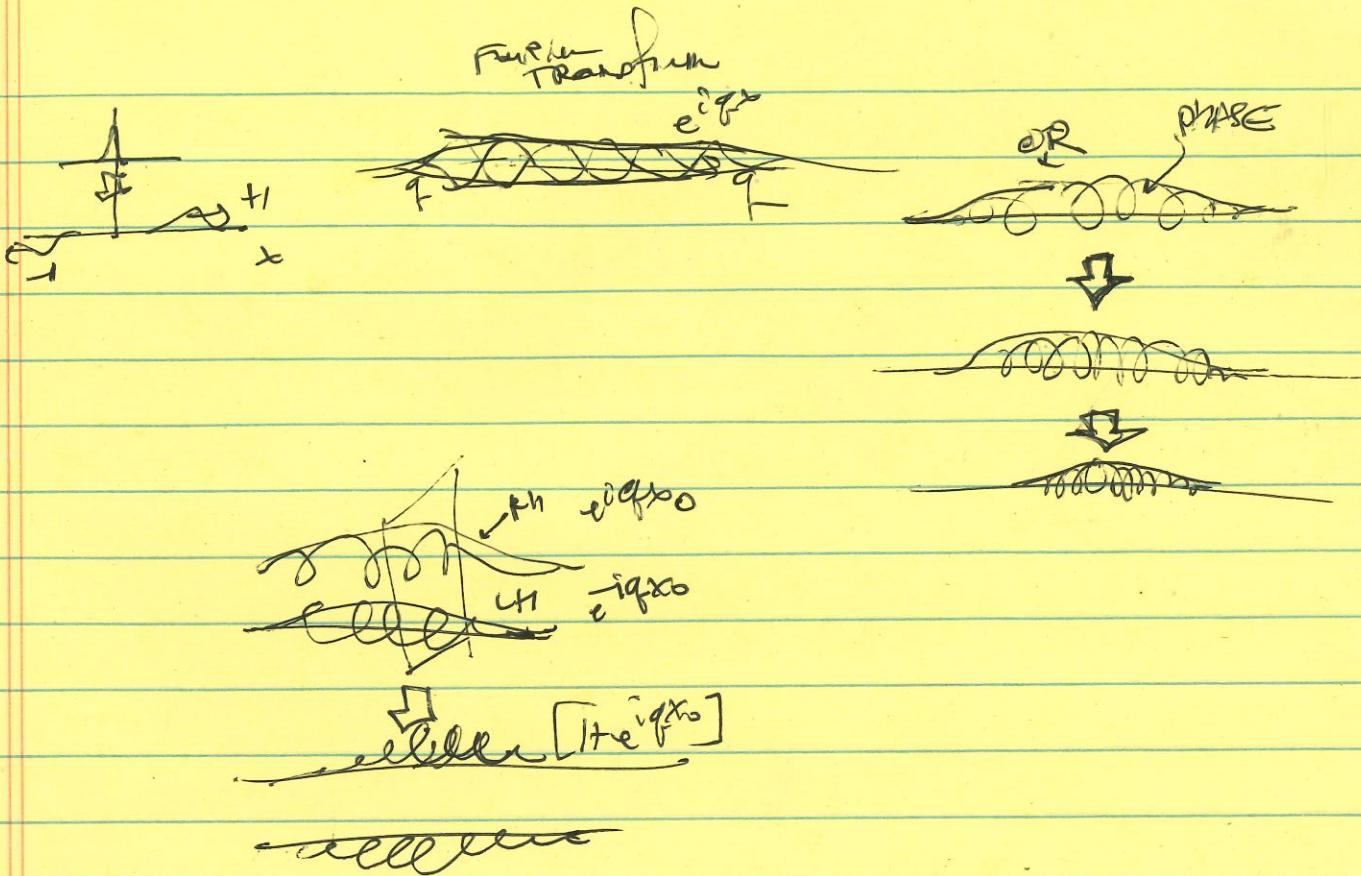
If we consider both branches:



The output is:   
AFTER WHAT TRANSFORM

Does this convolution operation look like a multiplication?

(7)



$$(a_1 e^{i\omega x_0} + b_1 e^{-i\omega x_0})(a_2 e^{i\omega x_0} + b_2 e^{-i\omega x_0}) \\ = a_1^2 e^{2i\omega x_0} + 2a_1 b_1 e^{i\omega x_0} + b_1^2 e^{-i\omega x_0}$$

$$a = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$b = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

I made some separate notes on Fourier Transform Analysis in SE(1), but didn't achieve simple conclusion.

