

A New Code for Orbit Response Matrix Analysis

Scaled Levenberg-Marquardt Algorithm

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Motivation

Why use **Orbit Response Matrix** ?

- To understand the lattice, how much different from design ?
 - BPM gains, tilt angles.
 - Orbit corrector strength.
 - Quadrupole strength, tilt angles.
- Correct errors of magnets.
- Symmetrize the lattice.

Why a new code ?

- Better convergence property.
- More robust, solve larger models with more parameters.
- Solve the coupling of parameters[1].
- Proven by three different lattices.



Orbit Response Matrix

The Orbit Response Matrix(ORM) is defined by

$$G_{ij} = \frac{dy_i}{d\theta_j^y} \quad y \text{ is horiz. or vert.} \quad (1)$$

- y_i is closed orbit deviation in horizontal or vertical direction at the i^{th} BPM.
- θ_j^y is horizontal or vertical bump angle at the j^{th} corrector.

The full matrix for m BPMs and n corrector is

$$\begin{pmatrix} x \\ z \end{pmatrix}_{2m} = \begin{pmatrix} G^{xx} & G^{xz} \\ G^{zx} & G^{zz} \end{pmatrix}_{2m \times 2n} \begin{pmatrix} \theta^x \\ \theta^z \end{pmatrix}_{2n} \quad (2)$$

where $G_{ij}^{xz} = dx_i/d\theta_j^z$, which is actually a Green function between j^{th} corrector and i^{th} BPM[2].

G^{xz} and G^{zx} will be zero, if no coupling presents.



χ^2 -Minimization

We can model the accelerator by minimizing the χ^2 of our model and experiment measurement*.

$$\begin{aligned} \chi^2(\mathbf{b}) &= \sum_{ij} \left(\frac{G_{ij}(\mathbf{b})^{\{x,z\},\text{model}} - G_{ij}^{\{x,z\},\text{meas}}}{\sigma_{M_{ij}}} \right)^2 + \sum_{x,z} \left(\frac{\nu_{x,z}(\mathbf{b})^{\text{model}} - \nu_{x,z}^{\text{meas}}}{\sigma_{\nu_{x,z}}} \right)^2 \\ &+ \sum_k \left(\frac{\phi_k(\mathbf{b})^{\{x,z\},\text{model}} - \phi_k^{\{x,z\},\text{meas}}}{\sigma_{\phi_{ij}}} \right)^2 \\ &\equiv \sum_i \left(\frac{f_i(\mathbf{b})^{\text{model}} - f_i^{\text{meas}}}{\sigma_{f_i}} \right)^2 \end{aligned} \quad (3)$$

\mathbf{b} is the parameters of our model, which can quadrupole strength, tilt angle and BPM gain, BPM tilt angle.

*ICA measurement can give the phase advance[1]



Steepest-descent(Gradient) method

Follow the inverse-gradient direction: **Increment vector for Steepest-descent method**

$$\delta_s = -\left(\frac{\partial \chi^2}{\partial b_1}, \frac{\partial \chi^2}{\partial b_1}, \dots, \frac{\partial \chi^2}{\partial b_k}\right)^T \quad (4)$$

and the increment of \mathbf{b} in iteration n^{th} step is

$$b_i^{(n+1)} = b_i^{(n)} + (\delta_s)_i \Delta_i \quad (5)$$

Δ_i is the step size for b_i , which can be a constant or adaptive.



Gauss-Newton(Taylor Series) method

$$\chi^2(\mathbf{b}) \equiv \sum_{i=1}^n \left(\frac{f_i^{\text{model}} - f_i^{\text{meas}}}{\sigma_i} \right)^2 \quad (6)$$

f_i can be M_{ij} , ν_x , ν_y , ϕ_x , ϕ_y and other quantities measured or determined by other analysis.

$$f_i \approx f_i^0 + \sum_{j=1}^m \frac{\partial f_i}{\partial b_j} (\delta_{\mathbf{g}})_j \equiv f_i^0 + \sum_{j=1}^m J_{ij} (\delta_{\mathbf{g}})_j \quad (7)$$

Increment vector for Gauss-Newton method

$$\mathbf{J}^T \mathbf{J} \delta_{\mathbf{g}} = \mathbf{J}^T (\mathbf{f}^{\text{meas}} - \mathbf{f}^0) \equiv \mathbf{g} \quad (8)$$



Scaled Levenberg-Marquardt method

Increment vector for scaled Levenberg-Marquardt method[3]

$$(\mathbf{J}^T \mathbf{J} + \lambda \mathbf{D}^T \mathbf{D}) \boldsymbol{\delta} = \mathbf{J}^T (\mathbf{f}^{\text{meas}} - \mathbf{f}^0) = \mathbf{g} \quad (9)$$

