

# Implementation of fully analytic orbit response analysis in python

Formulas used in this slides are derived from: <https://arxiv.org/abs/1711.06589v2>



S.Liuzzo, A.Franchi, February 13<sup>th</sup> 2024

The European Synchrotron

**Analytic formulas for the rapid evaluation of the orbit response matrix and chromatic functions from lattice parameters in circular accelerators**

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(Dated: September 23, 2018)

# OUTLINE

- **Part 1: The implementation (12')**
- **Part 2: The repository (1')**
- **Part 3: outlook (3')**

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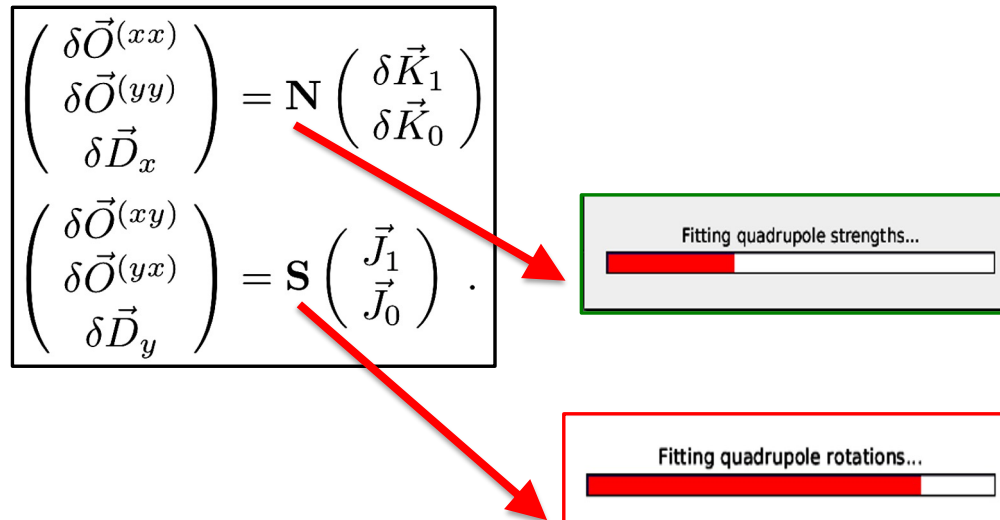
## JACOBIAN OF THE ORBIT RESPONSE MATRIX

- In order to perform a linear lattice modelling and correction, the Jacobian of the ORM needs to be computed, SVD pseudo-inverted & applied to the measured ORM & dispersion.

$$\begin{pmatrix} \delta\vec{O}^{(xx)} \\ \delta\vec{O}^{(yy)} \\ \delta\vec{D}_x \end{pmatrix} = \mathbf{N} \begin{pmatrix} \delta\vec{K}_1 \\ \delta\vec{K}_0 \end{pmatrix}$$
$$\begin{pmatrix} \delta\vec{O}^{(xy)} \\ \delta\vec{O}^{(yx)} \\ \delta\vec{D}_y \end{pmatrix} = \mathbf{S} \begin{pmatrix} \vec{J}_1 \\ \vec{J}_0 \end{pmatrix} .$$

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  - ✓ Pros: accurate, can be parallelized
  - ✓ Cons: time consuming, if optics or orbit unstable for quadrupole variation it can fail.



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- **Pseudo-analytic Jacobian: use textbook formulas to evaluate ORM after computing Twiss parameters at each quadrupole variation.**
  - ✓ Pros: faster than numerical, can be parallelized
  - ✓ Cons: no analytic formulas for off-diagonal ORM blocks (coupling), it can fail if optics or orbit unstable for quadrupole variation.

$$M_{ij} = \frac{\sqrt{\beta_j \beta_i}}{2 \sin \pi Q} \cos (|\phi_j - \phi_i| - \pi Q) .$$

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- **Fully analytic Jacobian: evaluate directly the Jacobian from Twiss parameters of the initial model (ideal or from beam-based measurements)**
  - ✓ Pros: coupling & dispersion included, only one computation of Twiss parameters needed, no orbit calculation needed, faster than pseudo-analytic, can be parallelized
  - ✓ Cons: tedious to code.

# JACOBIAN OF THE ORBIT RESPONSE MATRIX

- In order to perform a linear lattice modelling and correction, the Jacobian of the ORM needs to be computed

- Numerical strength

✓ Pros: accurate

✓ Cons: time-consuming  $w \rightarrow steerer$

- Pseudo-analytic for each quadrupole

✓ Pros: fast

✓ Cons: not accurate

- Fully analytic Jacobian: evaluate directly the Jacobian from beam-based measurements

✓ Pros: coupling & dispersion included, only one computation of Twiss parameters, pseudo-analytic, can be parallelized

✓ Cons: tedious to code.

$$N_{wj,m}^{(xx)} \simeq - \frac{\sqrt{\beta_{j,x}^{(mod)} \beta_{w,x}^{(mod)} \beta_{m,x}^{(mod)}}}{2 \sin(\pi Q_x^{(mod)})} \left\{ \begin{aligned} & \frac{\cos(\tau_{x,wj}^{(mod)})}{4 \sin(2\pi Q_x^{(mod)})} \left[ \cos(2\tau_{x,mj}^{(mod)}) + \cos(2\tau_{x,mw}^{(mod)}) \right] \\ & + \frac{\sin(\tau_{x,wj}^{(mod)})}{4 \sin(2\pi Q_x^{(mod)})} \left[ \sin(2\tau_{x,mj}^{(mod)}) - \sin(2\tau_{x,mw}^{(mod)}) \right] \\ & + \frac{1}{2} \sin(\tau_{x,wj}^{(mod)}) \left[ \Pi(m, j) - \Pi(m, w) + \Pi(j, w) \right] + \frac{\cos(\Delta\phi_{x,wj}^{(mod)})}{4 \sin(\pi Q_x^{(mod)})} \end{aligned} \right\},$$

$$\Pi(a, b) = 1 \quad \text{if } s_a < s_b, \quad \Pi(a, b) = 0 \quad \text{if } s_a \geq s_b$$

$$\tau_{z,ab} = \Delta\phi_{z,ab} - \pi Q_z, \quad z = x, y$$



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- **Fully analytic Jacobian: evaluate directly the Jacobian from beam-based measurements)**
  - ✓ Pros: coupling & dispersion included, only one computation per quadrupole, pseudo-analytic, can be parallelized
  - ✓ Cons: tedious to code.
- **Observation: the accuracy of the two analytic approaches can be poor if thin-element model is used. Corrections to account for the variation of Twiss parameters across magnets have been included which reduce dramatically the errors w.r.t. the numerical version (see next slide).**

$$\begin{aligned}
 \beta_m &\longrightarrow I_{\beta,m} = \frac{1}{L_m} \int_0^{L_m} \beta(s) ds, \\
 \beta_m \sin(2\tau_{mj}) &\longrightarrow I_{S,mj} = \frac{1}{L_m} \int_0^{L_m} \beta(s) \sin(2\tau_{sj}) ds \\
 \beta_m \cos(2\tau_{mj}) &\longrightarrow I_{C,mj} = \frac{1}{L_m} \int_0^{L_m} \beta(s) \cos(2\tau_{sj}) ds \\
 \sqrt{\beta_m} \sin(\tau_{mj}) &\longrightarrow J_{C,mj} = \frac{1}{L_m} \int_0^{L_m} \sqrt{\beta(s)} \sin(\tau_{sj}) ds \\
 \sqrt{\beta_m} \cos(\tau_{mj}) &\longrightarrow J_{C,mj} = \frac{1}{L_m} \int_0^{L_m} \sqrt{\beta(s)} \cos(\tau_{sj}) ds
 \end{aligned}$$