15 JUN 19

on  
R.A.D  
to

SALT WELLS  
SALT  
MINES  
P.W.A.M

$$\text{Forward } \begin{pmatrix} e^{ikx_0} \\ e^{-ikx_0} \end{pmatrix} \rightarrow \text{Forward } \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \text{FWD}$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \rightarrow \text{Forward } \begin{pmatrix} \delta(x_0) \\ \delta(k_0) \end{pmatrix} \quad \text{FWD}$$

~~15 JUN 19~~  $\frac{\partial}{\partial x_0}$

Imagine transform that  
just move east direction  
forward by  $\Delta x_0$  ...

$$\begin{pmatrix} e^{ikx_0} & 0 \\ 0 & e^{-ikx_0} \end{pmatrix}$$

check:

$$\begin{pmatrix} e^{ikx_0} & 0 \\ 0 & e^{-ikx_0} \end{pmatrix} \begin{pmatrix} e^{ikx_0} \\ e^{-ikx_0} \end{pmatrix} = \begin{pmatrix} e^{ik(x_0 + \Delta x_0)} \\ e^{-ik(x_0 - \Delta x_0)} \end{pmatrix}$$

Reversal ~~transform~~  
directions

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} e^{ikx_0} \\ e^{-ikx_0} \end{pmatrix} = \begin{pmatrix} e^{-ikx_0} \\ e^{ikx_0} \end{pmatrix}$$

E

Could do set directions as:

$$e^{ikx_0} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} e^{ikx_0} \\ e^{-ikx_0} \end{pmatrix}$$

$$\begin{pmatrix} e^{ikx_0} & 0 \\ 0 & e^{-ikx_0} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} e^{ikx_0} \\ e^{-ikx_0} \end{pmatrix}$$

Forward?

$$(0 = 0, 1) \quad e^{ikx_0} + e^{i\pi\theta} e^{-ikx_0}$$

$$\begin{pmatrix} x_0 \\ 1 \end{pmatrix}$$

multiple by

$$e^{i\pi\theta} e^{ikx_0} + e^{2i\pi\theta} e^{-ikx_0}$$



A  
B  
C  
D  
PARALLEL

ANTIPARALLEL

$$\boxed{\text{BASIS FUNCTIONS}} = \begin{pmatrix} +e^{ikx} \\ -e^{-ikx} \end{pmatrix}, \quad \begin{pmatrix} +e^{ikx} \\ -e^{-ikx} \end{pmatrix}$$

$$P(x) = \begin{pmatrix} A(x) \\ B(x) \end{pmatrix}$$

$$\tilde{P}_+(k) = \int_{-\infty}^{\infty} (e^{ikx} e^{-ibx}) \begin{pmatrix} A(x) \\ B(x) \end{pmatrix}$$

$$= \int_{-\infty}^{\infty} (e^{ikx} A(x)) + (e^{-ibx} B(x))$$

$$= \tilde{A}(k) + \tilde{B}(k)$$

$$\tilde{P}_-(k) = \int_{-\infty}^{\infty} (e^{ikx} - e^{-ibx}) \begin{pmatrix} A(x) \\ B(x) \end{pmatrix} = \tilde{A}(k) - \tilde{B}(k)$$

BASIS TRANSFORMATION:

$$\varepsilon = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$e^{i\theta\varepsilon} = e^{i\theta} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} e^{i\theta k} & 0 \\ 0 & e^{-i\theta b} \end{pmatrix}$$

$$P(x) = \left[ \tilde{P}_+(k) \begin{pmatrix} e^{ikx} \\ -e^{-ikx} \end{pmatrix} + \tilde{P}_-(k) \begin{pmatrix} +e^{ikx} \\ -e^{-ikx} \end{pmatrix} \right] dk$$

$$\text{move forward by } \begin{pmatrix} e^{ikx} & 0 \\ 0 & -e^{-ikx} \end{pmatrix} \rightarrow P(x) = \begin{pmatrix} e^{ikx} & 0 \\ 0 & e^{-ikx} \end{pmatrix} P(k)$$

$$\text{repeat and move by } \begin{pmatrix} 0 & e^{iq} \\ 0 & 0 \end{pmatrix}$$

$$= \tilde{P}_+(k) \begin{pmatrix} e^{ik(b+ba)} \\ -e^{-ik(b+ba)} \end{pmatrix} + \tilde{P}_-(k) \begin{pmatrix} +e^{ik(b+ba)} \\ -e^{-ik(b+ba)} \end{pmatrix}$$

see next page.

