

Mathematical and Numerical Methods
for Non-linear Beam Dynamics in Rings
(an introduction)

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http://cern.ch/Werner.Herr/CAS2013/lectures/Trondheim_methods.pdf

For many more details:

<http://cern.ch/Werner.Herr/METHODS>

Primary purpose of this lectures

- ▣ Assumption: familiar with **linear**, transverse dynamics
 - ▣ Need to introduce new tools for **non-linear** dynamics
 - ▣ Avoid mathematical derivations and proofs rather give "*raison d'etre*" and "*mode d'emploi*"
 - ▣ Give an overview of the *modern*^{*)} tools used in accelerator physics
 - ▣ Necessarily brief and incomplete
- An invitation to further studies ...

^{*)} *modern*: "contemporary", not "fashionable" !



Recommended Bibliography:

- [EF1] E. Forest, *Beam Dynamics - A New Attitude and Framework*, Harwood Academic Publishers, 1998.
- [AC1] A. Chao, *Lecture Notes on Topics in Accelerator Physics* SLAC, 2001.
- [AD] A. Dragt, *Lie Methods for Non-linear Dynamics with Applications to Accelerator Physics*
- [AC2] A. Chao and M. Tigner, *Handbook of Accelerator Physics and Engineering*, World Scientific Publishing, 1998.
- [MB] M. Berz, *Modern Map Methods in Particle Beam Physics*, Academic Press, 1999.
- [AW] A. Wolski, *Lecture Notes on non-linear single particle dynamics*, University of Liverpool.
- [HF] W. Herr and E. Forest, *Non-linear Dynamics in Accelerators*, Landolt-Börnstein, Vol.21C, "Accelerators and Colliders", edited by .S. Myers, H. Schopper.

Why Beam Dynamics in Rings ?

- Most lectures deal with rings
- Rings are periodic systems
- Implies stability (at least for some time) and confinement
- This restricts the methods and tools applicable to study of beam dynamics
- Applicable to other machine and beam lines !



Outline of this lectures

- Motivation, introduction and classical concepts
- New concepts and modern techniques
 - Maps
 - Computation: maps, symplectic integration
 - Hamiltonian theory (for our purpose)
 - Analysis: Lie transforms, normal forms
 - Analysis: Differential algebra
- Identify possible traps and pitfalls ...



Treatment of **LINEAR** dynamics in rings

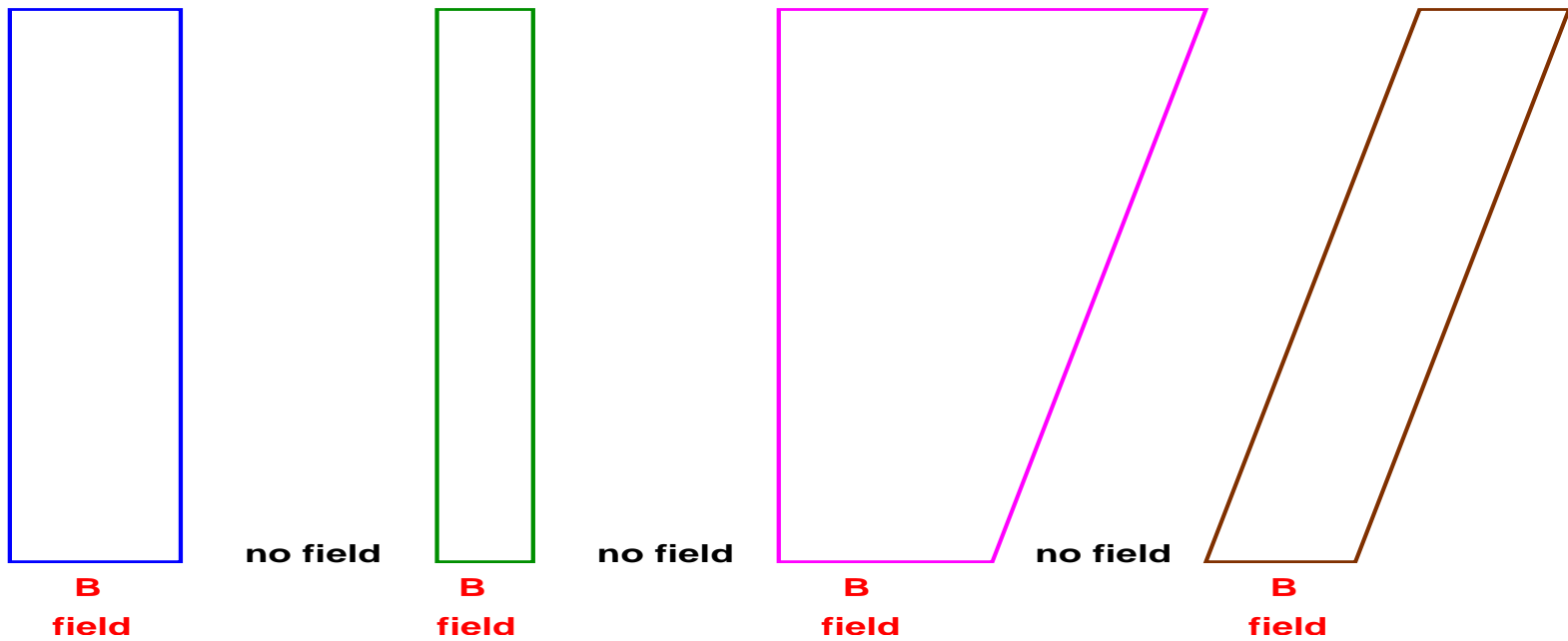
- Standard introduction using Hill's equation
(for simplicity: show for one dimension first):

$$\frac{d^2 x(s)}{ds^2} + K(s)x(s) = 0$$

- $K(s)$ periodic, smooth function
- Is that true ?
- No, normally not



Arrangement of beam line elements



❑ Cannot be described by Hill's equation

❑ Not smooth, not periodic



Treatment of **LINEAR** dynamics in rings

- Used to "derive" Courant-Snyder ansatz:

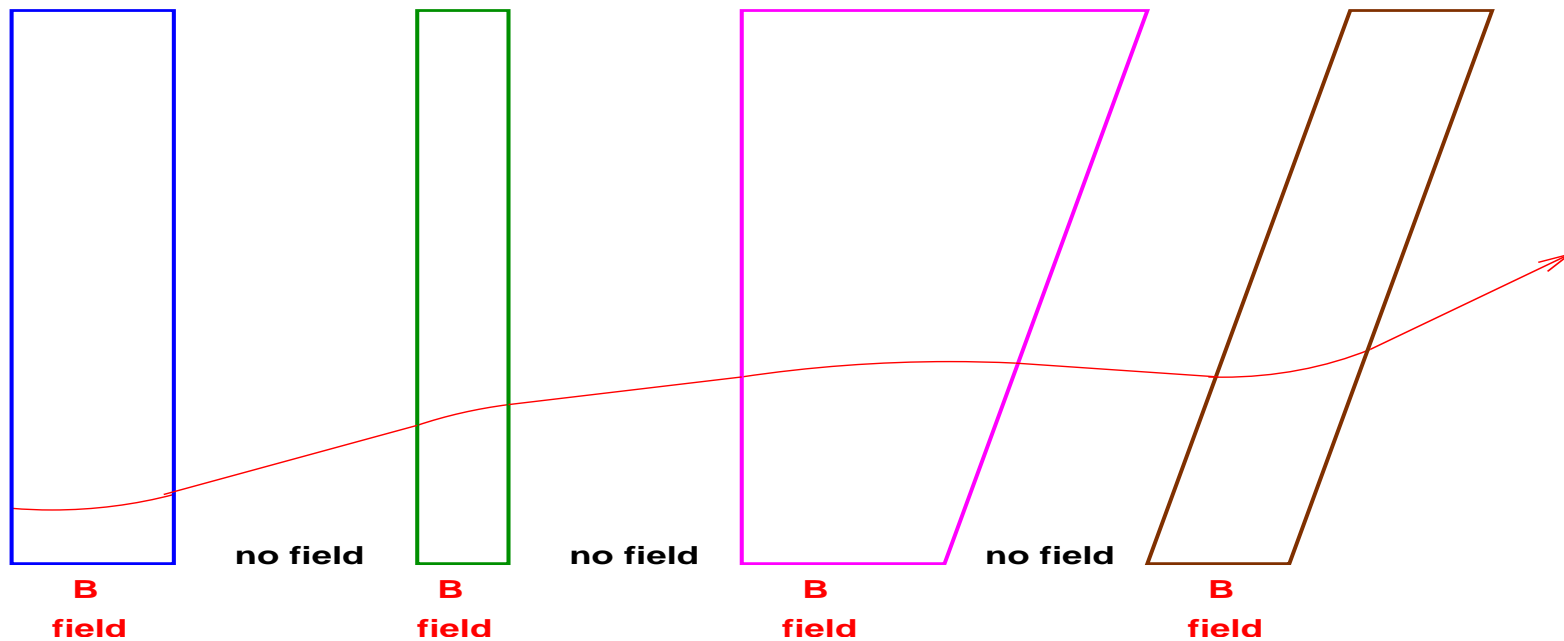
$$x(s) = \sqrt{\beta(s) \cdot \epsilon} \cdot \cos(\mu(s) + \mu_0)$$

$$x'(s) = \sqrt{\frac{\epsilon}{\beta(s)}} \cdot (\sin(\mu(s) + \mu_0) + \alpha \cdot \cos(\mu(s) + \mu_0))$$

- Is the solution to **any** system that is: confined and periodic !
- Do particles really move like this ?



Trajectories in beam line elements



Not a solution of the above

What if we put additional elements (distortions ?)



Treatment of **DISTORTED** dynamics

- Hill's equation with distortions, we have to re-write (similar for the other plane):

$$\frac{d^2 x(s)}{ds^2} + K(s)x(s) = -\frac{B_y(x, y, s)}{p}$$

- or in general as (any order) multipoles:

$$\frac{d^2 x(s)}{ds^2} + K(s)x(s) = \sum_{i,j,k,l \geq 0} p_{ijkl}(s) x^i x'^j y^k y'^l$$

- Very non-linear differential equation to solve ...

➔ Enter the field of non-linear dynamics



Can we deal with that ?

- Under certain circumstances (see lecture by Oliver Brüning):
 - All $p_{ijkl}(s)$ are perturbations, i.e. (very) small
 - Only a few $p_{ijkl}(s)$ are non-zero
 - You can avoid resonances
 - Perturbations are smooth or possibly periodic
 - Perturbation treatment to leading order is sufficient
 - Would you build a 3 billion Euro machine on these assumptions and approximations ?
-

What is normally not said

- Hill's equation, β -function, ...etc.:
 - All concepts developed for **synchrotrons** !
(Courant and Snyder, 1957)
 - Strictly speaking, not applicable to:
 - Beam lines, LINACs, cyclotrons,
 - Computer programs do not use Hill's equation
 - Can we find a better framework ?
-

A disclaimer ...

- Traditional treatment requires many approximations
- Useful to understand and demonstrate concepts
- See Oliver Brüning's, Bernhard Holzer's lectures
- For practical work on realistic machine:
 - New tools required
 - Should exploit modern computing techniques to the maximum
 - It is much easier that you think (.. and other people tell you !)



A better framework

Start with the differential equation:

$$\frac{d^2 x(s)}{ds^2} + K(s)x(s) = \sum_{i,j,k,l \geq 0} p_{ijkl}(s) x^i x'^j y^k y'^l$$



Bad news:

Description not very realistic (see above)

We have no global analytical solution



Good news: An analytical solution is not needed !



A better framework

➤ Why not ?

➤ We do **not** want to know:

➔ The particle's position and momentum at
2h 45min 22.3s ?

(Remember Thermodynamics !)

➤ We do **want** to know:

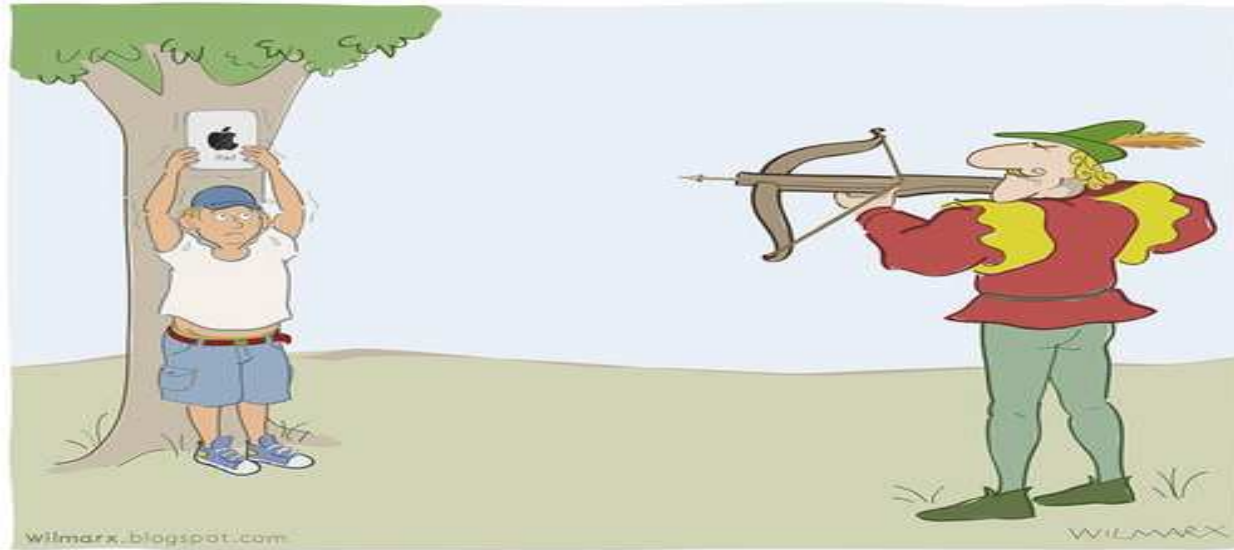
➔ Is the beam stable for a long time ?