or

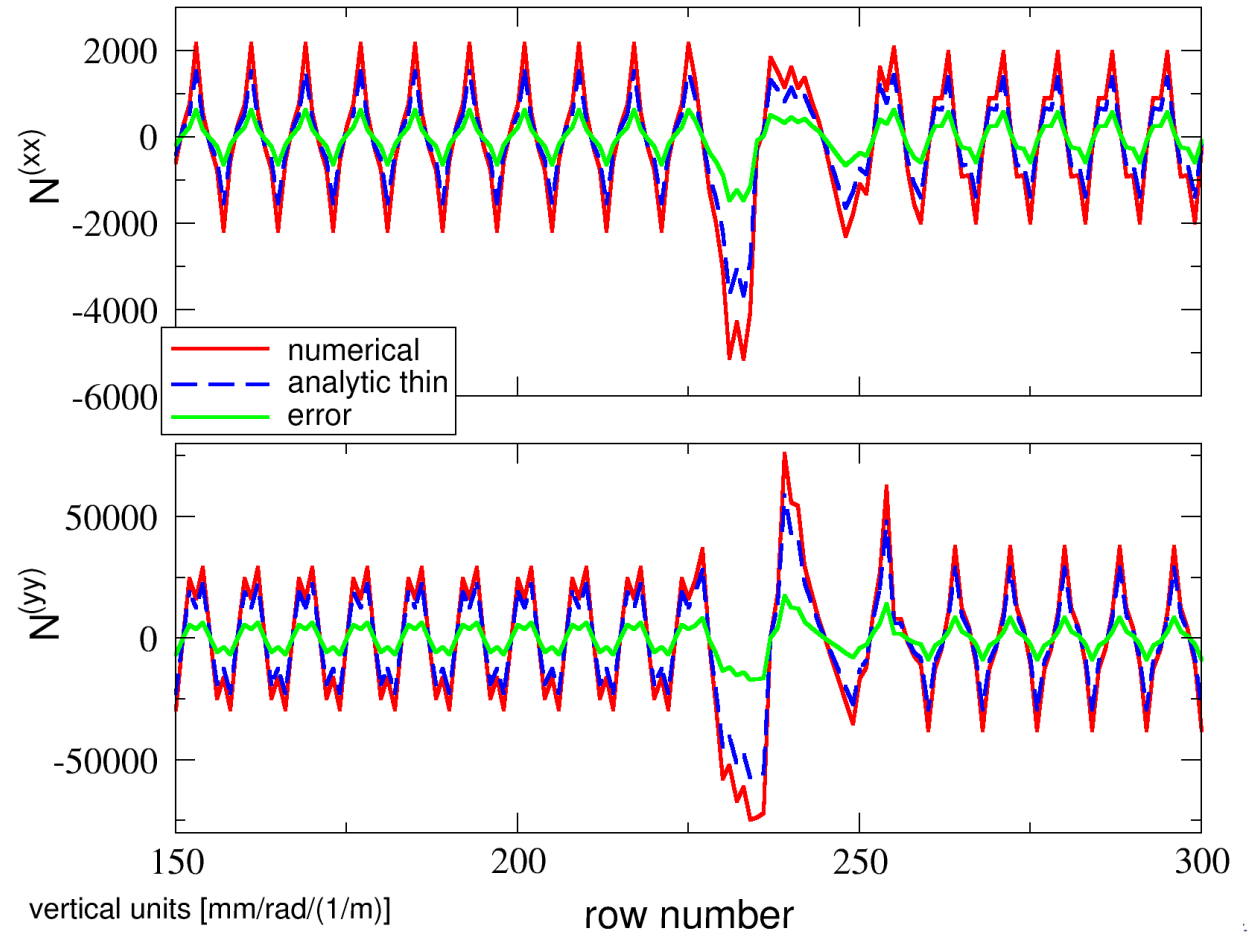
# NUMERICAL VS ANALYTIC JACOBIAN OF THE ORM: ACCURACY

Example: FCC quadrupole ORM Jacobian with

- 1600 BPMs
- 8 steerers
- 1 quadrupole QC1L1\_1

$$\begin{pmatrix} \delta \vec{O}(xx) \\ \delta \vec{O}(yy) \\ \delta \vec{D}_x \end{pmatrix} = \mathbf{N} \begin{pmatrix} \delta \vec{K}_1 \\ \delta \vec{K}_0 \end{pmatrix}$$

THIN



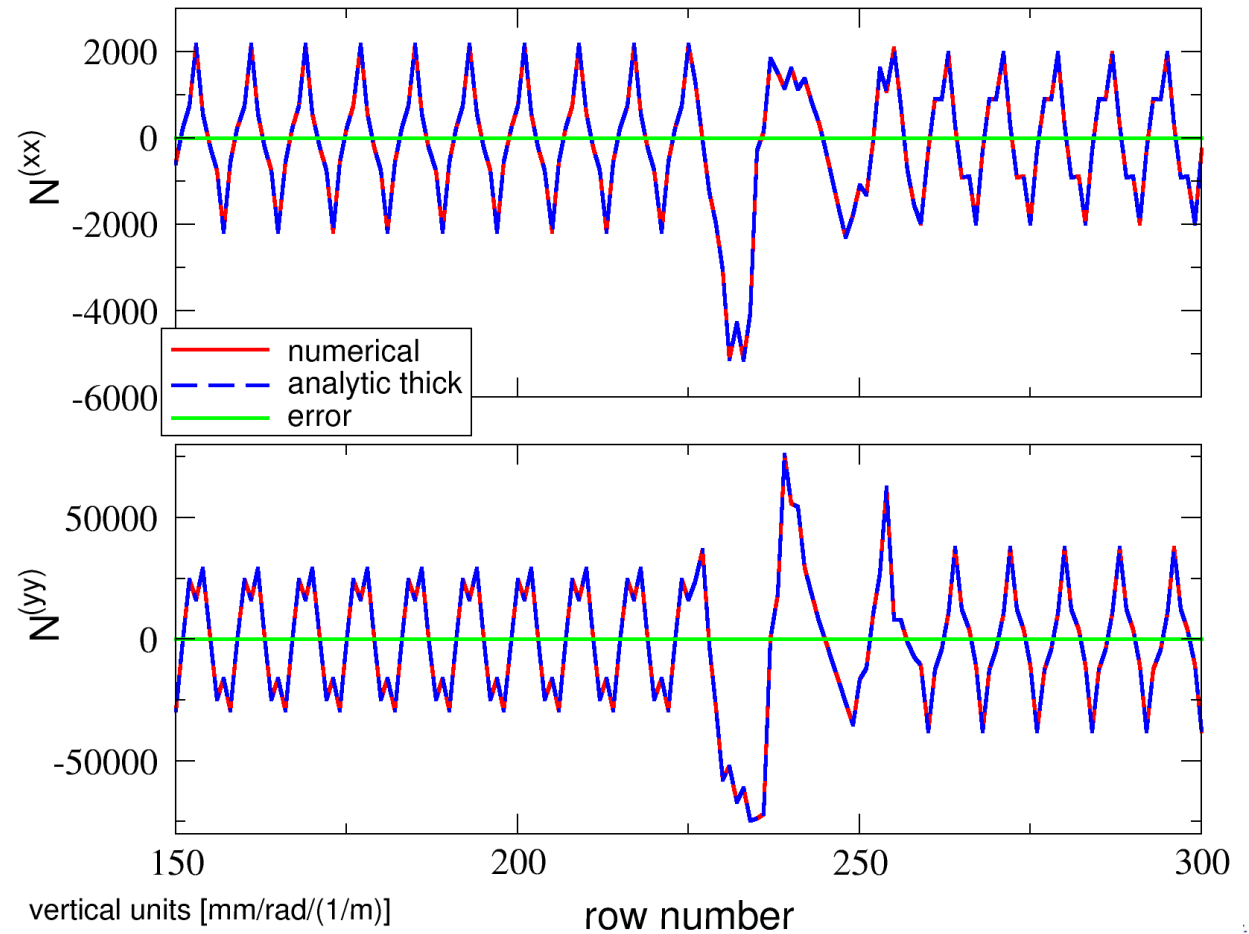
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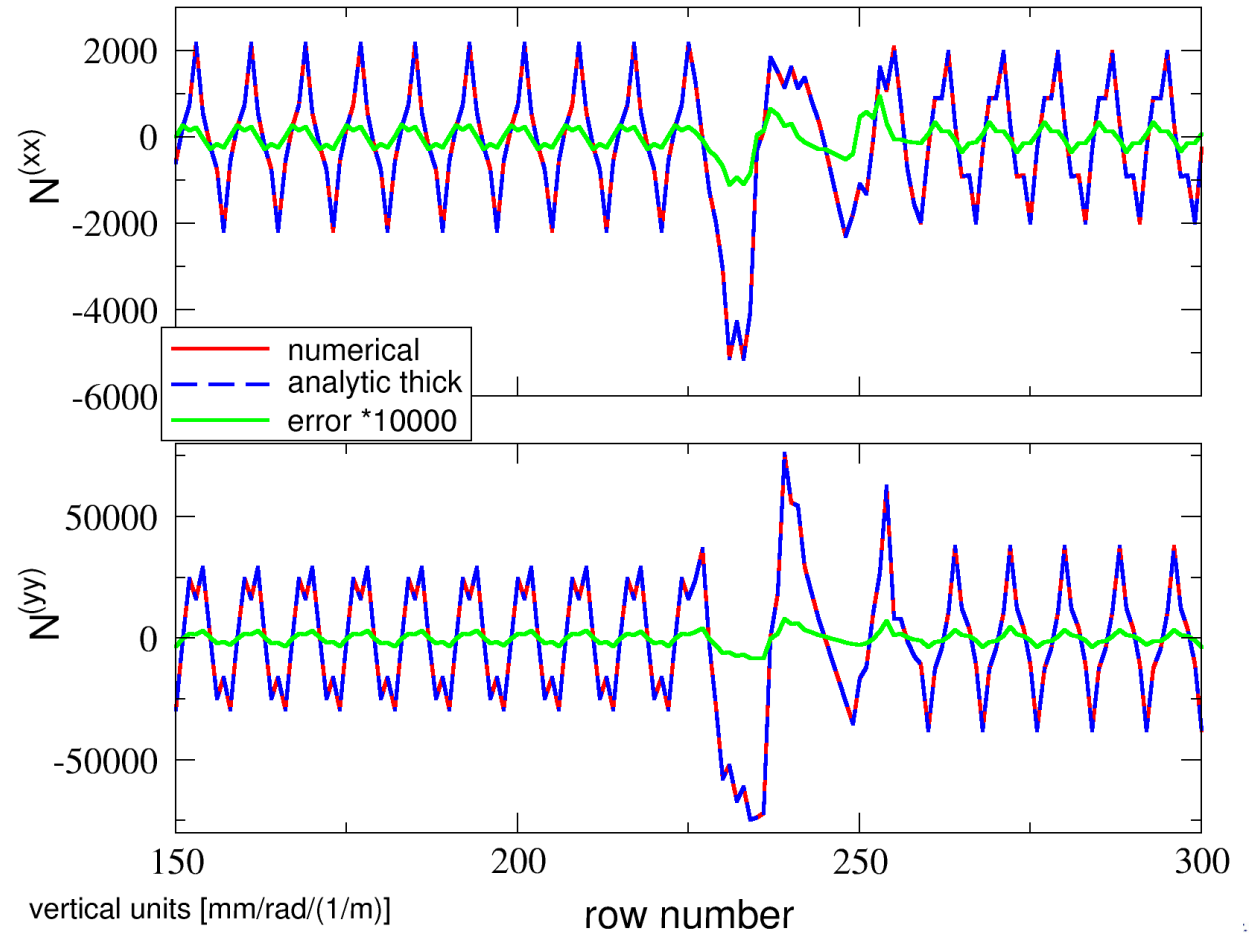
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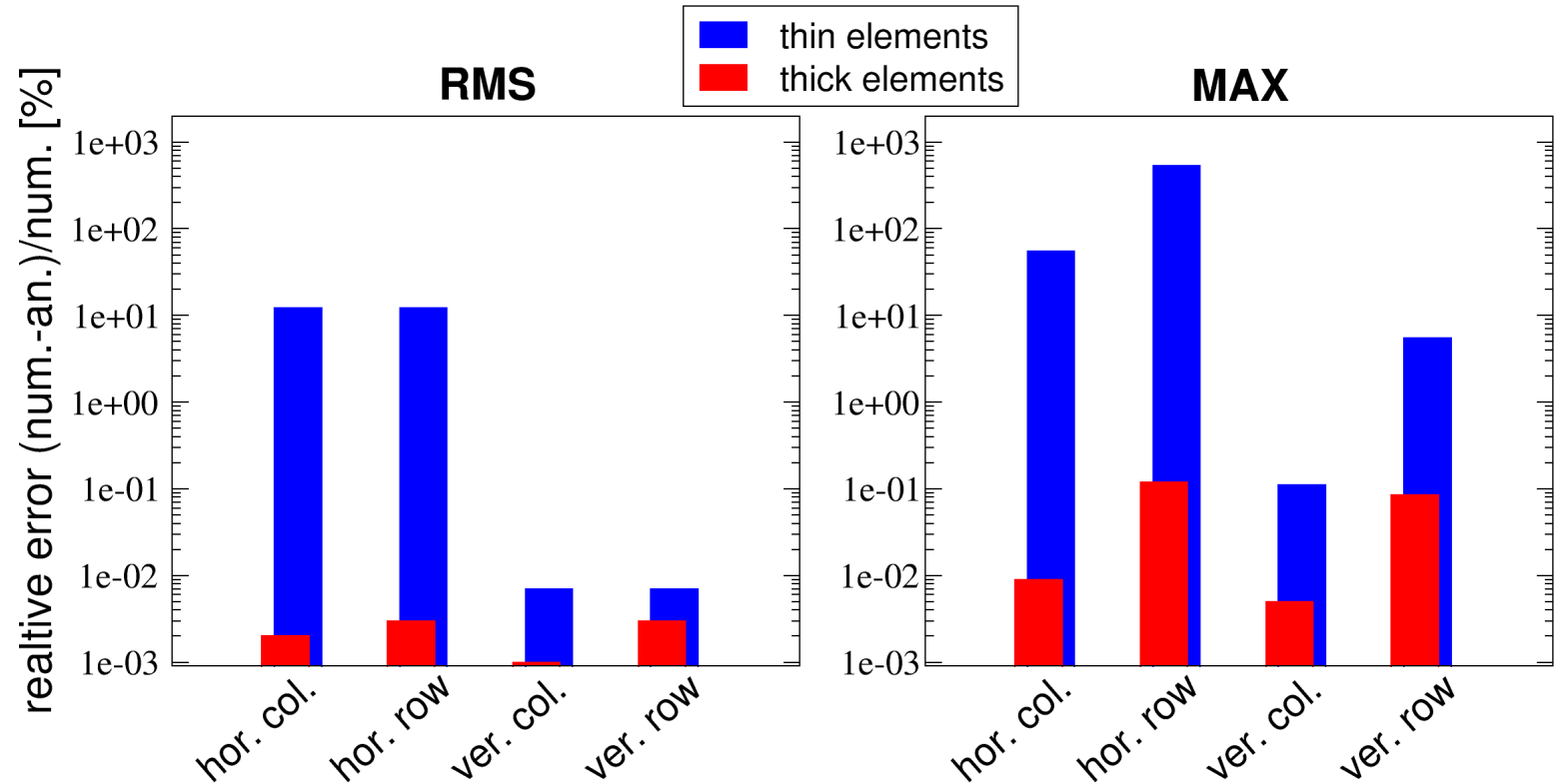
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$$\text{ORM} = \begin{pmatrix} \mathbf{O}^{(xx)} & \mathbf{O}^{(xy)} \\ \mathbf{O}^{(yx)} & \mathbf{O}^{(yy)} \end{pmatrix}$$