(B) 10

Let's work this out again:-

$$P(x) = \begin{pmatrix} A(x) \\ B(x) \end{pmatrix} = \int_{-\infty}^{\infty} \left[ \tilde{P}_+(k) \frac{e^{ikx}}{2} + \tilde{P}_-(k) \frac{e^{-ikx}}{2} \right] dk$$

Forward  
POINTERS  
BACKWARD  
POINTERS

check: PROPAGATION IN DIRECTION OF POINTERS IS SIMPLY  
multiplication  $\neq$  by  $e^{ikx}$ :

$$\begin{aligned} P(x) &= \int_{-\infty}^{\infty} \left[ \tilde{P}_+(k) \frac{e^{ik(x+\Delta x)}}{2} e^{-ik(x-\Delta x)} \right] dk + \tilde{P}_-(k) \left( \frac{e^{ik(x+\Delta x)}}{2} - \frac{e^{-ik(x-\Delta x)}}{2} \right) dk \\ &= \begin{pmatrix} A(x+\Delta x) \\ B(x-\Delta x) \end{pmatrix} \end{aligned}$$

$k \rightarrow -k$   
~~original equation~~

check: REVERSE POINTERS DIRECTION IS  $\hat{F}x(-\frac{\partial}{\partial x}) @ \tilde{P}_+(k) \neq \tilde{P}_-(k)$

$$\begin{aligned} P'(x) &= \int_{-\infty}^{\infty} \left[ \tilde{P}_+(k) \left( e^{-ik(x+\Delta x)} + e^{ik(b-\Delta x)} \right) \right] dk + \tilde{P}_-(k) \left( e^{ik(x-\Delta x)} - e^{-ik(b-\Delta x)} \right) dk \\ &= \begin{pmatrix} B(x) \\ A(x) \end{pmatrix} \end{aligned}$$

So... lets check two ROD convolutions:

$\xrightarrow{\text{FORWARD}}$   
 $\xleftarrow{\text{BACKWARD}}$

One transform:

$$e\left(\begin{pmatrix} u \\ v \end{pmatrix}\right) e\left(\begin{pmatrix} u' \\ v' \end{pmatrix}\right) = e^{i(k_x u - k_y v)} e^{i(k_x u' - k_y v')} = e\left(\begin{pmatrix} u \\ v \end{pmatrix}\right) e\left(\begin{pmatrix} u' \\ v' \end{pmatrix}\right)$$

uh wait... NOT WORKING  $\rightarrow$  the book

further reading  
the PIP?

$\hookrightarrow$  Not complex conjugate?  
 $\mathcal{R} \leftarrow \mathcal{L}$

(4) 11

$$\star F_p P(b) = P_{T+}(-k) = P_{T-}^*(b) \quad (\pm \text{ think})$$

$$F_{\text{flip}}(0,1) e^{ikx - \frac{\pi^2 k^2}{2}} \quad F_{\text{flip}}(0,1) e^{ikx - \frac{\pi^2 k^2}{2}}$$

$$F_{\text{flip}}(0,1) e^{ikx - \frac{\pi^2 k^2}{2}} \left( \begin{array}{cc} 0 & e^{-ikx - \frac{\pi^2 k^2}{2}} \\ e^{ikx} & 0 \end{array} \right)$$

$$= \text{[Diagram of two loops]} e^{-\frac{\pi^2 k^2}{2}} \text{[Diagram of two loops]} \quad (\dots)$$

looks  $\propto k$

$\stackrel{\leftrightarrow}{\sum}$

$$\left\{ \frac{1}{2} \left[ \text{[Diagram]} I + P(0,1) \right] \right\} e^{-\frac{\pi^2 k^2}{2}}$$

$$\sum_{m=1}^n \frac{1}{2^n} \binom{n}{m} e^{-\frac{\pi^2 k^2}{2}} \left[ \underset{x}{\text{[Diagram]}} + \underset{-x}{\text{[Diagram]}} \right]$$

- might work out o.k.

- The  $F^*$  thing is not very clear - mix  $k$  and  $-k$ ?

How about instead  $P_-(b) \rightarrow -P_+(b)$ ?

(8) (12)

Did not complete - REVISIT LMER

Harmonic mode analysis: First, do two-joint.

$$\boxed{\downarrow} \quad z = \int e^{-\frac{x^2}{2\sigma^2}} \frac{e^{-\frac{y^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} \delta(x-y) dy$$

$$= \int e^{-\frac{x^2}{2\sigma^2}} \frac{dx}{\sqrt{2\pi\sigma^2}} = \frac{\sqrt{2\pi(\frac{3}{2})}}{2\sigma^2} = \boxed{\frac{1}{\sqrt{2\pi\sigma^2}}}$$

$\begin{pmatrix} 0 & 0 \\ 0 & 4 \\ 0 & 0 \end{pmatrix}$   
 $\begin{pmatrix} 0 & ? \\ 2(1-\cos\pi) & 0 \end{pmatrix}$