➔ Is the motion confined ?

➔ Does the beam hit the target ?



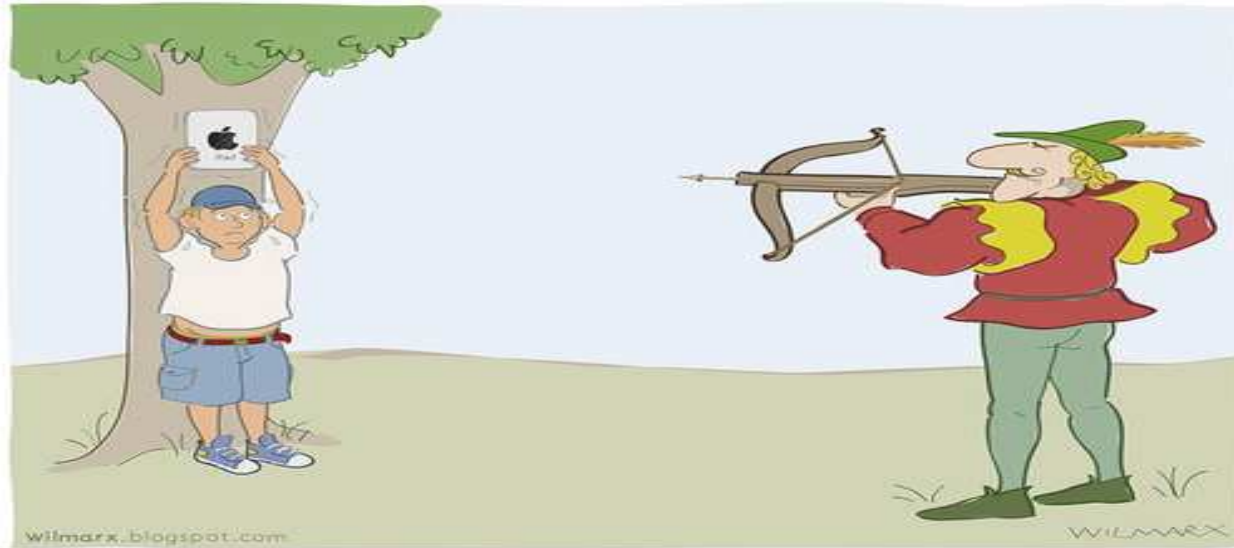
An every day example ...



- Not important to know trajectory as function of time
- Very important to know trajectory at end of flight
- Can we get a framework to get that (easily) ?



An every day example ...



- Not important to know trajectory as function of time
- Very important to know trajectory at end of flight
- Can we get a framework to get that (easily) ?
- ▣ Yes we can ! Should not go back **50** years !

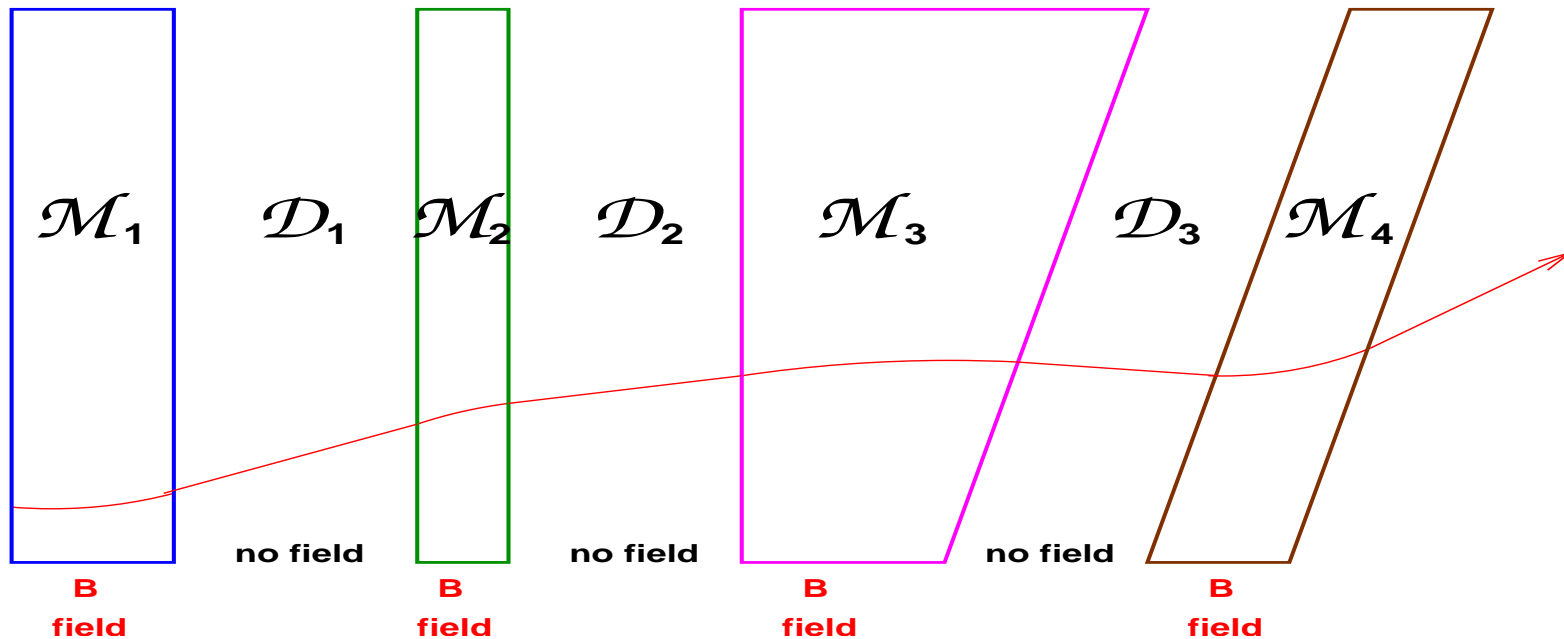
A better framework - go back 100 years ...



- "Old" to "New" classical dynamics:
 - ➔ Topology and properties of phase space (see Oliver's lecture)
 - ➔ Chaotic motion, non-integrable systems
 - ➔ Sensitivity to initial conditions



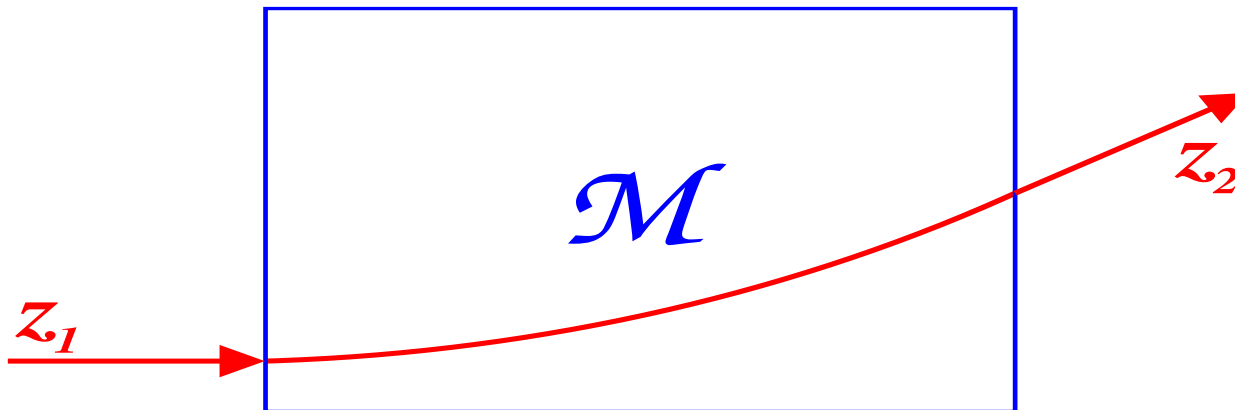
How is an **beam line** described ?



- ▣ Beam line (or ring) made of machine elements and drifts
- ▣ Described by **maps** for magnets (\mathcal{M}) and drifts (\mathcal{D})

How can an **element** really be described ?

- You need to describe what happens to the particle in **M** and in the drifts **D**



- In general: $\vec{z}_2 \neq \vec{z}_1$



How is an **element** described ?

- Let \vec{z}_1, \vec{z}_2 describe a quantity (coordinates, beam sizes ...) before and after the element
- Take an machine element (e.g. magnet) and build a mathematical model \mathcal{M}
 - ➔ In general: $\vec{z}_2 = \mathcal{M}(\vec{z}_1)$
 - ➔ \mathcal{M} is a so-called **map**
 - ➔ Very important: no need to know what happens in the rest of the machine !!
- The complete sequence of MAPS connects the pieces together to make a ring (or beam line)



MAPS transform coordinates through an element

■ Use coordinate vector: $\vec{z} = (x, x' = \frac{\partial x}{\partial s}, y, y' = \frac{\partial y}{\partial s})$ *)

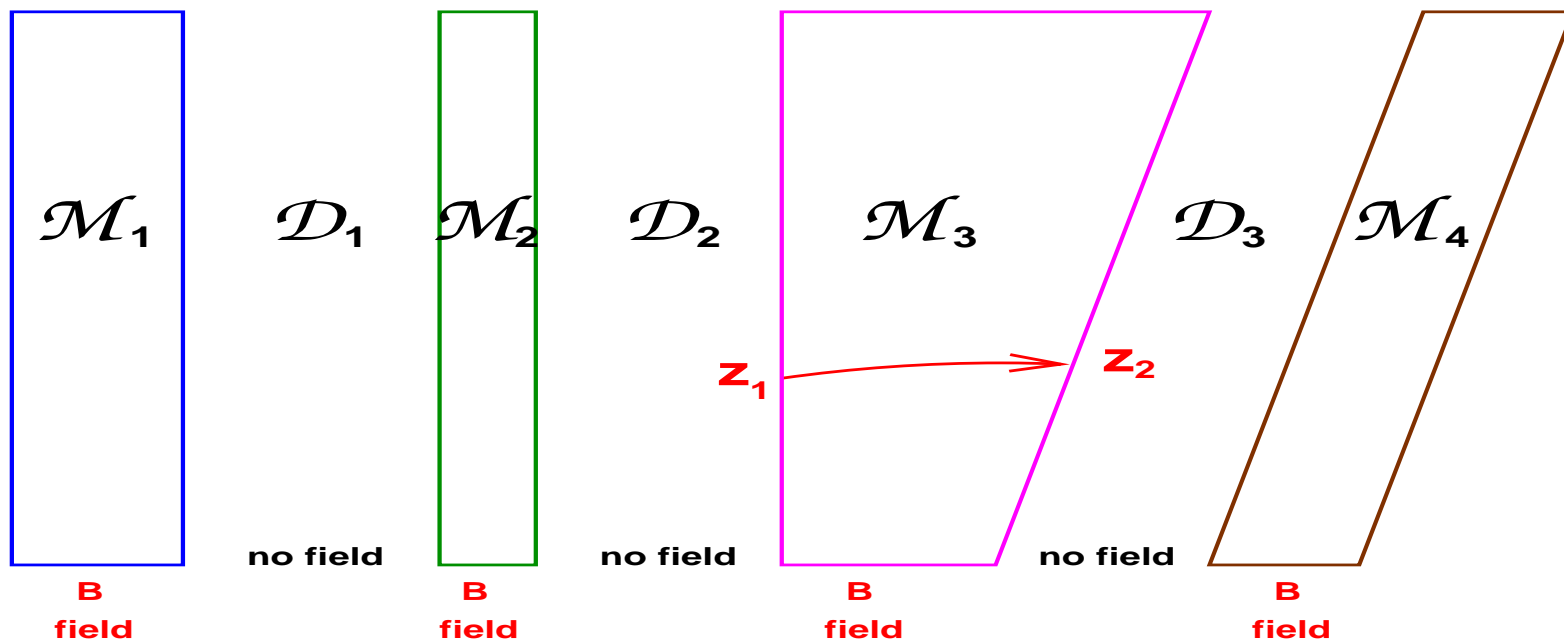
■ \mathcal{M}_3 transforms the coordinates $\vec{z}_1(s_1)$ through the magnet M_3 at position s_1 to new coordinates $\vec{z}_2(s_2)$ at position s_2 :


$$\vec{z}_2(s_2) = \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_2} = \mathcal{M}_3 \circ \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_1} = \mathcal{M}_3 \circ \vec{z}_1(s_1)$$

*) not unique, see later



MAPS transform coordinates through an element



 The MAP fully describes what happens inside the magnet

What can \mathcal{M} be ?