- RMS & MAX error computed over all columns & rows of the diagonal ORM blocks  $\mathbf{O}^{(xx)}$  &  $\mathbf{O}^{(yy)}$ .

## NUMERICAL VS ANALYTIC JACOBIAN OF THE ORM: CPU TIME

Example: FCC quadrupole ORM Jacobian  $N$  (diagonal blocks only) with

- 1600 BPMs
- 8 steerers
- 360 quadrupoles parallelized over 64 cores CPUs (for both numerical and analytic tests)

### Results

- Numeric: 1807.1 s [100%]
- fully analytic: 221.1 s [12%] (room for further optimization)

## NUMERICAL VS ANALYTIC JACOBIAN OF THE ORM: OPTICS CORRECTION

Table 2:  $\beta$ -beating, dispersion and emittances after correction of 10  $\mu\text{m}$  random alignment errors on dipole quadrupole and sextupole magnets for the EBS and FCC-ee lattices using analytic or numeric ORM derivative.

$\langle std \rangle_{50}$ units	$\frac{\Delta\beta_h}{\beta_{h,0}}$ %	$\frac{\Delta\beta_v}{\beta_{v,0}}$ %	$\Delta\eta_h$ mm	$\Delta\eta_v$ mm	$\Delta\epsilon_v$ pm rad
EBS					
err.	19.37	11.08	17.33	6.91	94.17
ana.	0.2	0.2	0.18	0.05	0.003
num.	0.2	0.2	0.18	0.05	0.003
FCC-ee Z					
err.	3.6	59.4	120.5	82.45	-
ana.	0.81	4.29	26.0	9.57	0.17
num.	0.82	4.30	25.98	9.64	0.18

From IPAC23 MOPL069

Table 3:  $\beta$ -beating, dispersion and emittances after correction of 10  $\mu\text{m}$  random alignment errors on dipole quadrupole and sextupole magnets for the FCC-ee lattice using analytic ORM derivative (1856 BPMs, 18 steerers). The input lattice is tested: without radiation, with radiation and with radiation and tapering. Reference lattice is in all cases without radiation.

$\langle std \rangle_{50}$ units	$\frac{\Delta\beta_h}{\beta_{h,0}}$ %	$\frac{\Delta\beta_v}{\beta_{v,0}}$ %	$\Delta\eta_h$ mm	$\Delta\eta_v$ mm	$\Delta\epsilon_v$ pm rad
4D err	3.63	61.37	118.7	82.36	-
4D cor	0.84	4.24	25.67	9.58	0.71
6D err	3.60	59.45	120.54	82.45	-
6D cor	0.81	4.29	26.0	9.57	0.17
6D err + tapering	3.61	61.33	119.59	82.96	-
6D cor + tapering	0.82	4.22	26.03	9.65	0.18

# ANALYTIC TUNE VARIATION WITH THICK QUADRUPOLES

EBS

Quad:  $\Delta QF1J=0.023\%$

error  $\Delta QH$  (num-ana):  $-0.16\%$

error  $\Delta Qv$  (num-ana):  $0.14\%$

THIN

$$\frac{\Delta Q}{\Delta K_j} = \frac{\beta_j}{4\pi}$$

average  $\beta$  function across quadrupole

$$\frac{\Delta Q}{\Delta K_j} = \frac{\frac{1}{2} \left[ \beta_j + \frac{\gamma_j}{K_j} \right] + \frac{\sin(2\sqrt{K_j L_j})}{4\sqrt{K_j L_j}} \left[ \beta_j - \frac{\gamma_j}{K_j} \right] + \frac{\alpha_j}{2K_j L_j} [\cos(2\sqrt{K_j L_j}) - 1]}{4\pi}$$

error  $\Delta QH$  (num-ana):  $-0.0005\%$

error  $\Delta Qv$  (num-ana):  $0.0024\%$

THICK

From IPAC23 MOPL069

(“Thick modelling” original idea and mathematical approach : Zeus Martí , ALBA)



# ANALYTIC TUNE VARIATION WITH THICK QUADRUPOLES

FCC-ee

Quad:  $\Delta Q_{FG2-1} = 0.184\%$

error  $\Delta Q_H$  (num-ana): 2.2%

error  $\Delta Q_v$  (num-ana): 0.38%

THIN

$$\frac{\Delta Q}{\Delta K_j} = \frac{\beta_j}{4\pi}$$

average  $\beta$  function across quadrupole

$$\frac{\Delta Q}{\Delta K_j} = \frac{\frac{1}{2} \left[ \beta_j + \frac{\gamma_j}{K_j} \right] + \frac{\sin(2\sqrt{K_j L_j})}{4\sqrt{K_j L_j}} \left[ \beta_j - \frac{\gamma_j}{K_j} \right] + \frac{\alpha_j}{2K_j L_j} [\cos(2\sqrt{K_j L_j}) - 1]}{4\pi}$$

error  $\Delta Q_H$  (num-ana): 0.0003 %

error  $\Delta Q_v$  (num-ana): -0.0008 %

THICK

From IPAC23 MOPL069

(“Thick modelling” original idea and mathematical approach : Zeus Martí , ALBA)