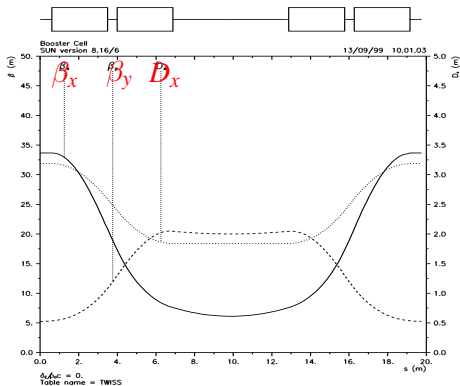
The main improvement is

- ① $\lambda = 0$, it's Gauss-Newton method. $\theta(\boldsymbol{\delta}_{lm}, \boldsymbol{\delta}_g) = 0$
- ② $\lambda \rightarrow \infty$, it's Steepest-descent (Gradient) method. $\theta(\boldsymbol{\delta}_{lm}, \boldsymbol{\delta}_s) = 0$
- ③ λ is an interpolation between Gauss-Newton and Gradient method.
- ④ \mathbf{D} is a diagonal although can be any nonsingular matrix which takes into account the scaling of the problem.
- ⑤ A step bound variable Δ is introduced to limit the length of increment vectors: $\|\mathbf{D}\boldsymbol{\delta}\| \leq \Delta$
- ⑥ The scheme for update Δ keeps the ratio of actual reduction of χ^2 and expected reduction at a reasonable level
- ⑦ More detailed consideration when solve the equation of $\boldsymbol{\delta}$

Strategy of D_k , Δ and λ .



Lattice of Fermi Lab Booster

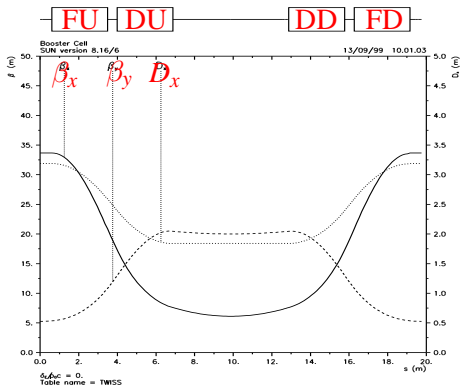


Fermilab Booster (12/05/2002)

- Circum., 474.2 [m]
- Lattice, (FOFODODO)*24
- L(FU,...), 2.8896[m].
- Drift, 6/1.2/0.5 [m]
- β_x , 33.67/6.12 m
- β_y , 20.46/5.27 m
- D_x , 3.19/1.84 m
- ϕ /cell, $\phi_{x,y} = 100.5, 102$
- Tune, $\nu_x = 6.7, \nu_y = 6.8$
- $K_1(F/D)$, 0.0542, $-0.0577 [m^{-2}]$.



Lattice of Fermi Lab Booster

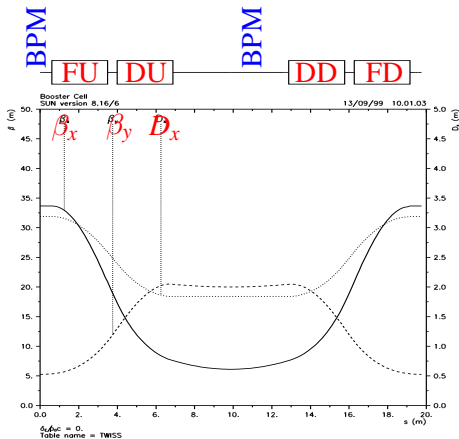


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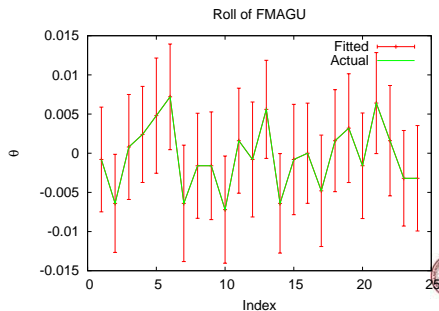
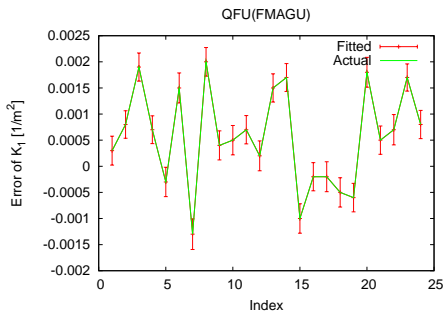
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Fermilab Booster: A model of 192 parameters, Quads. K_1 and θ

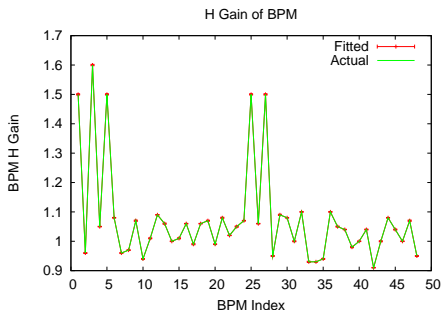
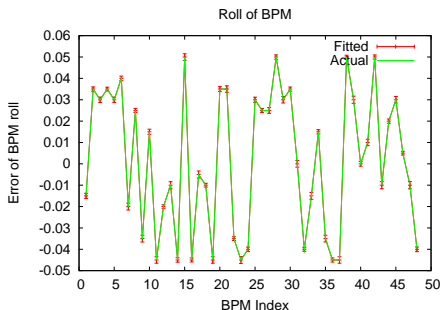
- Only show K_1 and θ_q for FU.
- the other fittings has similar accuracy.
- Probable uncertainties in the estimation of K_1 and θ_q

$$\sigma_{b_k}^2 = \sum_{i=1}^n \left(\frac{\partial b_k}{\partial f_i} \right)^2 \sigma_{f_i}^2 \quad (10)$$