- Any "description" to go from \vec{z}_1 to \vec{z}_2
 - This "description" can be:
 - A simple linear matrix or transformation
 - A non-linear transformation (Taylor series, Lie Transform ...)
 - High order integration algorithm
 - A computer program, subroutine etc.
 - Let us look at linear theory first !
Then generalize to non-linear theory
-

Simple examples (one dimensional)

First a drift space of length L

Two possible descriptions are (there are more):

➤ 1. Go straight from s_1 to s_2 !!

➤ 2. More formal:

$$\begin{pmatrix} x \\ x' \end{pmatrix}_{s_2} = \begin{pmatrix} 1 & L \\ 0 & 1 \end{pmatrix} \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1}$$



Simple examples (one dimensional)


Focusing quadrupole of length L and strength k :

$$\begin{pmatrix} x \\ x' \end{pmatrix}_{s_2} = \begin{pmatrix} \cos(L \cdot k) & \frac{1}{k} \cdot \sin(L \cdot k) \\ -k \cdot \sin(L \cdot k) & \cos(L \cdot k) \end{pmatrix} \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1}$$

Quadrupole with short length L (i.e.: $1 \gg L \cdot k^2$)

$$\begin{pmatrix} x \\ x' \end{pmatrix}_{s_2} = \begin{pmatrix} 1 & 0 \\ -k^2 \cdot L(= \frac{1}{f}) & 1 \end{pmatrix} \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1}$$

They are \mathcal{M} aps, describe the movement in an element (quadrupole)



Interlude: there was already a trap ... !

According to B. Holzer (lectures) or K. Wille (textbook):

$$k = \frac{1}{B\rho} \frac{dB_y}{dx}$$

According to "Handbook for Accelerator Physics" ([AC2]):

$$k^2 = \frac{1}{B\rho} \frac{dB_y}{dx}$$

→ The lesson: check what people use !!

(remember Air Canada 143)



Interlude: it can be worse ... !

You also find (and it may even be useful ...):

$$K^2 = k = \frac{1}{B\rho} \frac{dB_y}{dx}$$

Often different conventions in simulation programs !

Some programs want **fields**, not gradients !

Found this construction:

$$B_y = \frac{1}{0.1} \cdot k \cdot x \cdot B \cdot \rho$$



Interlude: what about 3D ... ?

Formally extended by adding more variables:

→ $(x, x', y, y', \Delta s, \frac{\Delta p}{p})$

→ $\Delta s = c\Delta t$: longitudinal displacement with respect to reference particle

→ $\frac{\Delta p}{p}$: relative momentum difference with respect to reference particle

⚠ Not all programs use this, but rather canonical variables

→ $(x, p_x/p_s, y, p_y/p_s, -c\Delta t, p_t = \frac{\Delta E}{p_s c})$

⚠⚠ p_s may be: $p_s = p_0$ or $p_s = p_0(1 + \delta_s) = m\beta_s\gamma_s$

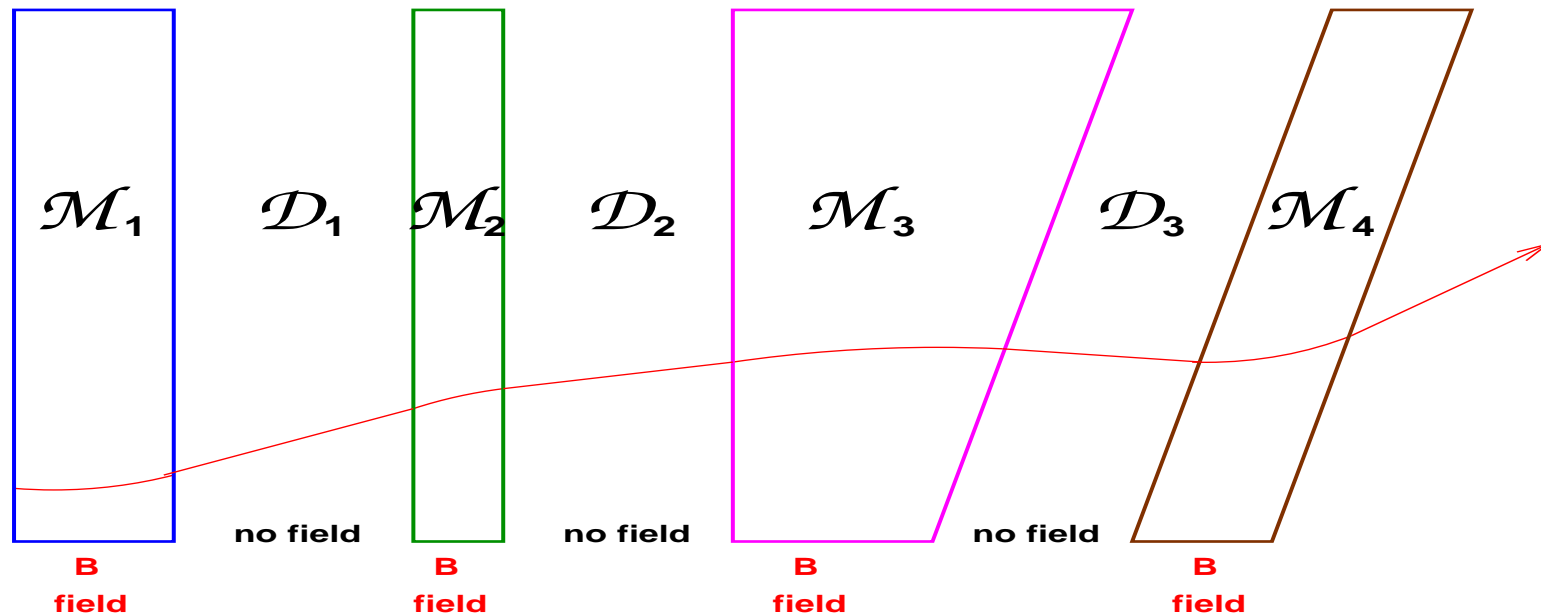
δ_s : difference between reference momentum and design momentum

Putting the "pieces" together

- We have to deal with many elements in our machines
- To make a **ring** or **beam line**:
 - Combine all elements maps together
 - Concatenated maps are a map again
 - Represents a bigger part of the machine
(or the whole machine ...)



How is an **beam line** described ?



$$\mathcal{M}_{\text{all}} = \mathcal{M}_4 \circ \mathcal{D}_3 \circ \mathcal{M}_3 \circ \mathcal{D}_2 \circ \mathcal{M}_2 \circ \mathcal{D}_1 \circ \mathcal{M}_1$$

▣ Beam line (or ring) is combination of all elements

Putting the "pieces" together

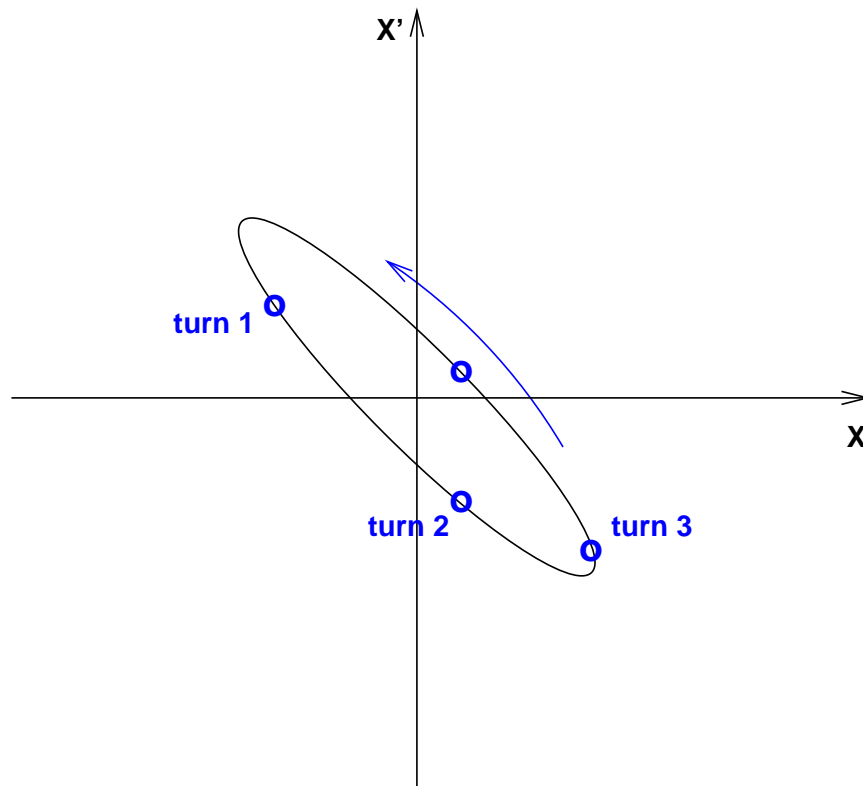
Starting from a position s_0 and applying all maps (for N elements) in sequence around a ring with circumference C to get the **One-Turn-Map** (OTM) for the position s_0 (for one dimension only):

$$\begin{pmatrix} x \\ x' \end{pmatrix}_{s_0 + C} = \mathcal{M}_1 \circ \mathcal{M}_2 \circ \dots \circ \mathcal{M}_N \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_0}$$

$$\Rightarrow \begin{pmatrix} x \\ x' \end{pmatrix}_{s_0 + C} = \mathcal{M}_{ring}(s_0) \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_0}$$



What does \mathcal{M}_{ring} do ?



■ Transforms coordinates in phase space once per turn

Analysis of the One-Turn-Map

- We have obtained a map for the whole ring
- In simplest (linear) case: multiply matrices to get a One-Turn-**Matrix**
- Have to get now the information we want:
 - Optics parameters (Tune, Twiss functions, ..)
 - Closed orbit
 - Stability
 - etc. ...
- How to analyse a MAP (first: a matrix) ???
(see also B. Holzer lecture, but practice comes here)



Normal forms

- Maps can be transformed into (Jordan) Normal Forms
- Original maps and normal form are equivalent, but ...
- Easily used to analyse the maps:
 - Get parameters (Q , Q' , Twiss function, ..)
 - Study invariants, etc.
 - Stability
 - For resonance analysis
 - etc. ...
- Idea is to make a transformation to get a simpler form for the map



Normal forms

Assume the map \mathcal{M}_{12} propagates the variables from location 1 to location 2, we try to find transformations $\mathcal{A}_1, \mathcal{A}_2$ such that:

$$\mathcal{A}_1 \mathcal{M}_{12} \mathcal{A}_2^{-1} = \mathcal{R}_{12}$$

■ The map \mathcal{R}_{12} is:

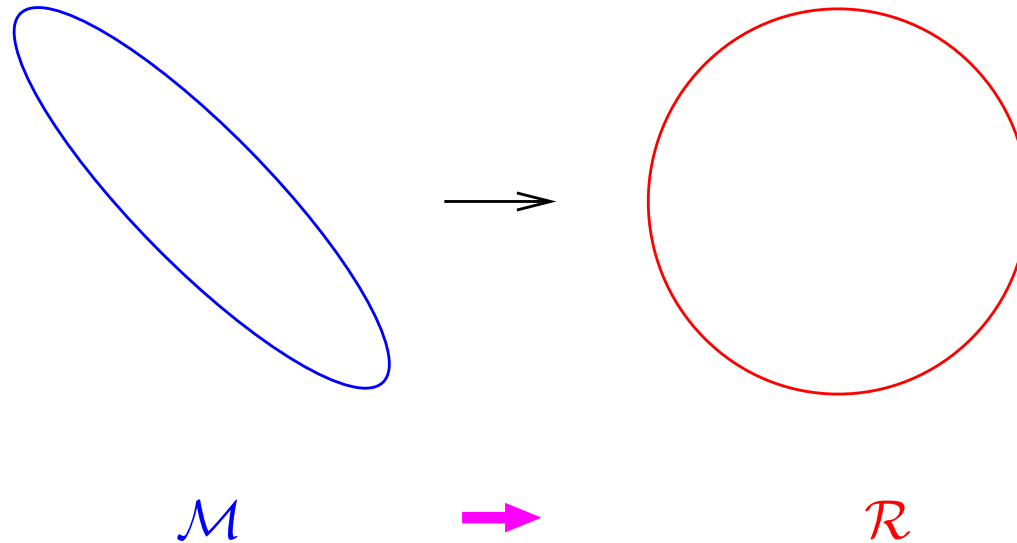
- A "Jordan Normal Form", (or at least a very simplified form of the map)
- Example: \mathcal{R}_{12} becomes a pure rotation

■ The map \mathcal{R}_{12} describes the same dynamics as \mathcal{M}_{12} , but:

- All coordinates are transformed
- The transformations $\mathcal{A}_1, \mathcal{A}_2$ "analyse" the motion