# THICK STEERERS IN CLASSIC ANALYTIC ORM FORMULA

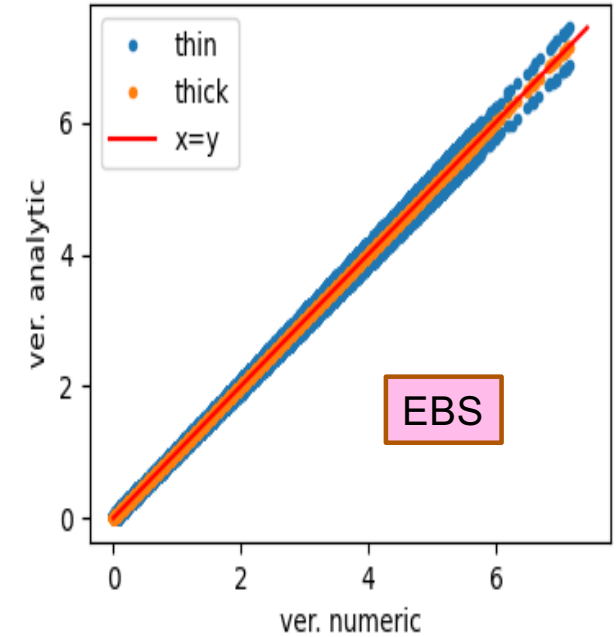
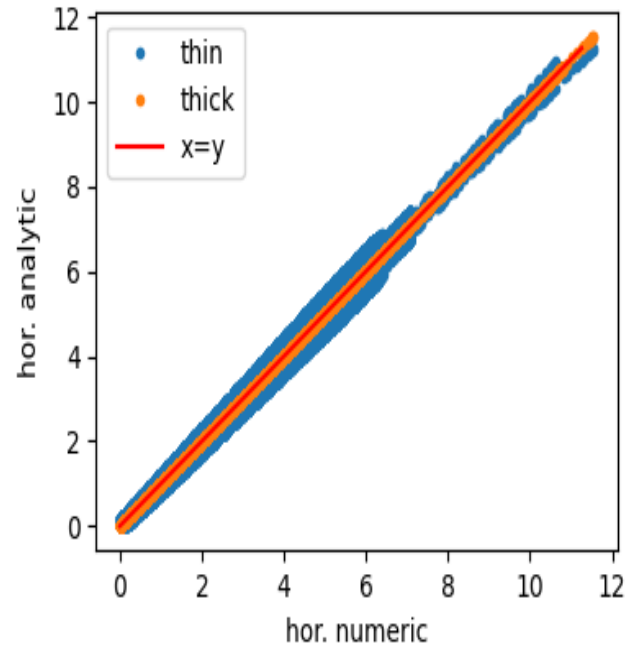
## THIN steerers ORM

$$\vec{x} = M\vec{\theta} \quad ,$$

$$M_{ij} = \frac{\sqrt{\beta_j \beta_i}}{2 \sin \pi Q} \cos (|\phi_j - \phi_i| - \pi Q) + \frac{\eta_j \eta_i}{L_0 \alpha_c}$$

## THICK steerers ORM

$$M_{i,j} = \frac{\sqrt{\beta_i}}{2 \sin \pi Q} \left[ \left( \sqrt{\beta_j} - \frac{\alpha_j L_j}{2\sqrt{\beta_j}} \right) \cos (|\phi_j - \phi_i| - \pi Q) + \frac{L_j}{2\sqrt{\beta_j}} \sin (|\phi_j - \phi_i| - \pi Q) \right] + \frac{\eta_i \eta_j}{L_0 \alpha_c}$$



# OUTLINE

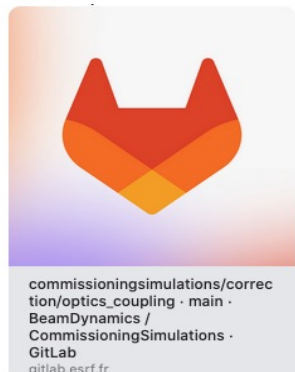
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## WHERE YOU CAN FIND US

<https://gitlab.esrf.fr/BeamDynamics/commissioningsimulations>

**Commissioning tools are still poor in terms of documentation, and debugging. The inclusion into the pyAT repository is pending such extensive validation tests.**

**Modules for the fully analytic ORM Jacobian can be found here:**



[https://gitlab.esrf.fr/BeamDynamics/commissioningsimulations/-/tree/main/commissioningsimulations/correction/optics\\_coupling](https://gitlab.esrf.fr/BeamDynamics/commissioningsimulations/-/tree/main/commissioningsimulations/correction/optics_coupling)

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# OUTLOOK: THICK-ELEMENT CORRECTION FOR RDTs

$$f_{jklm}(s) \propto \sum_m \beta_{m,x}^{(J+k)/2} \beta_{m,y}^{(l+m)/2} e^{i[(j-k)\Delta\phi_{w,x}^{(s)} + (l-m)\Delta\phi_{w,y}^{(s)}]}$$

$$f_{1001,j} = \frac{\sum_{m=1}^M J_{m,1} \sqrt{\beta_{m,x} \beta_{m,y}} e^{i(\Delta\phi_{x,mj} - \Delta\phi_{y,mj})}}{4 [1 - e^{2\pi i(Q_x - Q_y)}]} + O(J_1^2)$$

$$f_{1010,j} = \frac{\sum_{m=1}^M J_{m,1} \sqrt{\beta_{m,x} \beta_{m,y}} e^{i(\Delta\phi_{x,mj} + \Delta\phi_{y,mj})}}{4 [1 - e^{2\pi i(Q_x + Q_y)}]} + O(J_1^2)$$

$$f_{2000,j} = -\frac{\sum_{m=1}^M \beta_{m,x}^{(mod)} \delta K_{m,1} e^{2i\Delta\phi_{x,mj}^{(mod)}}}{1 - e^{4\pi i Q_x^{(mod)}}} + O(\delta K_1^2)$$

$$f_{0020,j} = \frac{\sum_{m=1}^M \beta_{m,y}^{(mod)} \delta K_{m,1} e^{2i\Delta\phi_{y,mj}^{(mod)}}}{1 - e^{4\pi i Q_y^{(mod)}}} + O(\delta K_1^2)$$

$$\beta_m^A \sin(\mathbf{B} \Delta\phi_{mj}) \rightarrow \frac{1}{L_m} \int_0^{L_m} \beta(s)^A \sin(\mathbf{B} \Delta\phi_s) ds$$

$$\beta_m^A \cos(\mathbf{B} \Delta\phi_{mj}) \rightarrow \frac{1}{L_m} \int_0^{L_m} \beta(s)^A \cos(\mathbf{B} \Delta\phi_s) ds$$

Thick-element corrections are being implemented to RDTs

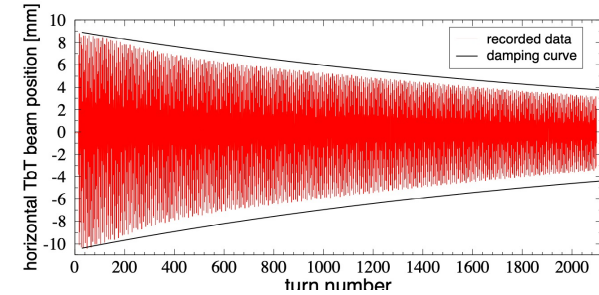
# OUTLOOK: A NEW WAY TO COMPUTE LINEAR RDTS

From analytic formulas

$$f_{2000,j} = -\frac{\sum_{m=1}^M \beta_{m,x}^{(mod)} \delta K_{m,1} e^{2i\Delta\phi_{x,mj}^{(mod)}}}{1 - e^{4\pi i Q_x^{(mod)}}} + O(\delta K_1^2)$$

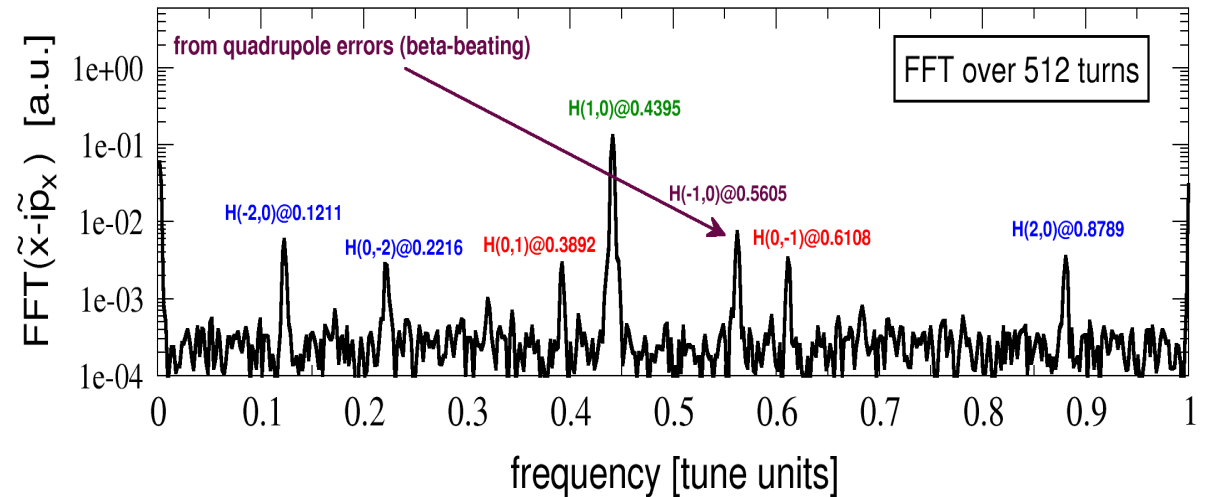
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From particle tracking + harmonic analysis