Compared to no-delta function:

$$z = \int e^{-\frac{x^2}{2\sigma^2}} \frac{e^{-\frac{y^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} dy = 1.$$

$$z = \int e^{-\frac{(x-a)^2}{2\sigma^2}} e^{-\frac{(x-b)^2}{2\sigma^2}} e^{-\frac{(x-c)^2}{2\sigma^2}} e^{-\frac{(x-d)^2}{2\sigma^2}} e^{-\frac{(x-e)^2}{2\sigma^2}} \frac{1}{(\sqrt{2\pi\sigma^2})^5} \frac{1}{\det[\frac{x_i - x_j}{\sigma^2}]}$$

$$= \int e^{-\frac{(x-a)^2}{2\sigma^2}} \left[ 2\pi b^2 c^2 d^2 e^2 + (-a-b-c-d-e)^2 \right] \text{ bubble bubble}$$

$$= \int e^{-\frac{(x-a)^2}{2\sigma^2}} \prod_{i=1}^{n-1} \prod_{j=i+1}^n \begin{pmatrix} a+b+c+d+e & 2^{-1} & \dots & -1 \\ 2 & 2 & \dots & 2 \\ \vdots & \vdots & \ddots & \vdots \\ 2 & 2 & \dots & 2 \end{pmatrix} \begin{pmatrix} 1 \\ b \\ c \\ d \\ e \end{pmatrix}$$

Alternatively,

$$z = \int e^{-\frac{(x-a)^2}{2\sigma^2}} \left[ \begin{pmatrix} x_1 & x_2 & x_3 & x_4 & x_5 \\ x_1 & x_2 & x_3 & x_4 & x_5 \end{pmatrix} \begin{pmatrix} 2^{-1} & & & & -1 \\ 1 & 2 & -1 & & \\ -1 & 2 & 1 & & \\ 1 & -2 & -1 & & \\ 1 & 1 & 1 & 2 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \right]$$

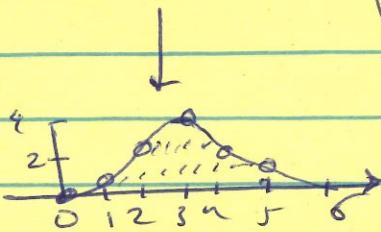
Normal  
Mode

$$A \begin{pmatrix} f_1 \\ f_2 \\ f_3 \\ f_4 \\ f_5 \end{pmatrix} = \sigma \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}$$

$$A \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{pmatrix} = \sigma \begin{pmatrix} e^{2\pi i \frac{1}{2}\sigma} \\ e^{2\pi i \frac{2}{2}\sigma} \\ e^{2\pi i \frac{3}{2}\sigma} \\ e^{2\pi i \frac{4}{2}\sigma} \\ e^{2\pi i \frac{5}{2}\sigma} \end{pmatrix}$$

$$\text{char. } 2 - e^{2\pi i \frac{1}{2}\sigma} + e^{2\pi i \frac{5}{2}\sigma}$$

$$A \begin{pmatrix} e^{j\frac{2\pi}{6}n\omega_0} \\ e^{j\frac{2\pi}{6}(n+1)\omega_0} \\ \vdots \\ e^{j\frac{2\pi}{6}(n+5)\omega_0} \end{pmatrix} = 2 \left[ 1 - \cos\left(\frac{2\pi}{6}\right) \right] \begin{pmatrix} 1 \\ e^{j\frac{2\pi}{6}\omega_0} \\ \vdots \\ e^{j\frac{2\pi}{6}(n+5)\omega_0} \end{pmatrix}$$



Intrinsic...

$$Z = \cancel{x} \times \frac{1}{\sqrt{2^5 [1 - \cos(\frac{2\pi}{6})][1 - \cos(\frac{2\pi}{6})]} \cdot [1 - \cos(\frac{2\pi}{6})]} \cdot [1 - \cos(\frac{2\pi}{6})]$$

zero mode  
5 vibrations mode



$$Z = \cancel{\omega} \frac{2\pi}{6} = \frac{1}{3}$$

closed.

$$= L \cancel{\left( \frac{1}{2\pi} \right)^5} \frac{1}{\sqrt{2^5 \left(\frac{1}{2}\right)^2 \cdot \left(\frac{2}{3}\right)^2 \cdot 2}}$$

$$= L \cancel{\left( \frac{1}{2\pi} \right)^5} \frac{1}{(3 \cdot 2)} = \frac{L}{\cancel{(2\pi)^5} \cdot \cancel{6} \cdot \cancel{5}}$$

~~$$Z_{\text{error}} = L \cdot \cancel{\left( \frac{1}{2\pi} \right)^5}$$~~

$$\frac{Z_{\text{error}}}{Z_{\text{true}}} = \frac{1}{\cancel{2\pi}^5 \times \cancel{6}} \stackrel{0.1\%}{=} \text{no } n \text{ should be } \frac{1}{\cancel{2\pi}^6 \cancel{5}}$$

How to resolve?

Go into MATLAB → calculate  
determinants  
in matrix  
check.

$$\cdot \text{ calculate } \prod_{m=1}^{N-1} \left( 1 - \cos \frac{2\pi m}{N} \right)$$

With mode analysis to  
estimate res. location too.  
from difference