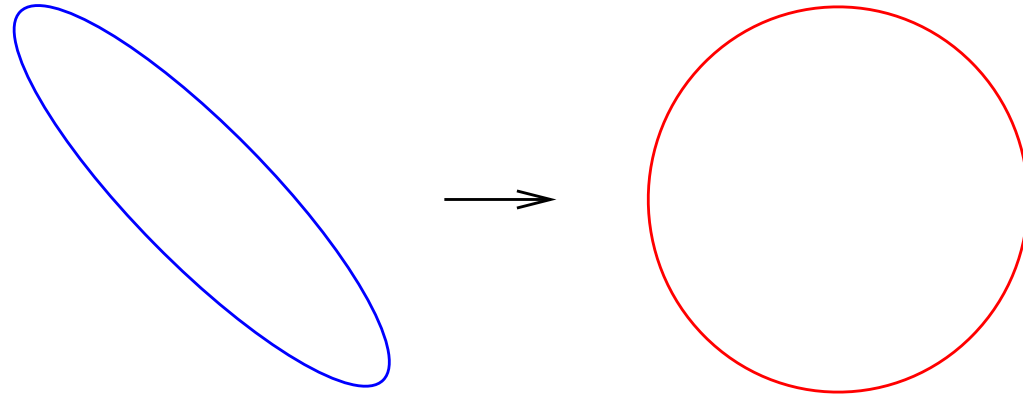
Normal forms - linear case



- Pictorial form of the transformation
- Motion on a complicated ellipse becomes motion on a circle (i.e. a pure rotation)



Normal forms - linear case



$$M = \mathcal{A} \circ \mathcal{R}(\Delta\mu) \circ \mathcal{A}^{-1} \quad \text{or :} \quad \mathcal{R}(\Delta\mu) = \mathcal{A}^{-1} \circ M \circ \mathcal{A}$$



Normal forms - linear case (1D)

Assume the one-turn-map (here a matrix) $\mathcal{M}(s)$ at the position s is (e.g. lecture on transverse dynamics):

$$\mathcal{M}(s) = \begin{pmatrix} \cos(\Delta\mu) + \alpha(s)\sin(\Delta\mu) & \beta(s)\sin(\Delta\mu) \\ -\gamma(s)\sin(\Delta\mu) & \cos(\Delta\mu) - \alpha(s)\sin(\Delta\mu) \end{pmatrix}$$

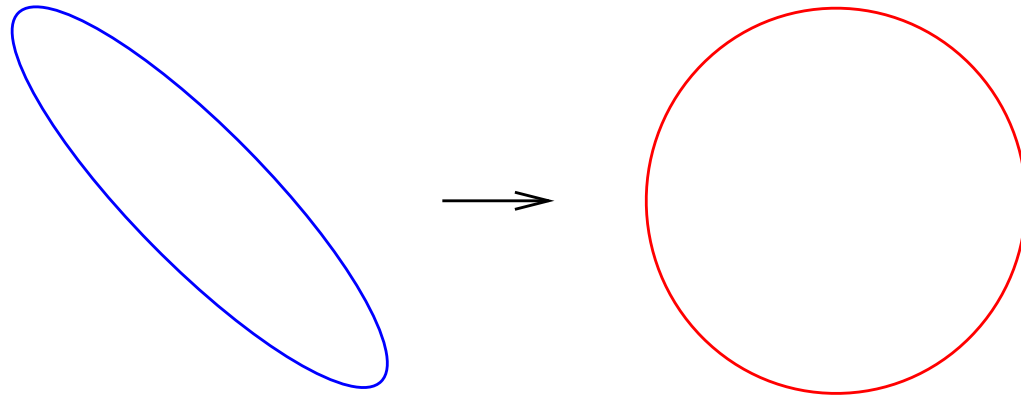
- Describes the motion on a phase space ellipse
- Re-write \mathcal{M} such that one part \mathcal{R} becomes a pure rotation (a circle), i.e.:

$$\mathcal{A}\mathcal{R}\mathcal{A}^{-1} = \mathcal{M}$$

- How ? Remember lectures on Linear Algebra (Eigenvectors, Eigenvalues ...)



Normal forms - linear case (1D)



$$M = \mathcal{A} \circ \mathcal{R}(\Delta\mu) \circ \mathcal{A}^{-1} \quad \text{or :} \quad \mathcal{R}(\Delta\mu) = \mathcal{A}^{-1} \circ M \circ \mathcal{A}$$

with

$$\mathcal{A} = \begin{pmatrix} \sqrt{\beta(s)} & 0 \\ -\frac{\alpha}{\sqrt{\beta}} & \frac{1}{\sqrt{\beta(s)}} \end{pmatrix} \quad \text{and} \quad \mathcal{R} = \begin{pmatrix} \cos(\Delta\mu) & \sin(\Delta\mu) \\ -\sin(\Delta\mu) & \cos(\Delta\mu) \end{pmatrix}$$



Normal forms - linear case (1D)

We had:

$$M = \mathcal{A} \circ \mathcal{R}(\Delta\mu) \circ \mathcal{A}^{-1} \quad \text{or :} \quad \mathcal{R}(\Delta\mu) = \mathcal{A}^{-1} \circ M \circ \mathcal{A}$$

with

$$\mathcal{A} = \begin{pmatrix} \sqrt{\beta(s)} & 0 \\ -\frac{\alpha}{\sqrt{\beta}} & \frac{1}{\sqrt{\beta(s)}} \end{pmatrix} \quad \text{and} \quad \mathcal{R} = \begin{pmatrix} \cos(\Delta\mu) & \sin(\Delta\mu) \\ -\sin(\Delta\mu) & \cos(\Delta\mu) \end{pmatrix}$$

- This is just the Courant-Snyder transformation to get β, α, \dots etc., $\Delta\mu$ is the tune !
- That is: the Courant-Snyder analysis is just a **normal form transform** of the linear one turn matrix
- Works in more than one dimension



Normalized variables:

Please note that:

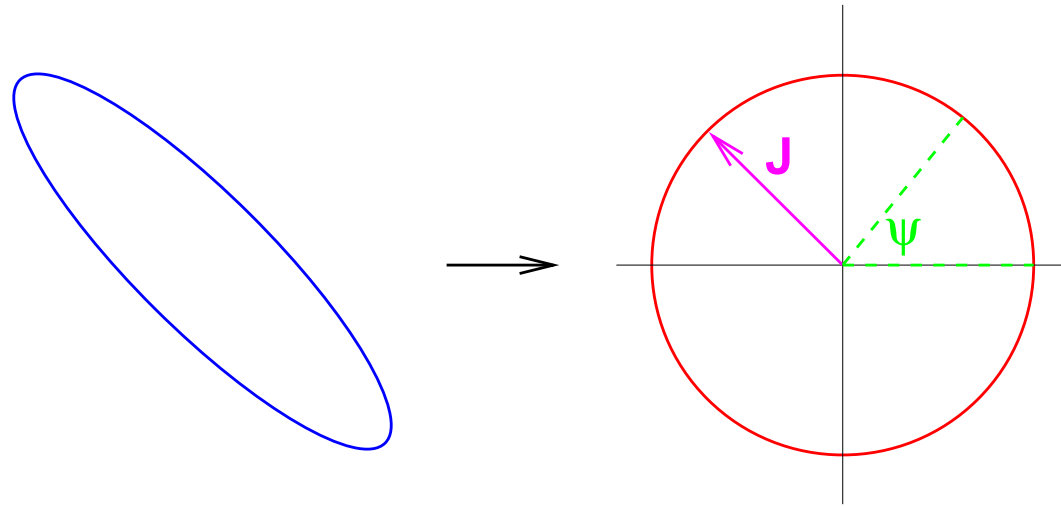
$$\begin{pmatrix} x_n \\ x'_n \end{pmatrix} = \mathcal{A}^{-1} \circ \begin{pmatrix} x \\ x' \end{pmatrix}$$

is just a variable transformation to new, normalized variables.

- Tune $(\Delta\mu)$ in the normalized map, stability for real values of phase advance $(\Delta\mu)$
- Optical functions (β, α, \dots) in the normalizing map
- No need to make any assumptions, ansatz, approximation, ...



Interlude: action - angle variables



■ Once the particles "travel" on a circle:

- Radius (say: $\sqrt{2J}$, with $J = \frac{x_n^2 + x_n'^2}{2}$) is constant (invariant of motion): action J
- Phase advances by constant amount: angle Ψ



Another example: coupling (2D)

Assume a one-turn-matrix in 2D:

$$T = \begin{pmatrix} M & n \\ m & N \end{pmatrix}$$

M, m, N, n are 2-by-2 matrices. In case of coupling: $m \neq 0, n \neq 0$.
we can try to re-write as:

$$T = \begin{pmatrix} M & n \\ m & N \end{pmatrix} = V R V^{-1}$$

with:

$$R = \begin{pmatrix} A & 0 \\ 0 & B \end{pmatrix} \quad \text{and} \quad V = \begin{pmatrix} \gamma I & C \\ -C^t & \gamma I \end{pmatrix}$$



What have we obtained ?

■ The matrix R is our simple rotation:

➤ A and B are the one-turn-matrices for the "normal modes"

➤ The matrix C contains the "coupling coefficients"

➤ The matrix V transforms from the coordinates (x, x', y, y') into the "normal mode" coordinates (w, w', v, v') via the expression:

$$(x, x', y, y') = V(w, w', v, v')$$

■ The last 2 slides: normally 1 hour lecture



Normal forms - linear case

This is extremely useful when map is applied k times (e.g. k turns):

$$M^k(x, x') = AR^k A^{-1}(x, x') = AR^k(X, X')$$

- For multi-turns: study effect of map in normalized coordinates
- Multiplying a matrix k (e.g. 4x4) can be quite a job !
- Easier to apply k times using the simple map (e.g. a rotation of μ becomes just a rotation $k \cdot \mu$)
- The A just transforms back to physical coordinates at the end (once !)



The general philosophy (linear systems):

- Describe your elements by a **linear** map
- Combine all maps into a ring or beam line to get the **linear** one turn matrix
- Normal form analysis of the **linear** one turn matrix will give all the information

No need for any assumptions !

No need for any approximations !

Works in more than 1D and with coupling !



The general philosophy (non-linear systems):

- Describe your elements by a **non-linear** map
- Combine all maps into a ring or beam line to get the **non-linear** one turn map
- Normal form analysis of the **non-linear** one turn map will give all the information

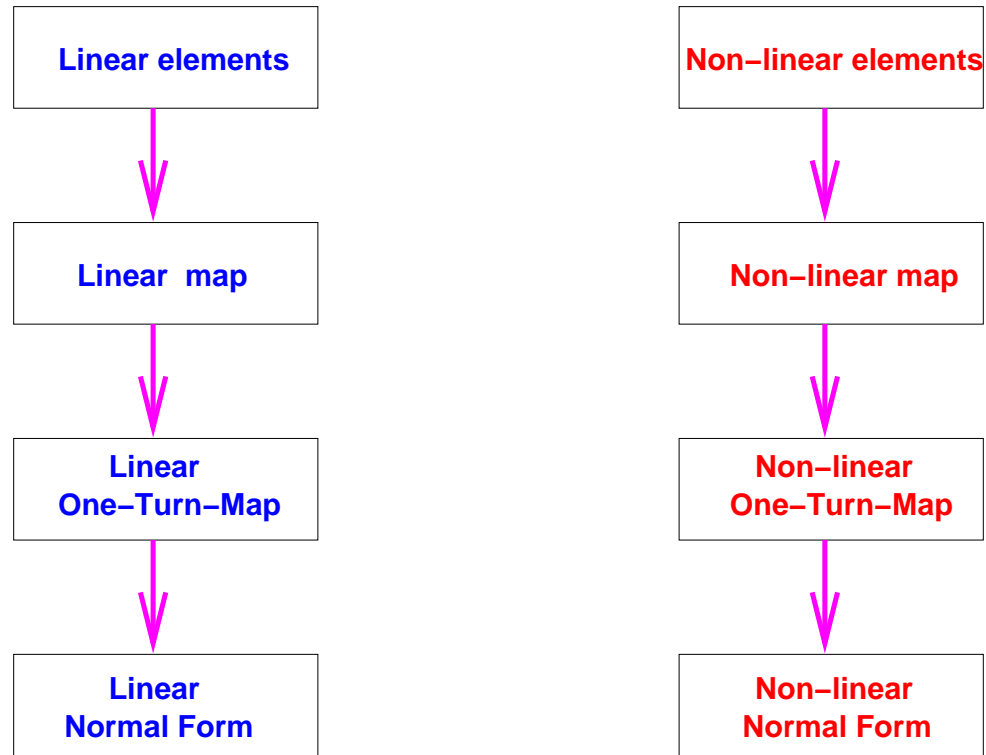
No need for any assumptions !

No need for any approximations !

Works in more than 1D and non-linearities !



The general philosophy



General formalism for all cases !



A small complication ...



Non-linear maps are not matrices !

Various types of non-linear MAPS

■ Choice depends on the application

➤ Taylor maps

➤ Symplectic integration techniques

➤ Lie transformations

➤ Truncated power series algebra (TPSA), can also generate Taylor map from tracking

➤ ...



(A key concept: Symplecticity)

▣ Not all possible maps are allowed !