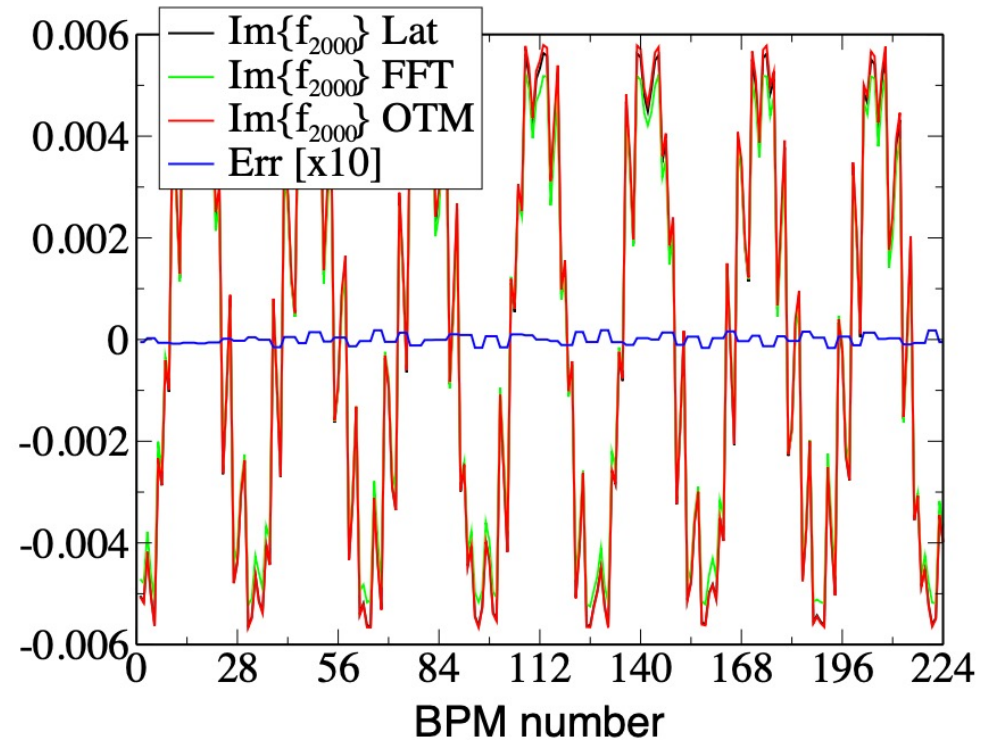
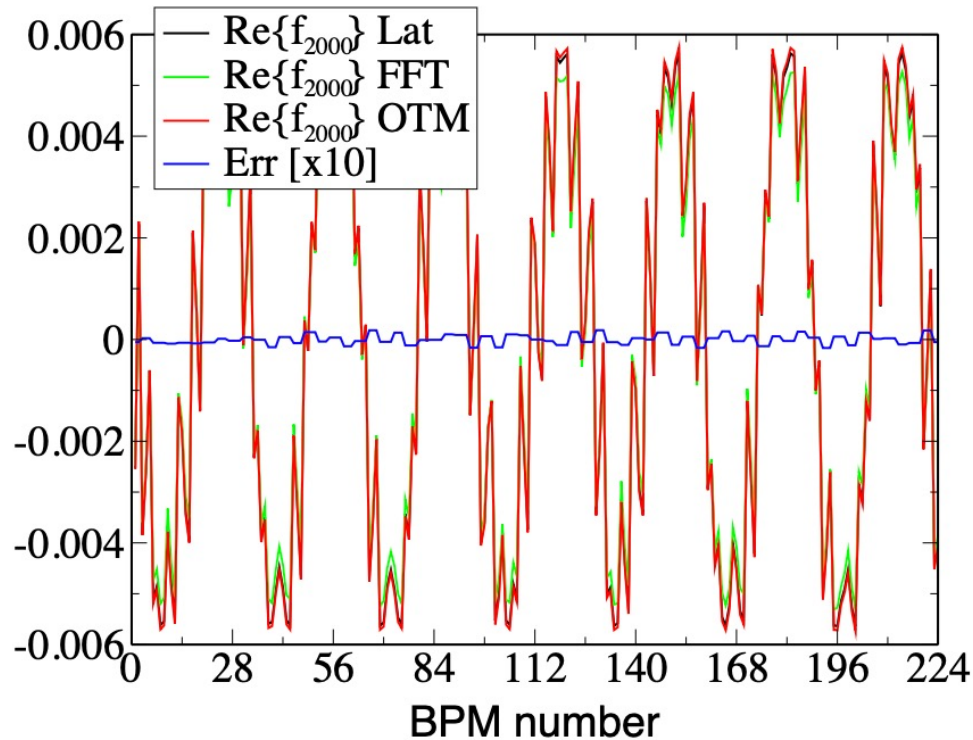


## New formulas from OTM

$$\vec{X}^{(N+1)} = \mathbf{M}\vec{X}^{(N)}, \quad \vec{X} = \begin{pmatrix} x \\ p_x \\ y \\ p_y \end{pmatrix}$$



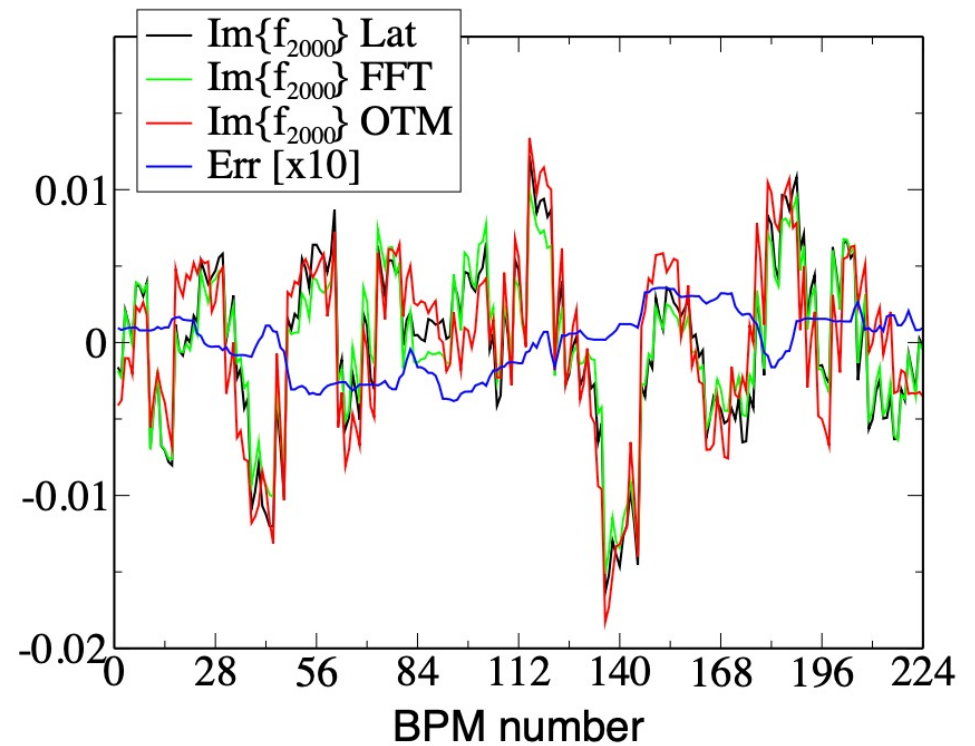
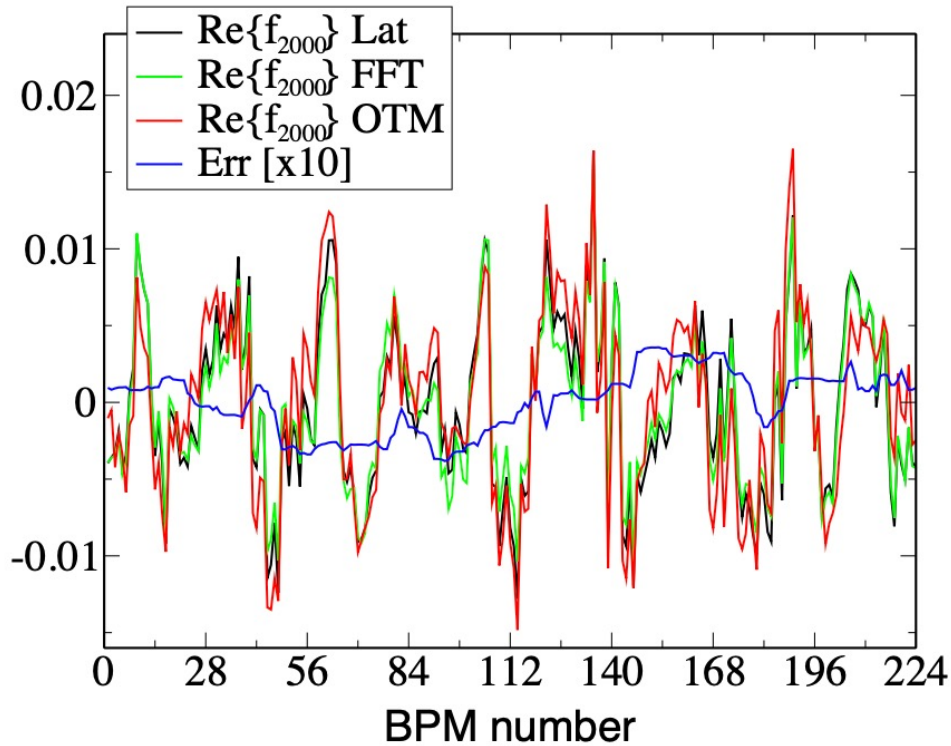
# OUTLOOK: A NEW WAY TO COMPUTE LINEAR RDTS



**Very accurate for single-quad error (thus ok for response matrix)**



# OUTLOOK: A NEW WAY TO COMPUTE LINEAR RDTS



Less accurate for distributed errors (2<sup>nd</sup> order & coupling terms)

