

Generation of Transfer Maps



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Motivation

Goal: identification of magnetic field errors

- compare predicted to measured trajectories
- minimize discrepancy
 - ▣ vary multipole strengths
 - ▣ gradient-based optimization algorithms

Require

tracking differentiable w.r.t. multipole strengths

Hamiltonian Dynamics

Hamiltonian

$$H = c\sqrt{m^2c^2 + \left(\vec{p} - e\vec{A}(\vec{q})\right)^2} + e\phi(\vec{q}) \quad (1)$$

eqn. of motion

$$\dot{q}_i = \frac{\partial H}{\partial p_i} \quad (2)$$

$$\dot{p}_i = -\frac{\partial H}{\partial q_i} \quad (3)$$

transfer map \Leftrightarrow solution of eqn. of motion

Formal Solution of Eqn. of Motion

Poisson bracket

$$\{f, g\} = \sum_{i=1}^n \frac{\partial f}{\partial q_i} \frac{\partial g}{\partial p_i} - \frac{\partial g}{\partial p_i} \frac{\partial f}{\partial q_i} \quad (4)$$

express eqn. of motion via Poisson bracket

$$\vec{z} = [q_1, \dots, q_n, p_1, \dots, p_n]^T \quad (5)$$

$$\dot{\vec{z}} = - \{H, \vec{z}\} = - \underbrace{\{H, \cdot\}}_{\text{diff. operator}} \vec{z} \quad (6)$$

formal solution

$$\vec{z}(t) = e^{-t\{H, \cdot\}} \vec{z}(0) \quad (7)$$

Transfer Map

transfer map \mathcal{M}

$$\vec{z}(t) = \mathcal{M}\vec{z}(0) = e^{-t\{H,\cdot\}}\vec{z}(0) \quad (8)$$

Thin-lens Approximation

- factorize

$$e^{-t\{H,\cdot\}} \approx \prod_i e^{-t c_i \{H_{\text{drift}},\cdot\}} e^{-t d_i \{H_{\text{kick}},\cdot\}}$$

- calculate each factor analytically

TPSA

- truncate to power N

$$e^{-t\{H,\cdot\}} \approx \sum_{k=0}^N \frac{1}{k!} (-t\{H,\cdot\})^k$$

Dual Numbers

- extend real numbers $z_{\text{dual}} = a + \epsilon b$
 - ▣ $a, b \in \mathbb{R}$
 - ▣ special element $\epsilon \neq 0: \epsilon^2 = 0$
- all arithmetic operations defined
 - ▣ $(a + \epsilon b) + (c + \epsilon d) = (a + c) + \epsilon(b + d)$
 - ▣ $(a + \epsilon b) * (c + \epsilon d) = (a + c) + \epsilon(ad + bc)$
 - ▣ etc...
- generalization to hyperdual numbers $\epsilon_1, \epsilon_2, \epsilon_3, \dots$

Dual Numbers: Application

let $f : \mathbb{R} \rightarrow \mathbb{R}$

- using dual numbers yields $f(a)$ and derivative $f'(a)$

$$f(a + \epsilon b) = f(a) + \epsilon f'(a)b \quad (9)$$

- hyperdual numbers: series expansion of f
- works for any computer program

differentiation defined

- $\partial [f(a) + \epsilon f'(a)] = f'(a)$
- implementation of differential operators \Rightarrow Poisson bracket

Generation of Transfer Map

(hyper-) dual numbers

- calculation of

$$\mathcal{M} = e^{-t\{H(\vec{z}), \cdot\}} \vec{z} \approx \sum_{i=1}^N \left(\frac{-t\{H(\vec{z}), \cdot\}}{i!} \right)^i \vec{z} \quad (10)$$

for some numerical \vec{z}

⇒ get \mathcal{M} as a power series expanded around $\vec{z} = 0$

Outlook

- generate thick-lens based transfer map
 - ▣ differentiable w.r.t. to multipole components
 - ▣ enables application of gradient-based optimizers
 - ▣ successfully implemented
- alternative to thin-lens tracking
 - ▣ avoid choice of split scheme
 - ▣ potentially more computationally efficient
 - one map per element vs many drifts / kicks

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