▣ Requires for a matrix $\mathcal{M} \rightarrow \mathcal{M}^T \cdot S \cdot \mathcal{M} = S$

with:

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

▣ It basically means: \mathcal{M} is area preserving and

$$\lim_{n \rightarrow \infty} \mathcal{M}^n = \text{finite} \quad \implies \quad \det \mathcal{M} = 1$$



Introducing non-linear elements

Effect of a (short) quadrupole depends **linearly** on amplitude (re-written from the matrix form):

$$\vec{z}(s_2) = \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_2} = \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_1} + \begin{pmatrix} 0 \\ k_1 \cdot x_{s_1} \\ 0 \\ k_1 \cdot y_{s_1} \end{pmatrix}$$

→ $\vec{z}(s_2) = \mathbf{M} \cdot \vec{z}(s_1)$

→ \mathbf{M} is a matrix



Non-linear elements (e.g. sextupole)

Effect of a (thin) sextupole with strength k_2 is:

$$\vec{z}(s_2) = \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_2} = \begin{pmatrix} x \\ x' \\ y \\ y' \end{pmatrix}_{s_1} + \begin{pmatrix} 0 \\ \frac{1}{2}k_2 \cdot (x_{s_1}^2 - y_{s_1}^2) \\ 0 \\ k_2 \cdot (x_{s_1} \cdot y_{s_1}) \end{pmatrix}$$

→ $\vec{z}(s_2) = \mathcal{M} \circ \vec{z}(s_1)$

→ \mathcal{M} is **not** a matrix, i.e. cannot be expressed by matrix multiplication



Non-linear elements

Cannot be written in linear matrix form !

We need something like:

$$\begin{aligned} z_1(s_2) = x(s_2) = & R_{11} \cdot x + R_{12} \cdot x' + R_{13} \cdot y + \dots \\ & + T_{111} \cdot x^2 + T_{112} \cdot xx' + T_{122} \cdot x'^2 + \\ & + T_{113} \cdot xy + T_{114} \cdot xy' + \dots \\ & + U_{1111} \cdot x^3 + U_{1112} \cdot x^2x' + \dots \end{aligned}$$

and the equivalent for all other variables ...



Higher order (Taylor -) MAPS:

We have (for: $j = 1\dots 4$):

$$z_j(s_2) = \sum_{k=1}^4 R_{jk} z_k(s_1) + \sum_{k=1}^4 \sum_{l=1}^4 T_{jkl} z_k(s_1) z_l(s_1)$$

Let's call it : $\mathcal{A}_2 = [R, T]$ (second order map \mathcal{A}_2)

Higher orders can be defined as needed ...

$$\mathcal{A}_3 = [R, T, U] \implies + \sum_{k=1}^4 \sum_{l=1}^4 \sum_{m=1}^4 U_{jklm} z_k(s_1) z_l(s_1) z_m(s_1)$$



Higher order (Taylor -) MAPS:

Example: complete second order map for a (thick) sextupole with length L and strength K (in 4D):

$$\begin{aligned}x_2 &= x_1 + Lx'_1 - K \left(\frac{L^2}{4}(x_1^2 - y_1^2) + \frac{L^3}{6}(x_1x'_1 - y_1y'_1) + \frac{L^4}{24}(x_1'^2 - y_1'^2) \right) \\x'_2 &= x'_1 - K \left(\frac{L}{2}(x_1^2 - y_1^2) + \frac{L^2}{2}(x_1x'_1 - y_1y'_1) + \frac{L^3}{6}(x_1'^2 - y_1'^2) \right) \\y_2 &= y_1 + Ly'_1 + K \left(\frac{L^2}{4}x_1y_1 + \frac{L^3}{6}(x_1y'_1 + y_1x'_1) + \frac{L^4}{24}(x'_1y'_1) \right) \\y'_2 &= y'_1 + K \left(\frac{L}{2}x_1y_1 + \frac{L^2}{2}(x_1y'_1 + y_1x'_1) + \frac{L^3}{6}(x'_1y'_1) \right)\end{aligned}$$

 Definition of K not unique, can differ by some factor !!

$$\text{e.g.} \quad \left(\frac{\partial^2 x}{\partial t^2} = S \cdot x^2 \quad \text{versus} \quad \frac{\partial^2 x}{\partial t^2} = \frac{k}{2} \cdot x^2 \right)$$


Symplecticity for higher order MAPS

- Truncated Taylor expansions are not matrices !!
- It is the associated Jacobian matrix \mathcal{J} which must fulfil the symplecticity condition:

$$\mathcal{J}_{ik} = \frac{\partial z_2^i}{\partial z_1^k} \quad \left(\text{e.g. } \mathcal{J}_{xy} = \frac{\partial z_2^x}{\partial z_1^y} \right)$$

$$\mathcal{J} \text{ must fulfil: } \mathcal{J}^t \cdot \mathcal{S} \cdot \mathcal{J} = \mathcal{S}$$

- In general: $\mathcal{J}_{ik} \neq \text{const}$ \rightarrow for truncated Taylor map can be difficult to fulfil for all z



Symplecticity for higher order MAPS

Take the sextupole map (for simplicity in one dimension):

$$\begin{aligned}x_2 &= x_1 + Lx'_1 - K \left(\frac{L^2}{4} x_1^2 + \frac{L^3}{6} x_1 x'_1 + \frac{L^4}{24} x_1'^2 + \mathcal{O}(3) \right) \\x'_2 &= x'_1 - K \left(\frac{L}{2} x_1^2 + \frac{L^2}{2} x_1 x'_1 + \frac{L^3}{6} x_1'^2 + \mathcal{O}(3) \right)\end{aligned}$$

we compute:

$$\mathcal{J}^T \cdot S \cdot \mathcal{J} = \begin{pmatrix} 0 & 1 + \Delta S \\ -1 - \Delta S & 0 \end{pmatrix} \neq S$$

is non-symplectic with error:

$$\Delta S = \frac{K^2}{72} L^4 (L^2 x'^2 + 6Lxx' + 6x^2)$$



Symplecticity for higher order MAPS

Take the sextupole map (for simplicity in one dimension):

$$\begin{aligned}x_2 &= x_1 + Lx'_1 - K \left(\frac{L^2}{4} x_1^2 + \frac{L^3}{6} x_1 x'_1 + \frac{L^4}{24} x_1'^2 + \mathcal{O}(3) \right) \\x'_2 &= x'_1 - K \left(\frac{L}{2} x_1^2 + \frac{L^2}{2} x_1 x'_1 + \frac{L^3}{6} x_1'^2 + \mathcal{O}(3) \right)\end{aligned}$$

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The way out: thin magnets

- Real magnets have a finite length, i.e. thick magnets
- Thick magnet: field and length used to compute effect, i.e. the map
- Consequence: they are not always linear elements (even not dipoles, quadrupoles)
- For thick, non-linear magnets closed solution for maps often does not exist

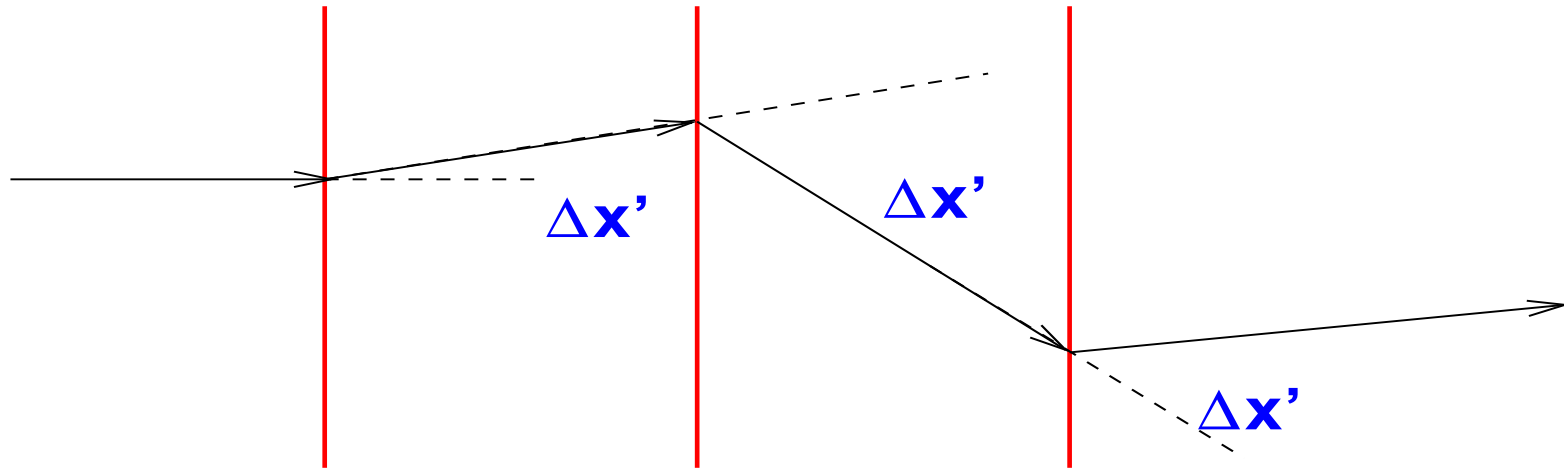


Thick versus thin magnets

- Thin "magnet": let the length go to zero, but keep field integral finite (constant)
- Thin dipoles and quadrupoles are linear elements
- Thin elements are much easier to use ...



Moving through thin elements



■ No change of amplitudes x and y

■ The momenta x' and y' receive an amplitude dependent deflection (**kick**)

→ $x' \rightarrow x' + \Delta x'$ and $y' \rightarrow y' + \Delta y'$

Using thin elements

- Can we approximate a thick element by thin element(s) ?
 - Yes, when the length is small or does not matter
 - Yes, when we can model the thick magnet correctly
 - What about accuracy, symplecticity etc. ??
 - Demonstrate with some simple examples



Thick → thin quadrupole

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} \cos(L \cdot K) & \frac{1}{K} \cdot \sin(L \cdot K) \\ -K \cdot \sin(L \cdot K) & \cos(L \cdot K) \end{pmatrix}$$

- Exact map (matrix) for quadrupole
- What happens when we make it thin ?
 - Accuracy ?
 - Symplecticity ?
- (What follows is valid for all elements)



Accuracy of thin lenses

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} \cos(L \cdot K) & \frac{1}{K} \cdot \sin(L \cdot K) \\ -K \cdot \sin(L \cdot K) & \cos(L \cdot K) \end{pmatrix}$$

- Start with exact map
- Taylor expansion in "small" length L :

$$L^0 \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + L^1 \cdot \begin{pmatrix} 0 & 1 \\ -K^2 & 0 \end{pmatrix} + L^2 \cdot \begin{pmatrix} -\frac{K^2}{2} & 0 \\ 0 & -\frac{K^2}{2} \end{pmatrix} + \dots$$

Accuracy of thin lenses (B)

- Keep up to first order term in L

$$\mathcal{M}_{s \rightarrow s+L} = L^0 \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + L^1 \cdot \begin{pmatrix} 0 & 1 \\ -K^2 & 0 \end{pmatrix}$$

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} 1 & L \\ -K^2 \cdot L & 1 \end{pmatrix} + \mathcal{O}(L^2)$$

- Precise to first order $\mathcal{O}(L^1)$
- $\det \mathcal{M} \neq 1$, non-symplectic

Accuracy of thin lenses (C)

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} 1 & L \\ -K^2 \cdot L & 1 \end{pmatrix} + \mathcal{O}(L^2)$$



$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} 1 & L \\ -K^2 \cdot L & 1 - K^2 L^2 \end{pmatrix}$$

- Precise to first order $\mathcal{O}(L^1)$
- "symplectified" by adding term $-K^2 L^2$
(it is wrong to $\mathcal{O}(L^2)$ anyway ...)

Accuracy of thin lenses

- Keep up to second order term in L

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} 1 - \frac{1}{2}K^2L^2 & L \\ -K^2 \cdot L & 1 - \frac{1}{2}K^2L^2 \end{pmatrix} + \mathcal{O}(L^3)$$

- Precise to second order $\mathcal{O}(L^2)$
- More accurate than (C), but not symplectic



Accuracy of thin lenses (D)

➤ Symplectification like:

$$\mathcal{M}_{s \rightarrow s+L} = \begin{pmatrix} 1 - \frac{1}{2}K^2L^2 & L - \frac{1}{4}K^2L^3 \\ -K^2 \cdot L & 1 - \frac{1}{2}K^2L^2 \end{pmatrix} + \mathcal{O}(L^3)$$

➤ Precise to second order $\mathcal{O}(L^2)$

➤ Fully symplectic

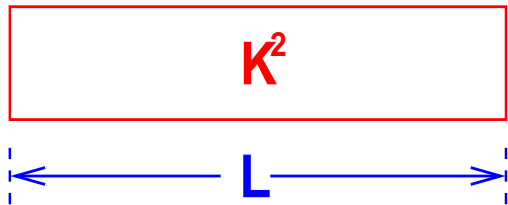


Accuracy of thin lenses

- Looks like we made some arbitrary changes and called it "symplectification"
- Is there a physical picture behind the approximations ?
- Yes, **geometry** of thin lens kicks ...
- A thick element is split into thin elements with drifts between them