## OUTLOOK: A NEW WAY TO COMPUTE LINEAR RDTs

New formulas for coupling RDTs from OTM are still to be derived

$$\vec{X}^{(N+1)} = \mathbf{M}\vec{X}^{(N)}, \quad \vec{X} = \begin{pmatrix} x \\ p_x \\ y \\ p_y \end{pmatrix}$$

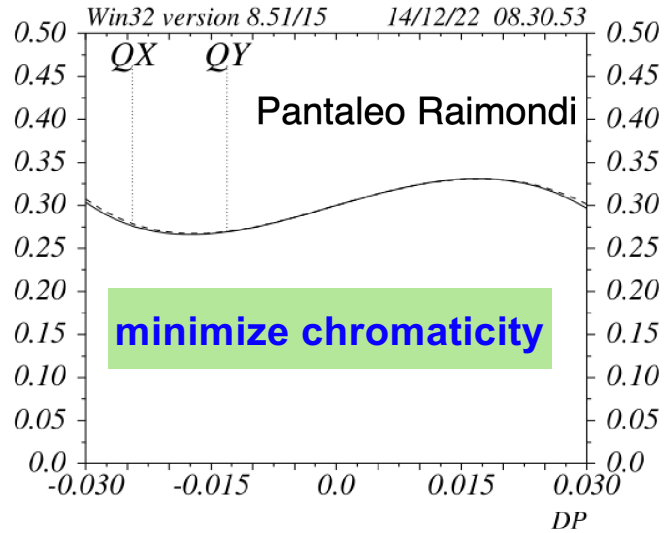
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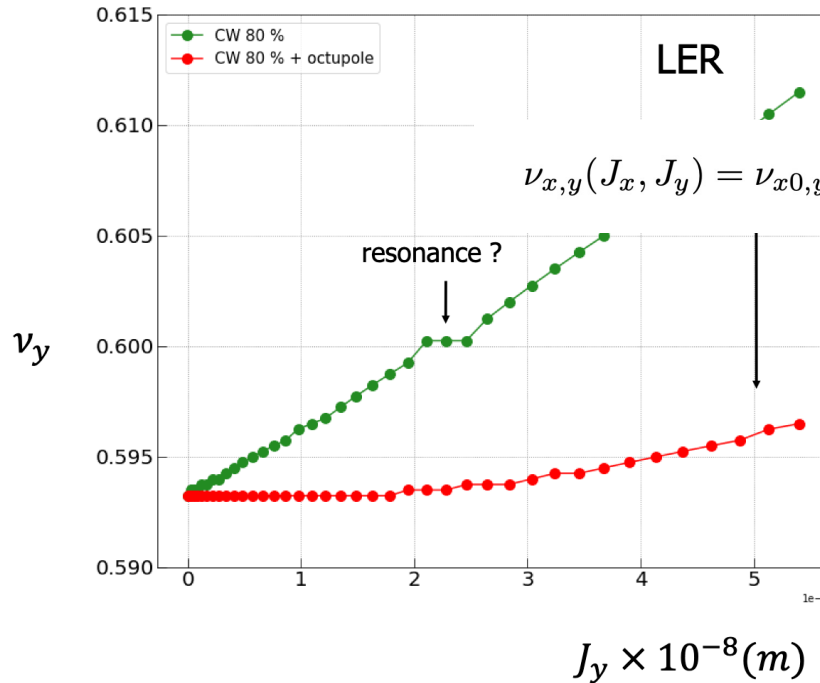
**Thank you!**

# SOME THOUGHTS ON NON-LINEAR OPTIMIZATION

Y. Ohnishi / KEK



The ARC+LSS has zero second order chromaticity in both planes, only third and higher orders remain.



uncorrected

$$\nu_{x,y}(J_x, J_y) = \nu_{x0,y0} + \left( \frac{\partial \nu_{x,y}}{\partial J_x} J_x + \frac{\partial \nu_{x,y}}{\partial J_y} J_y \right) + \dots$$

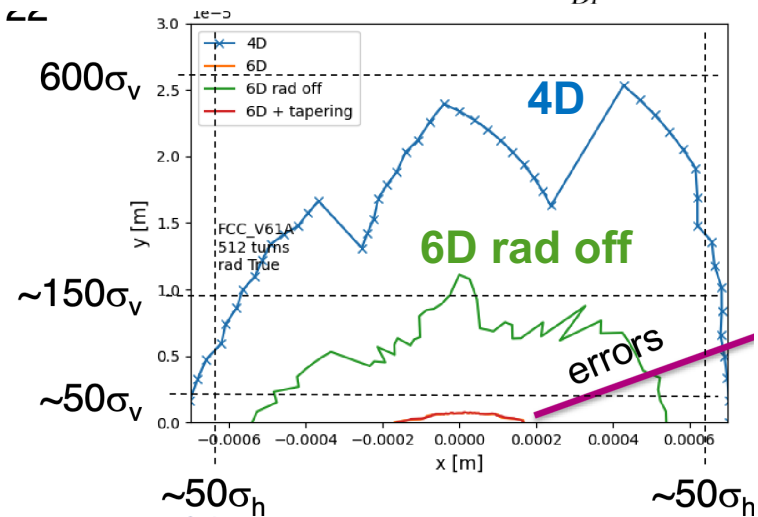
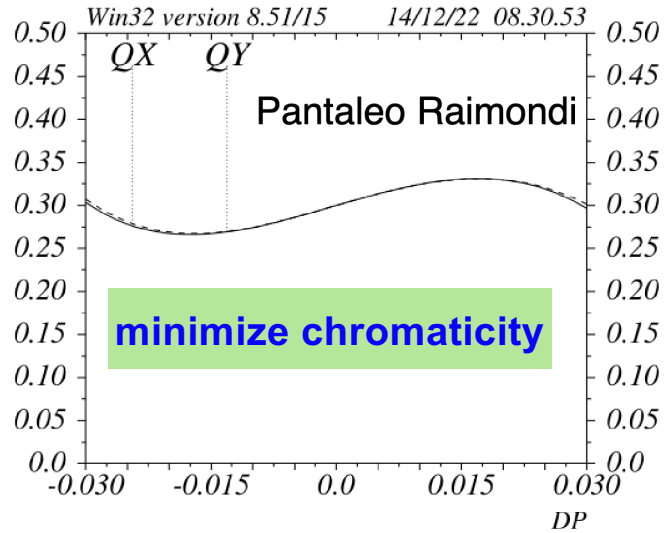
corrected

minimize detuning with amplitude

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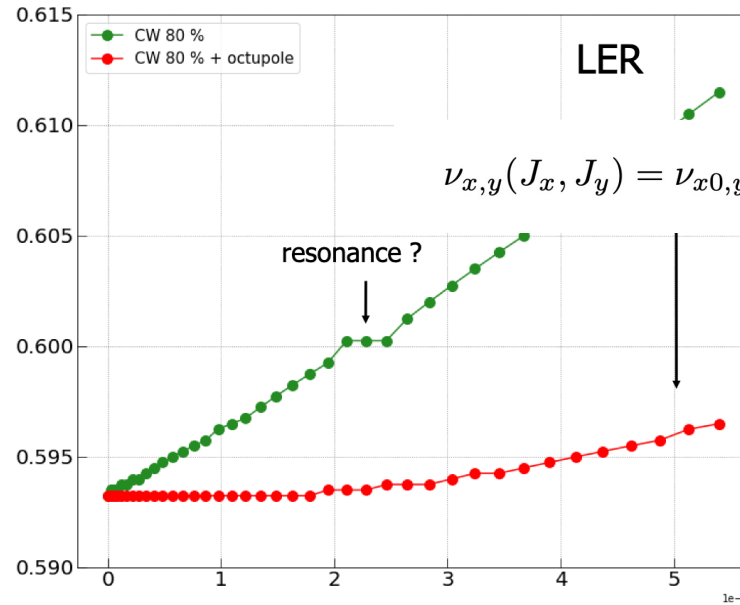
Y. Ohnishi / KEK

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lers | 26th June 2023 | S.Liuzzo et al.

Simone Liuzzo



uncorrected

corrected

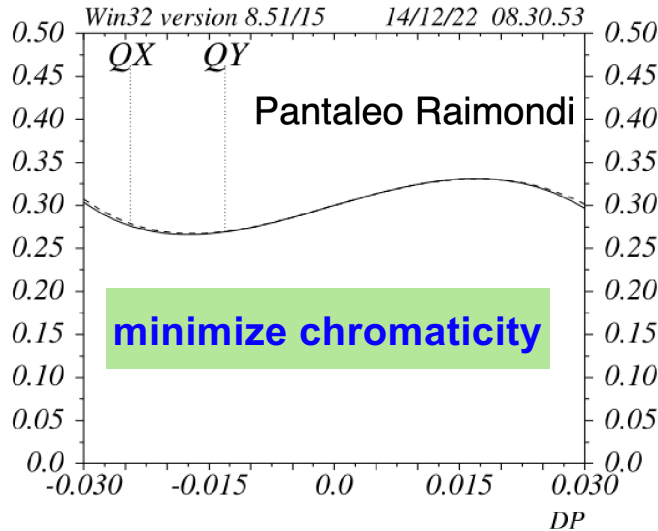
$$\nu_{x,y}(J_x, J_y) = \nu_{x0,y0} + \left( \frac{\partial \nu_{x,y}}{\partial J_x} J_x + \frac{\partial \nu_{x,y}}{\partial J_y} J_y \right) + \dots$$