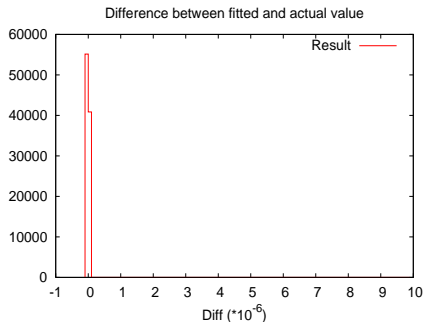
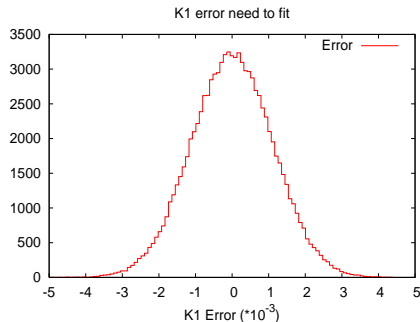
When BPM errors are considered, BPM b_x, b_y, θ_b

- Model has 192+144 parameters, Quads and BPM gain, roll.
- Same accuracy for K_1 and θ_q .



Statistics: Gaussian Distribution

- Model has 96 parameters, i.e. only K_1 errors.
- K_1^{actual} has Gaussian distribution, $\sigma \approx 2\%K_1(F)$
- 1002 random parameter sets fitted.



$$\frac{|K_1^{\text{fit}} - K_1^{\text{actual}}|}{K_1(F)} < 10 \times 10^{-6} / 0.05 = 0.02\%$$



BPM noise

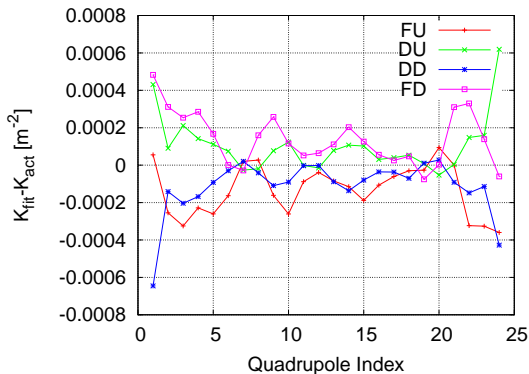


Figure: An example of $\delta_n = 10^{-2}$, $10\mu\text{m}/1\text{mrad}$. The new algorithm breaks the coupling between nearby quadrupoles in [1] and give the right converged solution.



Design Lattice of Taiwan Photon Source

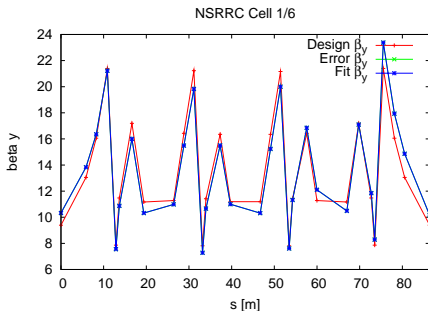


Figure: Restore betatron amplitude function β_x

24 Superperiod, 518.4m. (24p18K1). $\nu_x = 26.226$, $\nu_y = 12.280$. 48 types of quadrupoles, 168 BPMs. 600 parameters: quadrupole gradient and roll, BPM gains and roll. 112,898 terms in χ^2



Source of Vertical Orbit Oscillation

What is the source of vertical orbit oscillation ?

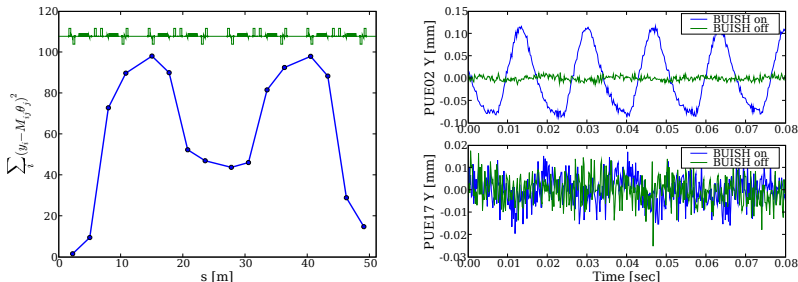


Figure: NSLS VUV ring, vertical orbit oscillation, (FDBK off).

After turning off the BUIISH, the vertical orbit oscillation disappeared.



Conclusion

- 1 Quadrupole and BPM errors are fitted successfully for Fermilab Booster lattice, Taiwan Photon Source design lattice.
- 2 The coupling of parameters for nearby quadrupoles[?] can be resolved by new code.
- 3 New code can fit BPM data with noise.
- 4 It can be used for beam diagnosis.



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