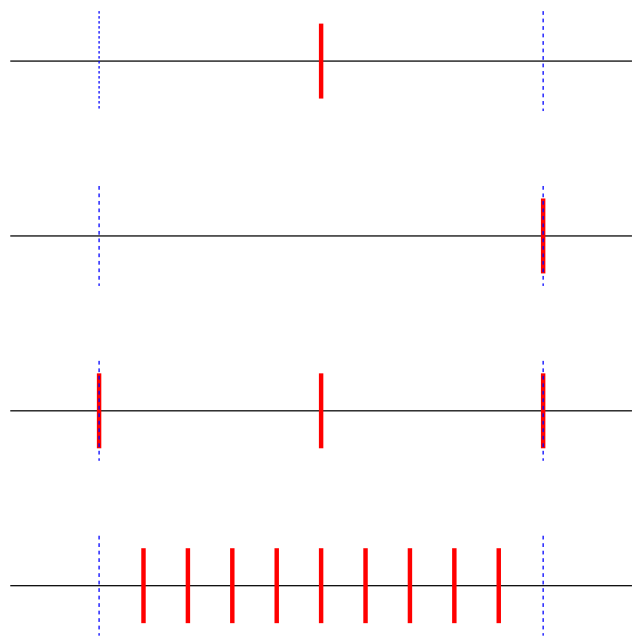


Thick → thin quadrupole



quadrupole of finite length

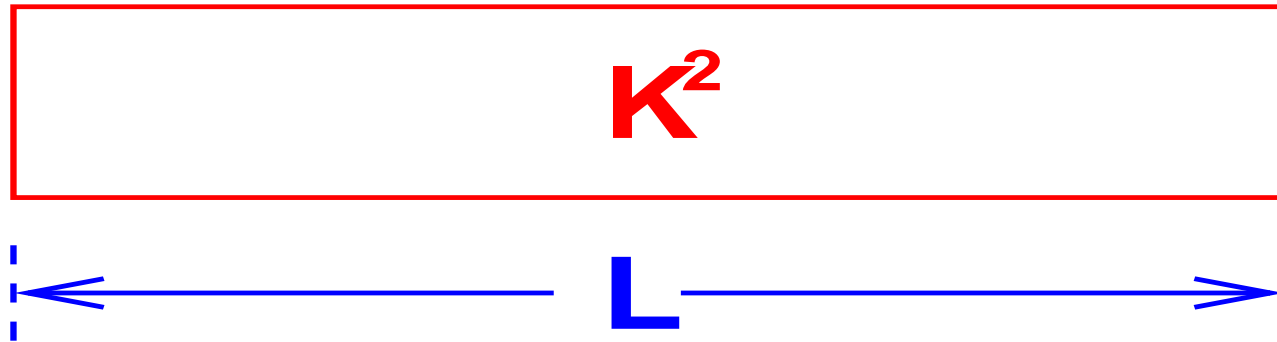
options:



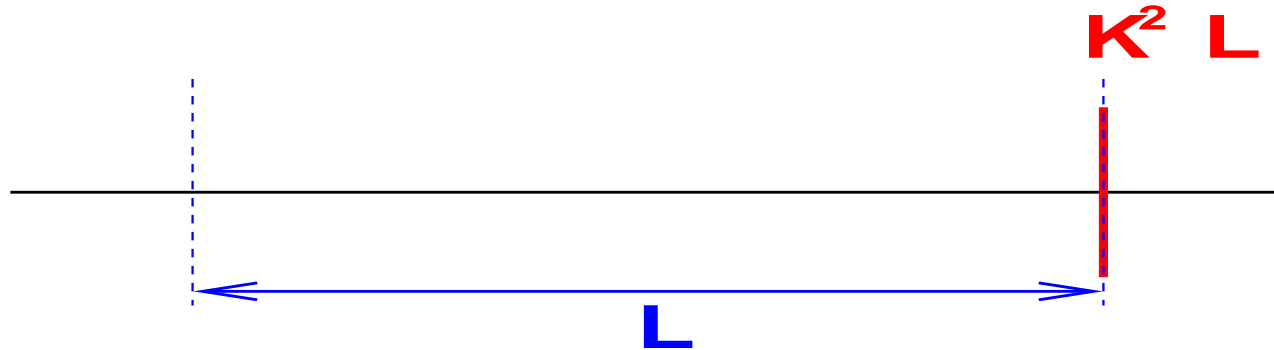
Which is a good strategy ? → accuracy and speed



Thick quadrupole ..



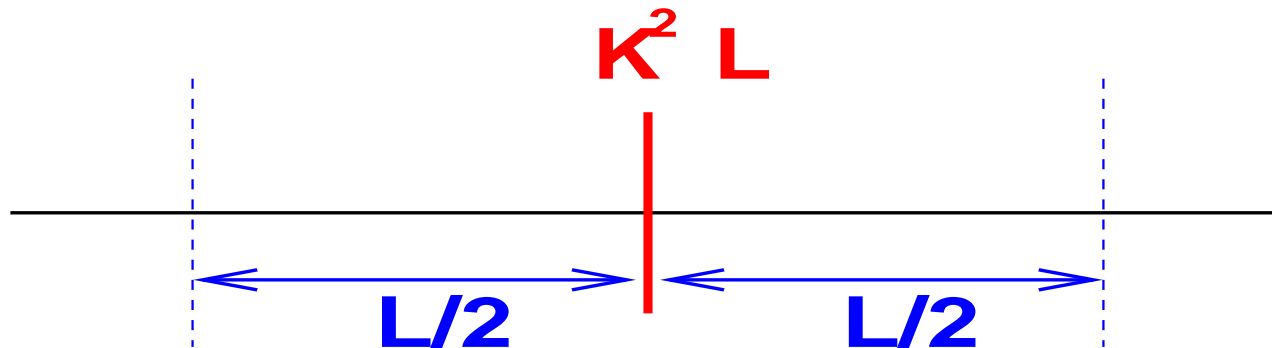
First order ..



- One thin quadrupole "kick" and one drift combined
- Resembles "symplectification" of type (C)

$$\begin{pmatrix} 1 & 0 \\ -K^2 \cdot L & 1 \end{pmatrix} \begin{pmatrix} 1 & L \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & L \\ -K^2 \cdot L & 1 - K^2 L^2 \end{pmatrix}$$

Second order ..



- One thin quadrupole "kick" between two drifts
- Resembles more accurate, symplectic model of type (D)

$$\begin{pmatrix} 1 & \frac{1}{2}L \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ -K^2 \cdot L & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{1}{2}L \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 - \frac{1}{2}K^2L^2 & L - \frac{1}{4}K^2L^3 \\ -K^2 \cdot L & 1 - \frac{1}{2}K^2L^2 \end{pmatrix}$$

Accuracy of thin lenses

■ One kick at the end (or beginning):

→ Error (inaccuracy) of first order $\mathcal{O}(L^1)$

■ One kick in the centre:

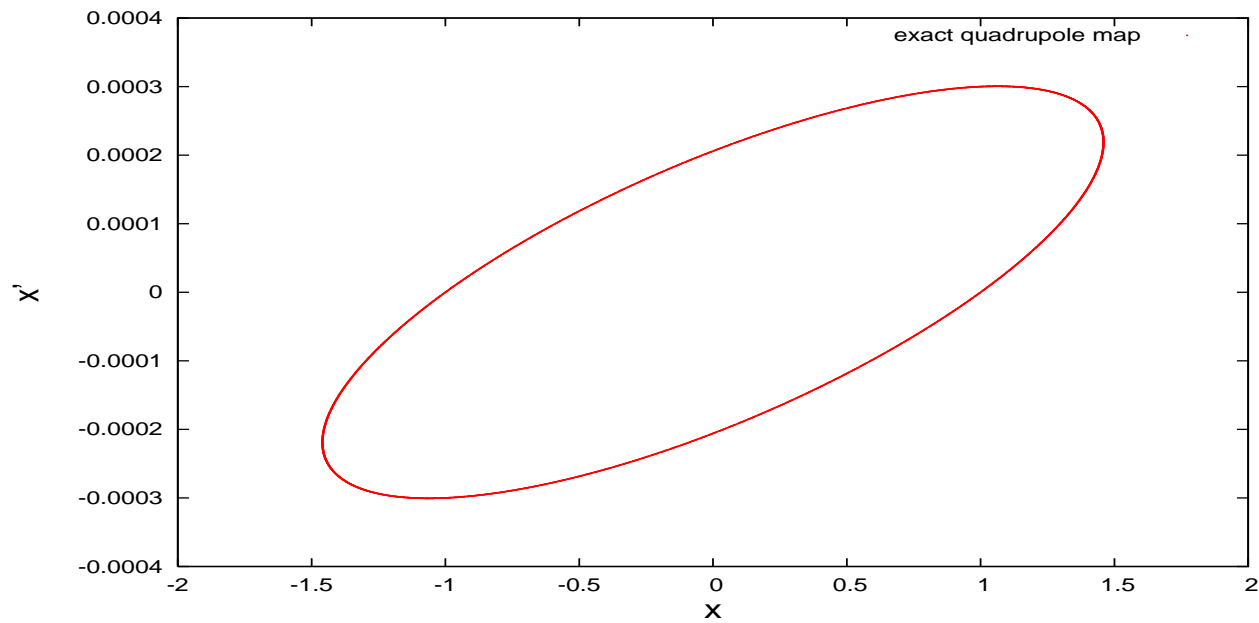
→ Error (inaccuracy) of second order $\mathcal{O}(L^2)$

➤ It is very relevant **how** to apply thin lenses

➤ Aim should be to be precise and fast (and simple to implement)



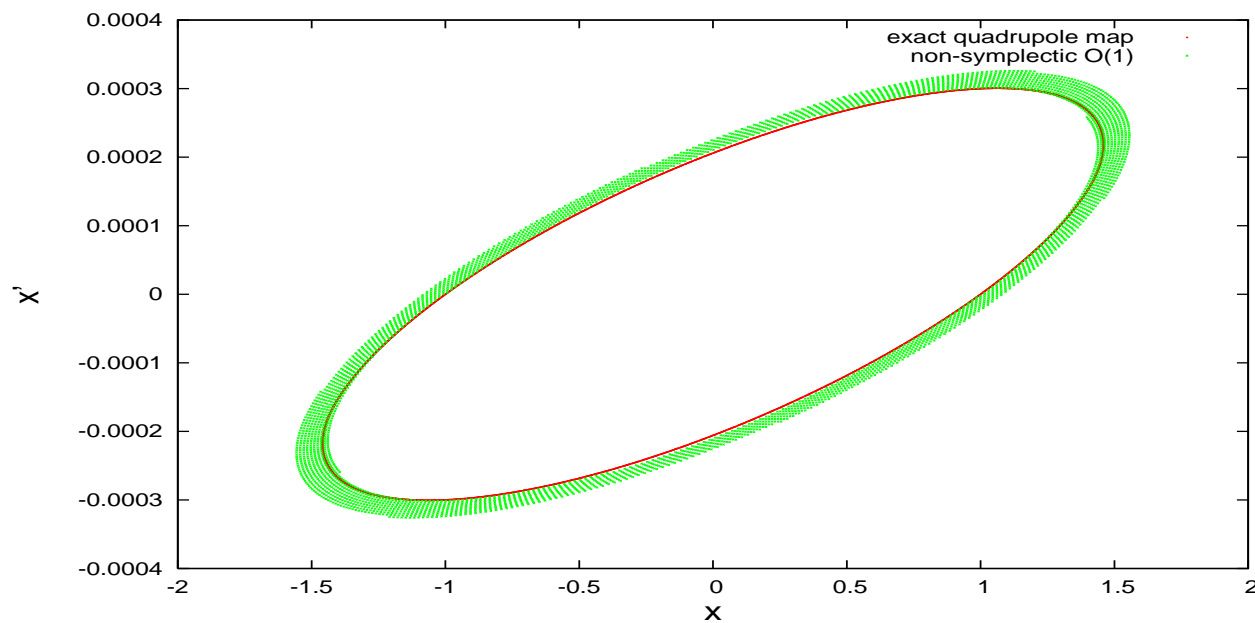
What is the point ???