$J_y \times 10^{-8} (m)$

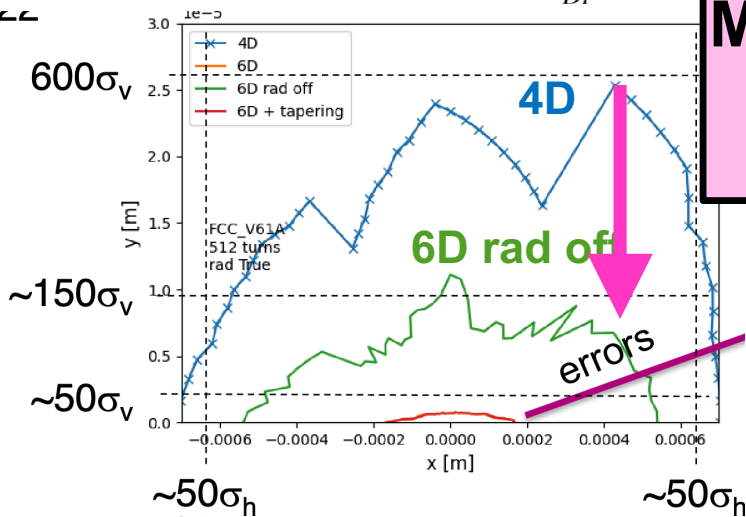
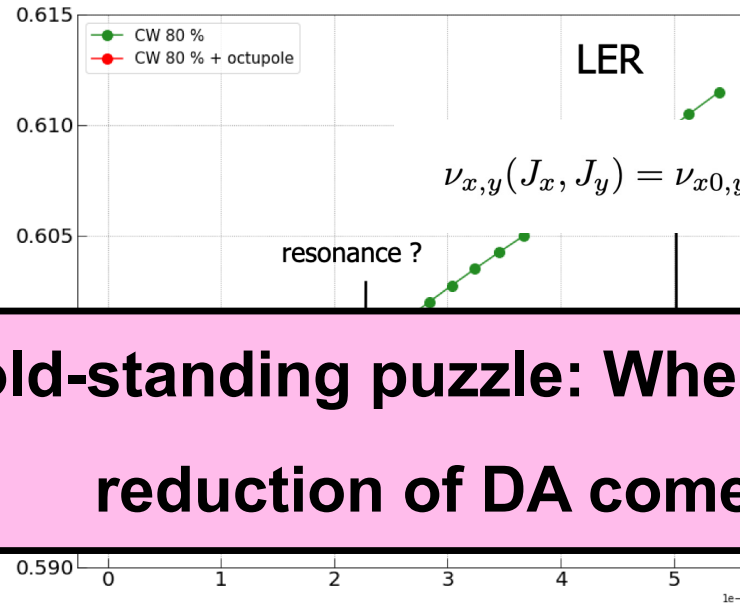
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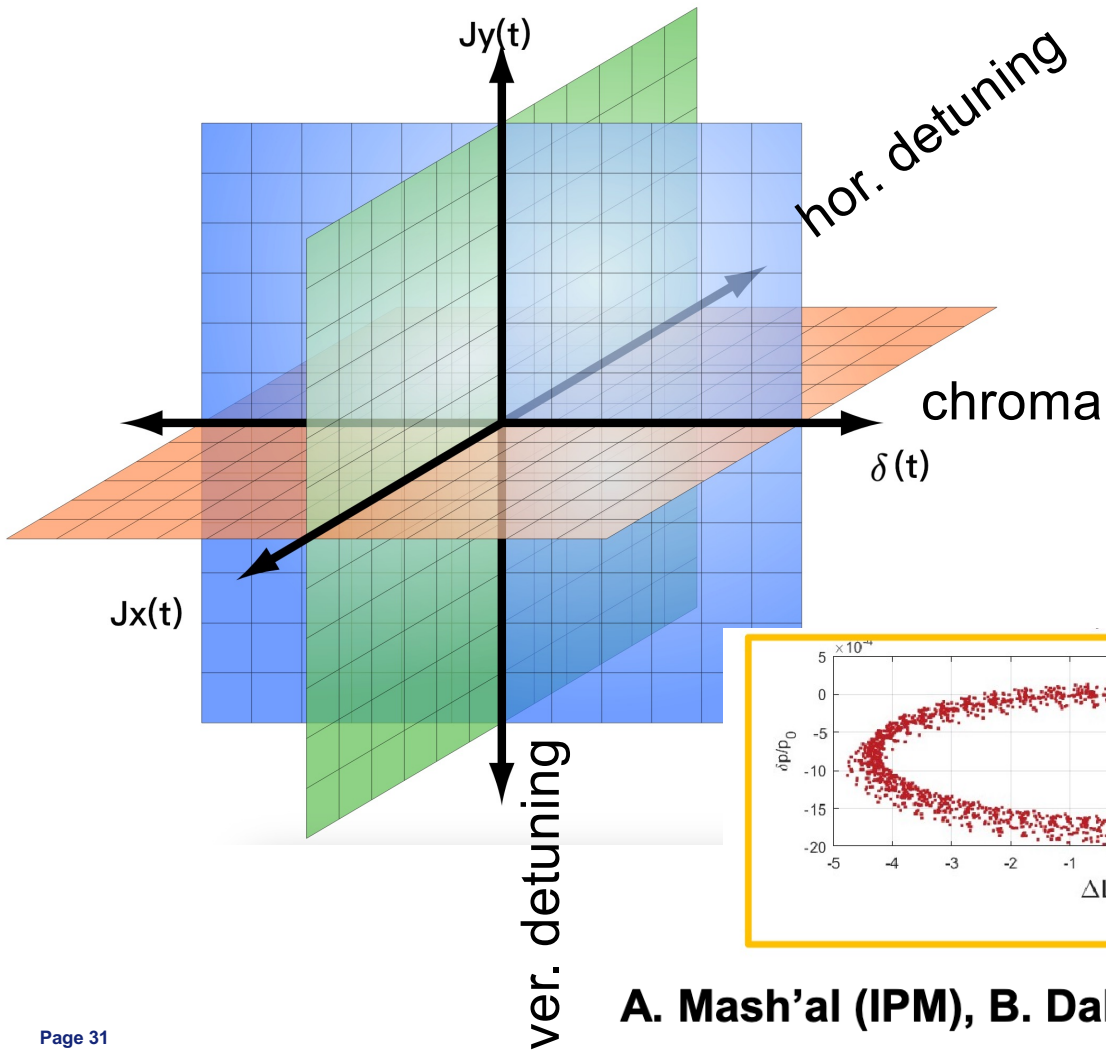


Simone Liuzzo

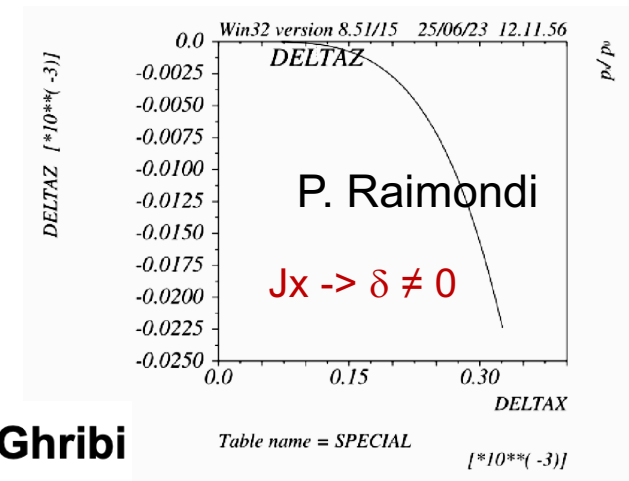
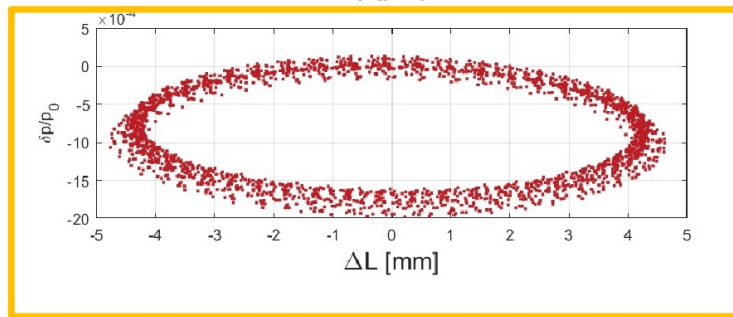
$J_y \times 10^{-8} (m)$