➤ Phase ellipse - quadrupole exact solution

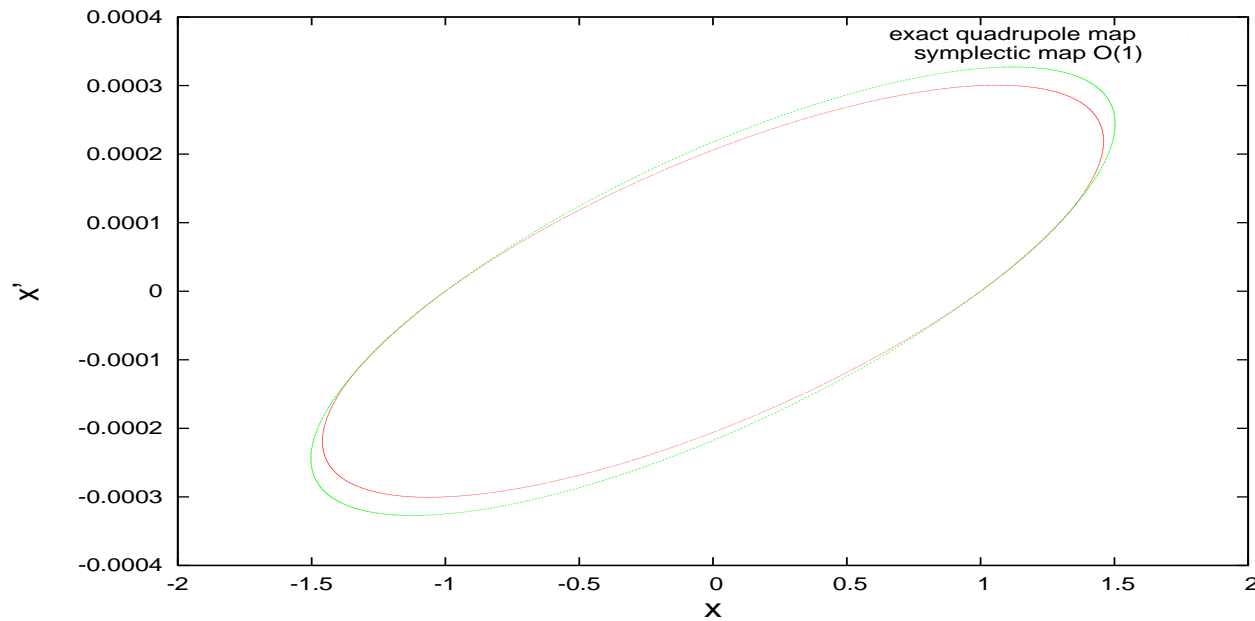


Quadrupole non-symplectic solution



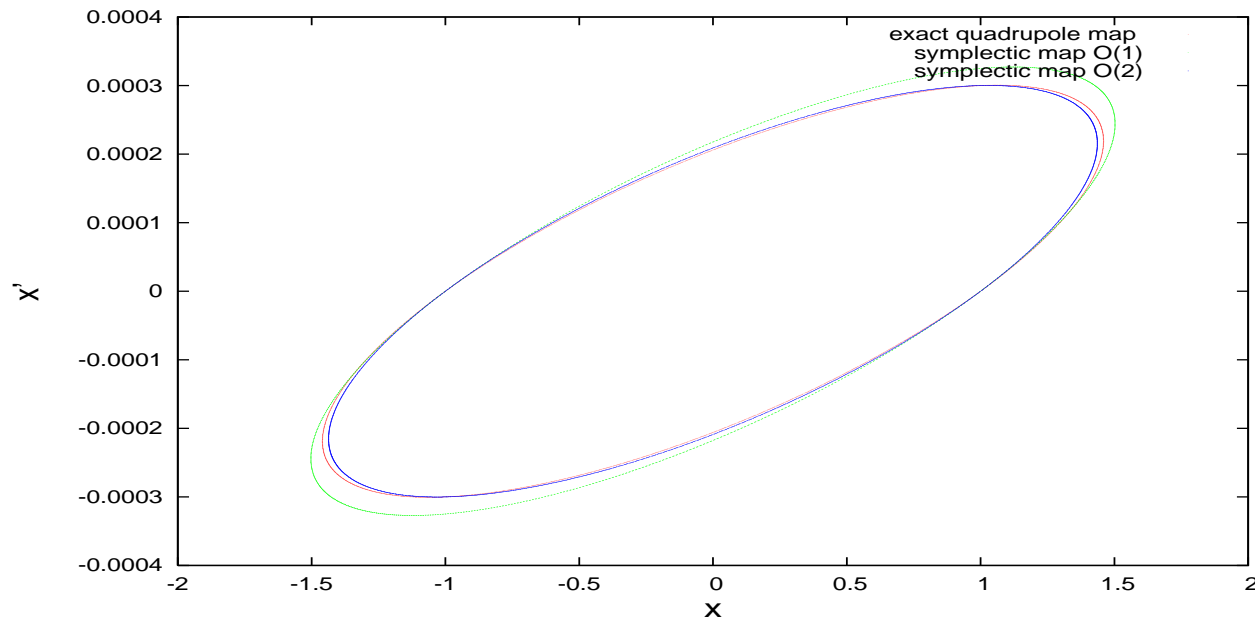
➤ Non-symplecticity: particles spiral towards outside

Quadrupole symplectic $\mathcal{O}(L^1)$ solution



➤ symplecticity: but phase space ellipse not accurate

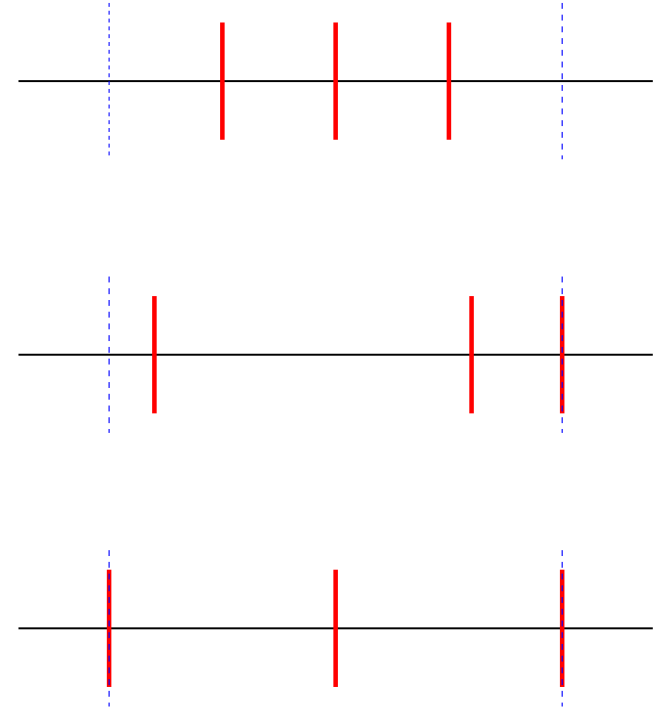
Quadrupole symplectic $\mathcal{O}(L^2)$ solution



➤ symplecticity: phase space ellipse accurate enough

Can we do better ?

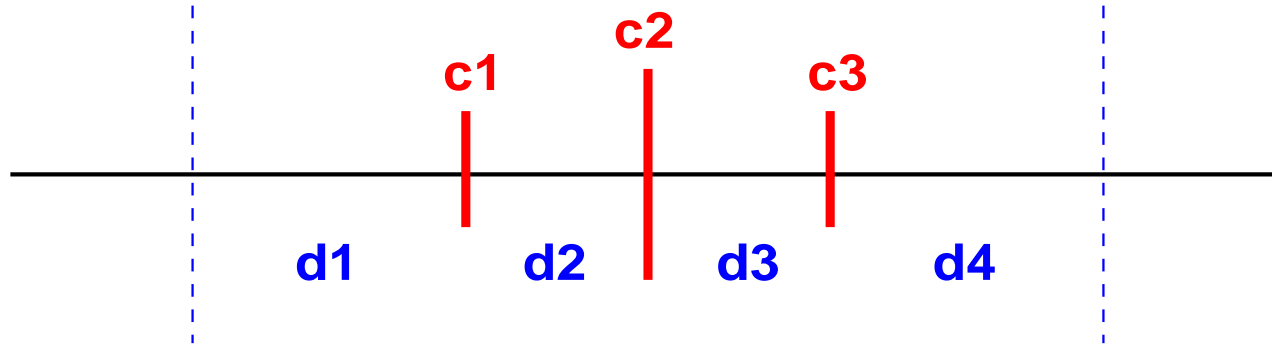
- Try more slices, e.g. 3 kicks:
- How to put them ?
- Hope you are already alerted ...
- Allow that they are at different positions **and** have different strengths
- Minimize the inaccuracy



Question: is one of the options obviously wrong ? If yes, why ?

Can we do better ?

- Try a model with 3 kicks:

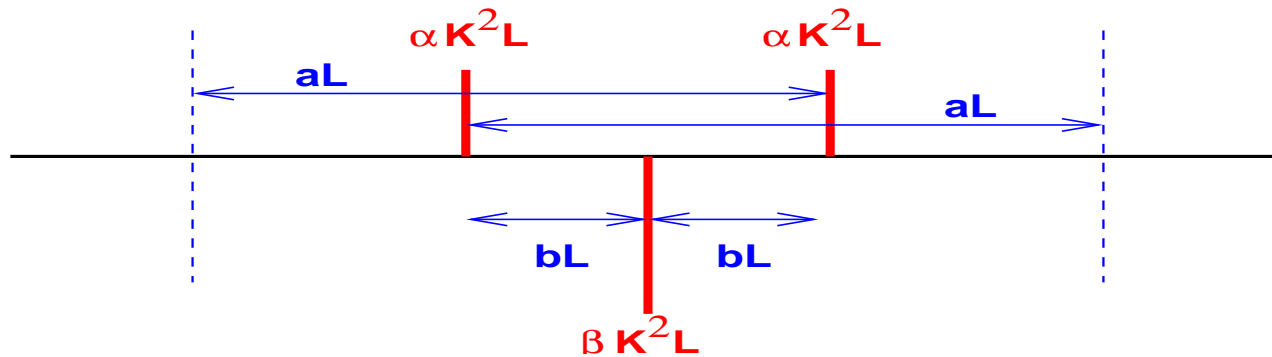


- ➡ To get best accuracy (i.e. deviation from exact solution):
 - Optimize kicks **c1**, **c2**, **c3**
 - Optimize drifts **d1**, **d2**, **d3**, **d4**



Can we do better ?

➤ Try a model with 3 kicks:



➤ with:

$$a \approx 0.6756, \quad b \approx -0.1756, \quad \alpha \approx 1.3512, \quad \beta \approx -1.7024$$

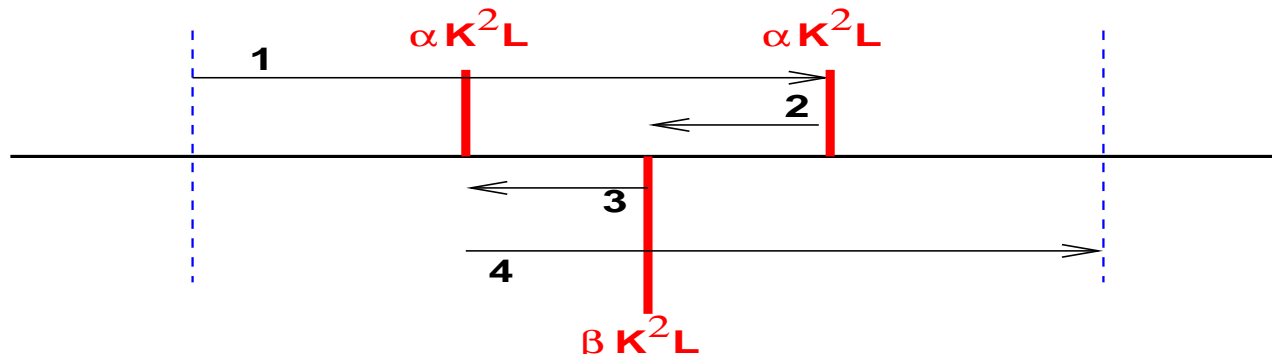
➤ we have a $\mathcal{O}(4)$ integrator ...

➤ (a $\mathcal{O}(6)$ integrator would require 9 kicks (!) ...)



Can we do better ?

- Try a model with 3 kicks:



- Must track backwards ! Change interpretation !
- Thin lenses not a new sequence of magnets (a la MAD)
- What about space charge calculations ?



Symplectic integration

- What we do is **Symplectic Integration**
- From a lower order integration scheme (1 kick), construct higher order scheme
- Formally (for the formulation of $S_k(t)$ see later):
 - From a 2nd order scheme (1 kick) $S_2(t)$ we construct a 4th order scheme (3 kicks = 3 x 1 kick) like:

$$S_4(t) = S_2(x_1 t) \circ S_2(x_0 t) \circ S_2(x_1 t) \quad \text{with:}$$

$$x_0 = \frac{-2^{1/3}}{2 - 2^{1/3}} \approx -1.7024 \quad x_1 = \frac{1}{2 - 2^{1/3}} \approx 1.3512$$

Symplectic integration

■ Can be considered as an iterative scheme (see e.g. H. Yoshida, 1990, E. Forest, 1998²⁾):

➤ If $S_{2k}(t)$ is a symmetric integrator of order $2k$, then:

$$S_{2k+2}(t) = S_{2k}(x_1 t) \circ S_{2k}(x_0 t) \circ S_{2k}(x_1 t) \quad \text{with:}$$

$$x_0 = \frac{-\sqrt{2}^{2k+1}}{2 - \sqrt{2}^{2k+1}} \quad x_1 = \frac{1}{2 - \sqrt{2}^{2k+1}}$$

■ Higher order integrators can be obtained in a similar way

 ²⁾ E. Forest, "Beam Dynamics, A New Attitude and Framework", 1998 

Symplectic integration

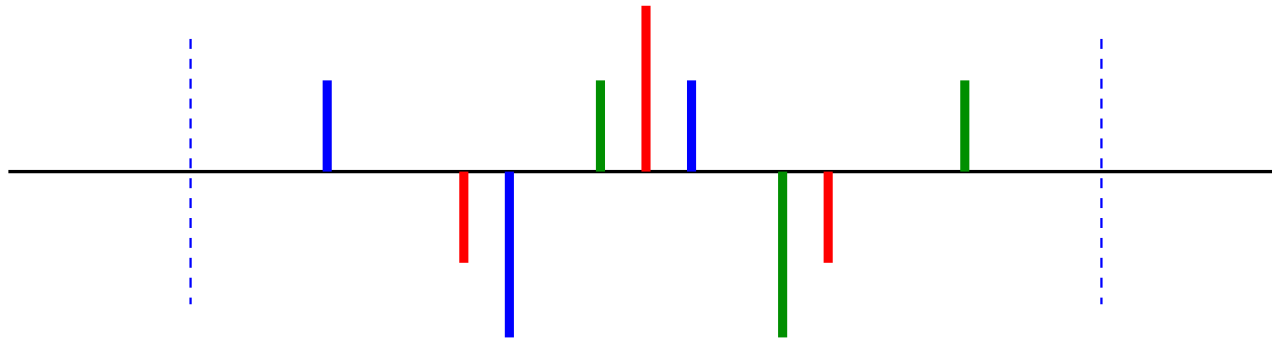
■ **Example: From a 4th order to 6th order**

$$S_6(t) = S_4(x_1t) \circ S_4(x_0t) \circ S_4(x_1t)$$

■ **We get 3 times 4th order with 3 kicks each, we have the 9 kick, 6th order integrator mentioned earlier**



Integrator of order 6





- Requires 9 kicks
- We have 3 interleaved 4th order integrators
- Can be used in iterative scheme



Some remarks:

- We have used a linear map (quadrupole) to demonstrate the integration
 - Can that be applied for other maps (solenoids, higher order, non-linear maps) ?
 - Yes !!
 - We get the same integrators !
 - Proof and systematic (and easy) extension in the form of Lie operators²⁾ (see later)
- Best accuracy for thin lenses !