minimize detuning with amplitude

# SOME THOUGHTS ON NON-LINEAR OPTIMIZATION

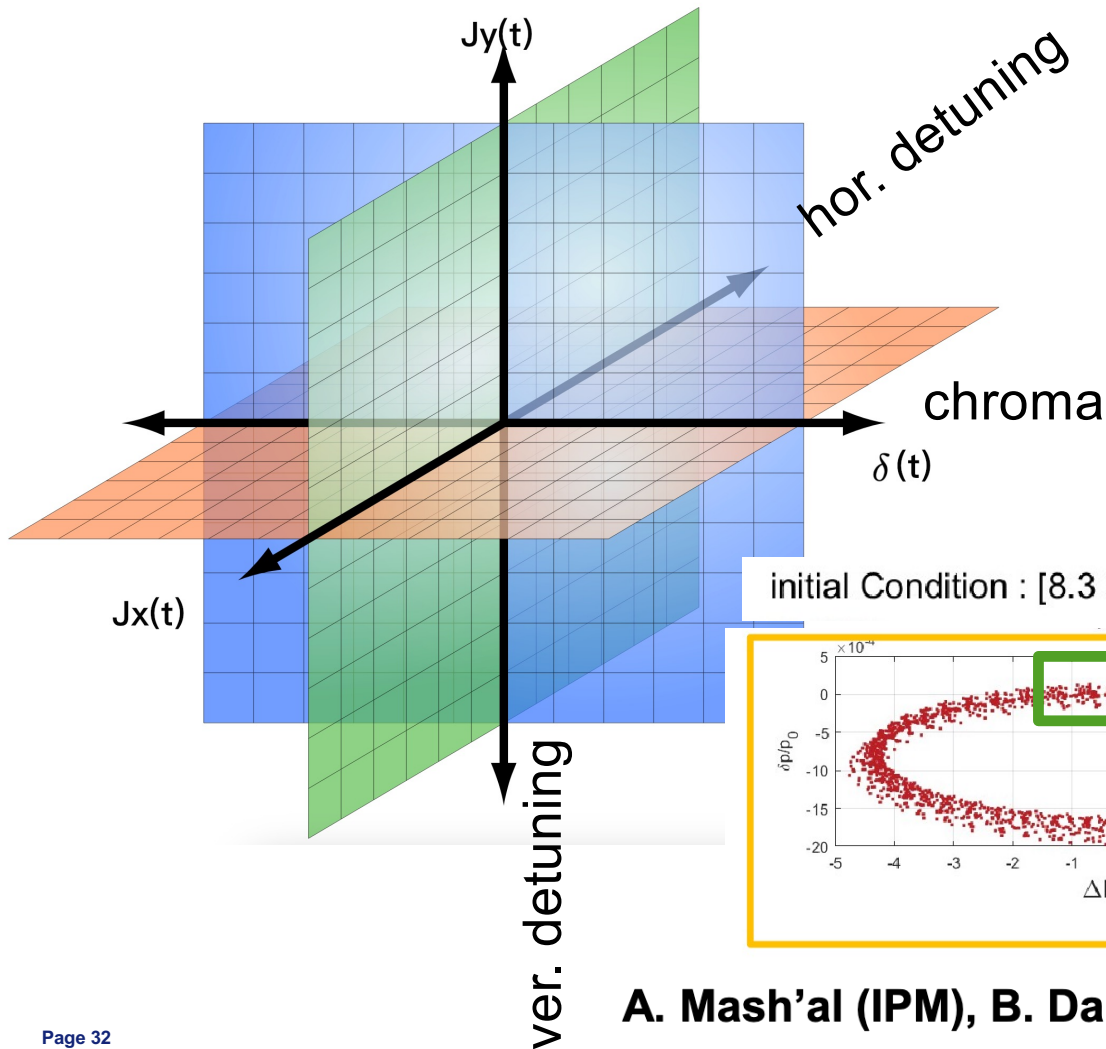


- My guess: Blame path lengthening and synchrotron motion for that.

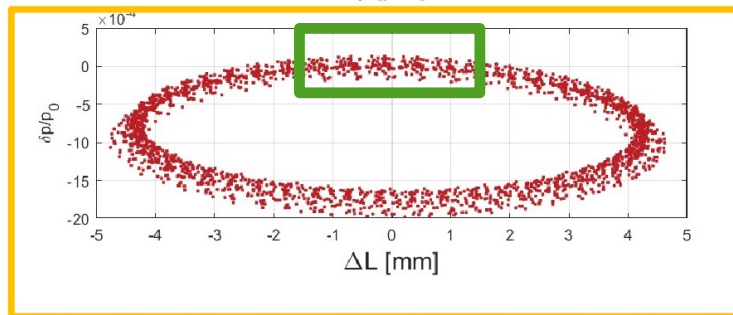


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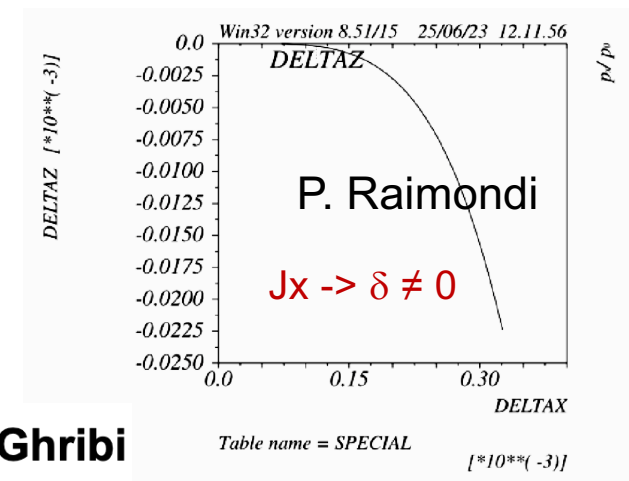
# SOME THOUGHTS ON NON-LINEAR OPTIMIZATION



initial Condition : [8.3 mm, 0, 1  $\mu$ m, 0, 0, 0]



- My guess: Blame path lengthening and synchrotron motion for that.
- Most of the time they will be off-axis ( $J_x \neq 0$ ) and off-energy ( $\delta \neq 0$ ), thus, outside the 3 axis of the tune space.



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# SOME THOUGHTS ON NON-LINEAR OPTIMIZATION

Consequence: Minimizing **cross-term (betatron+chromatic) detuning terms** could be as much, if not more, effective than minimizing **purely higher-order betatron or chromatic terms**

$$Q(J_x(t), \delta(t)) = Q(J_x(0), \delta(0)) +$$

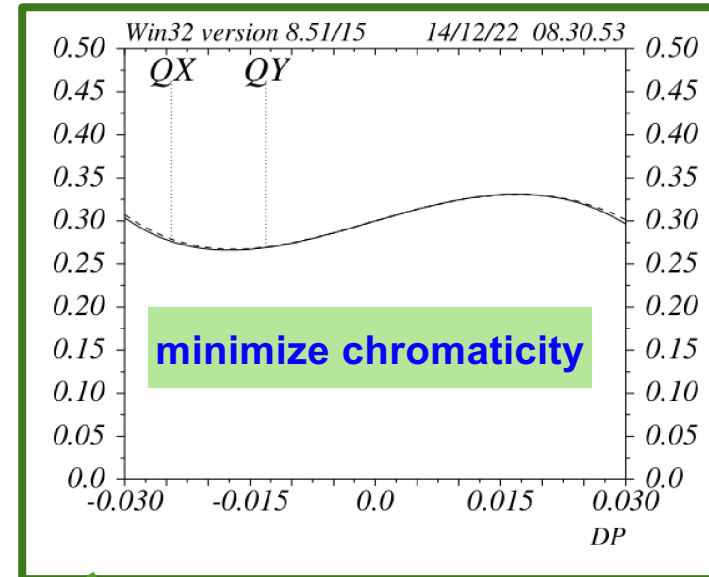
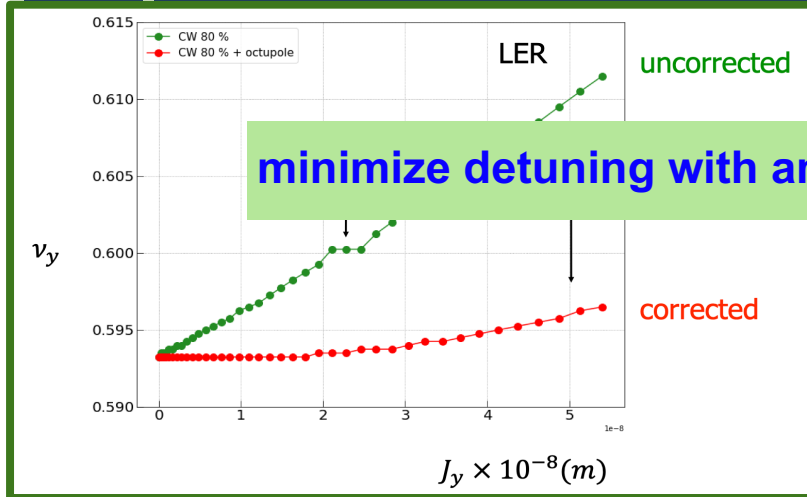
$\left| \begin{array}{l} \text{ANHX, ANHY} \\ n(\varepsilon_1), n(\varepsilon_2), \\ n(\delta_p) \end{array} \right|$

$$\begin{aligned}
 & \frac{\partial Q}{\partial J_x} J_x(t) + \frac{\partial Q}{\partial \delta} \delta(t) + \\
 & \frac{1}{2} \left\{ \frac{\partial^2 Q}{\partial J_x^2} J_x^2(t) + \frac{\partial^2 Q}{\partial \delta^2} \delta^2(t) + \frac{\partial^2 Q}{\partial J_x \partial \delta} J_x(t) \delta(t) \right\} +
 \end{aligned}$$

PTC\_NORMAL,

....

# SOME THOUGHTS ON NON-LINEAR OPTIMIZATION



$$Q(J_x(t), \delta(t)) = Q(J_x(0), \delta(0)) +$$

$$\frac{\partial Q}{\partial J_x} J_x(t) + \frac{\partial Q}{\partial \delta} \delta(t) + \frac{1}{2} \left\{ \frac{\partial^2 Q}{\partial J_x^2} J_x^2(t) + \frac{\partial^2 Q}{\partial \delta^2} \delta^2(t) + \frac{\partial^2 Q}{\partial J_x \partial \delta} J_x(t) \delta(t) \right\} + \dots$$

Maybe we shall also minimize the cross term along with higher-order terms

## SOME THOUGHTS ON NON-LINEAR OPTIMIZATION

Blame path lengthening: **Actually, minimizing linear chroma is still very helpful,**  
since path lengthening and chroma are highly correlated

$$\Delta C = -2\pi(J_X \xi_X + J_Y \xi_Y). \quad (31)$$

$$Q(J_x(t), \delta(t)) = Q(J_x(0), \delta(0)) +$$
$$\frac{\partial Q}{\partial J_x} J_x(t) + \frac{\partial Q}{\partial \delta} \delta(t) +$$

PHYSICAL REVIEW SPECIAL TOPICS - ACCELERATORS AND BEAMS **8**, 094001 (2005)

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**Dependence of average path length betatron motion in a storage ring**

Yoshihiko Shoji\*