 ²⁾ H. Yoshida, Physics Letters A, Volume 150 (1990) 262. 

Accuracy of thin lenses

What about accuracy of **non-linear** elements ?

assume a general case:

$$x'' = f(x)$$

- Disadvantage : usually a closed solution through the element does not exist, integration necessary
 - Advantage : They are usually thin (thinner than dipoles, quadrupoles ...)
 - Dipoles: ≈ 14.3 m
 - Quadrupole: $\approx 2 - 5$ m
 - Sextupoles, Octupoles: ≈ 0.30 m
- ➔ Can try our simplest thin lens approximation first ...

Accuracy of thin lenses - our $\mathcal{O}(2)$ model

$$1.Step \quad \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1+L/2} = \begin{pmatrix} 1 & \frac{L}{2} \\ 0 & 1 \end{pmatrix} \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1}$$

$$2.Step \quad \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1+L/2} = \begin{pmatrix} x \\ x' + \Delta x' \end{pmatrix}_{s_1+L/2}$$

$$3.Step \quad \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1+L} = \begin{pmatrix} 1 & \frac{L}{2} \\ 0 & 1 \end{pmatrix} \circ \begin{pmatrix} x \\ x' \end{pmatrix}_{s_1+L/2}$$



Accuracy of thin lenses

Assume the general case:

$$x'' = f(x) (= \Delta x')$$

➤ Using this thin lens approximation (type D, $\mathcal{O}(2)$) gives:

$$x'(L) \approx x'_0 + Lf\left(x_0 + \frac{L}{2}x'_0\right)$$

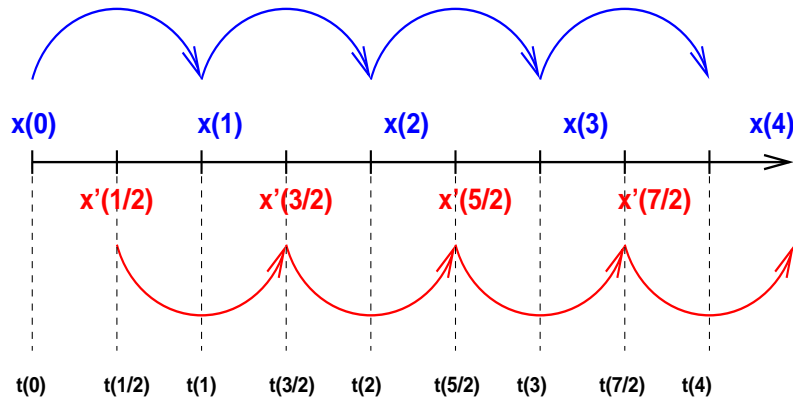
$$x(L) \approx x_0 + \frac{L}{2}(x'_0 + x'(L))$$

➤ This is also called "leap frog" algorithm/integration

➤ It is symplectic (... and time reversible) !!



Interlude ...



For any: $x'' = f(x, x', t)$ we can solve it by:

$$x'_{i+3/2} \approx x'_{i+1/2} + f(x_{i+1})\Delta t$$

$$x_{i+1} \approx x_i + x'_{i+1/2}\Delta t$$



Accuracy of thin lenses

Accuracy of "leap frog" algorithm/integration"

the (exact) Taylor expansion gives:

$$x(L) = x_0 + x'_0 L + \frac{1}{2} f(x_0) L^2 + \frac{1}{6} x'_0 f'(x_0) L^3 + \dots$$

the (approximate) "leap frog" algorithm gives:

$$x(L) = x_0 + x'_0 L + \frac{1}{2} f(x_0) L^2 + \frac{1}{4} x'_0 f'(x_0) L^3 + \dots$$

➤ Errors are $\mathcal{O}(L^3)$ (of course)

➤ For small L acceptable, and symplectic, extend to our symplectic integration

Accuracy of thin lenses

For bar/coffee discussions:

why did I write:

$$x'' = f(x)$$

and not:

$$x'' = f(x, x')$$



Accuracy of thin lenses

An application, assume a (1D) sextupole with (definition of k not unique !):


$$x'' = k \cdot x^2 = f(x)$$

using the thin lens approximation (type D) gives:

$$x(L) = x_0 + x'_0 L + \frac{1}{2} k x_0^2 L^2 + \frac{1}{2} k x_0 x'_0 L^3 + \frac{1}{8} k x_0'^2 L^4$$

$$x'(L) = x'_0 + k x_0^2 L + k x_0 x'_0 L^2 + \frac{1}{4} k x_0'^2 L^3$$

Map for thick sextupole of length L in thin lens approximation, accurate to $\mathcal{O}(L^2)$



Accuracy of thin lenses

An application, assume a (1D) sextupole with (definition of k not unique !):


$$x'' = k \cdot x^2 = f(x)$$

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$$x'(L) = x'_0 + k x_0^2 L + k x_0 x'_0 L^2 + \frac{1}{4} k x_0'^2 L^3$$

Map for thick sextupole of length L in thin lens approximation, accurate to $\mathcal{O}(L^2)$



Some comments:

- We have interleaved kicks with drifts
- Is that always necessary ?
 - No !
 - Can be any map with an exact expression
 - In most cases the exact map is a linear map (matrix)
- We have derived element maps for tracking from the equation of motion using this technique → can track now



Simulation and tracking

We have now sufficient tools for a simulation code

- Main purpose of such a code: Propagate particles around a ring or along a beam line
- Results (amongst others):
 - Phase space topology (Poincare sections,..)
 - Global properties (after some analysis), e.g. stability, detuning, invariants, frequency map analysis
- In our terminology: Tracking codes produce maps (i.e. relate output to the input)!
- Can we extract more "analytical" maps ?



So far ...

- Concept and representation by MAPS
- Computation and analysis of One-Turn-Maps
- Normal form analysis of LINEAR MAPS
- Introduction of Taylor maps
- Introduction of symplectic integration



Mathematical and Numerical Methods
for Non-linear Beam Dynamics in Rings
(an introduction)

Part 2

Werner Herr, CERN

http://cern.ch/Werner.Herr/CAS2013/lectures/Trondheim_methods.pdf

For many more details:

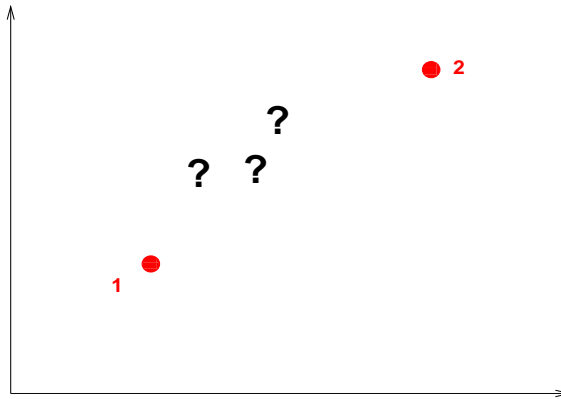
<http://cern.ch/Werner.Herr/METHODS>

The plan now ...

- Extend all concepts to non-linear dynamics
 - Lagrangian and Hamiltonian dynamics
 - How to use that → Lie transforms
 - How to analyse that → Non-linear normal forms
 - How to analyse that better → Differential Algebra (DA)
 - Avoid abstract definitions and formulation, but:
 - Intuitive (but correct !) treatment
 - Useful formulae and examples
 - Real life examples and demonstration (DA)
-

Hamilton principle

- Problem: describe the motion of a system (e.g. 1 or more particles) between times t_1 and t_2



- Describe by coordinates q_i ($i = 1, n$)
 n are degrees of freedom of the system
(Goldstein convention)

Hamilton principle

■ Describe motion by a function L

$$L(q_1, \dots, q_n, \dot{q}_1, \dots, \dot{q}_n, t)$$

(q_1, \dots, q_n) ... generalized coordinates

$(\dot{q}_1, \dots, \dot{q}_n)$... generalized velocities

■ The function L defines the **Lagrange function**

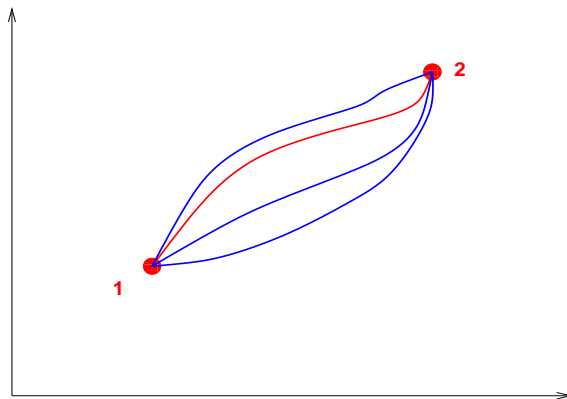
■ The integral $I = \int L(q_i, \dot{q}_i, t) dt$ defines the **action**

Without proof or derivation:

$$L = T - V = \text{kinetic energy} - \text{potential energy}$$


Hamilton principle

$$I = \int_1^2 L(q_i, \dot{q}_i, t) dt = \textit{extremum}$$



- **Hamiltonian principle:** system moves such that the action I becomes an extremum



Extremum principle ?

■ Not new:

- Used in optics: Fermat principle
- Quantum mechanics (path integrals)
- General relativity
- ...



Lagrange formalism

Without proof:

$$I = \int_1^2 L(q_i, \dot{q}_i, t) dt = \textit{extremum}$$

is fulfilled when:

$$\frac{d}{dt} \frac{\partial L}{\partial \dot{q}_i} - \frac{\partial L}{\partial q_i} = 0$$

(Euler - Lagrange equation)

From Lagrangian to Hamiltonian ..

- Lagrangian $L(q_1, \dots, q_n, \dot{q}_1, \dots, \dot{q}_n, t)$ in generalized coordinates and velocities
- Provides (n) second order differential equations
- Try to get:
 - Generalized momenta instead of velocities
 - First order differential equations (always solvable)

Corresponding (so-called conjugate) momenta p_j are:

$$p_j = \frac{\partial L}{\partial \dot{q}_j}$$



From Lagrangian to Hamiltonian ..

■ Lagrangian:

- n second order equations
- n -dimensional **coordinate space**

■ Hamiltonian:

- $2n$ first order equations
- $2n$ -dimensional **phase space**



From Lagrangian to Hamiltonian ..

Once we know what the canonical momenta p_i are: the **Hamiltonian** is a transformation of the **Lagrangian**:

$$H(q_j, p_j, t) = \sum_i \dot{q}_i p_i - L(q_j, \dot{q}_j, t)$$

without proof:

$H = T + V =$ kinetic energy + potential energy

we obtain 2 first order equation of motion:

$$\frac{\partial H}{\partial q_j} = -\dot{p}_j = -\frac{dp_j}{dt},$$

$$\frac{\partial H}{\partial p_j} = \dot{q}_j = \frac{dq_j}{dt}$$



Hamiltonian of particle in EM fields

For the Hamiltonian of a (relativistic) particle in a electro-magnetic field we have ($q \rightarrow x$):

$$\mathcal{H}(\vec{x}, \vec{p}, t) = c\sqrt{(\vec{p} - e\vec{A}(\vec{x}, t))^2 + m_0^2c^2} + e\Phi(\vec{x}, t)$$

where $\vec{A}(\vec{x}, t)$, $\Phi(\vec{x}, t)$ the vector and scalar scalar potential

Using canonical variables and the design path length s as independent variable (bending in x-plane):

$$\mathcal{H} = -\left(1 + \frac{x}{\rho}\right) \cdot \sqrt{(1 + \delta)^2 - p_x^2 - p_y^2} + \frac{x}{\rho} + \frac{x^2}{2\rho^2} - \frac{A_s(x, y)}{B_0\rho}$$

where $\delta = (p - p_0)/p$ is relative momentum deviation and $A_s(x, y)$ longitudinal component of the vectorpotential [MB].



Hamiltonian of particle in EM fields

The magnetic fields can be described with the multipole expansion:

$$B_y + iB_x = \sum_{n=1} (b_n + ia_n)(x + iy)^{n-1}$$

and since $\vec{B} = \nabla \times \vec{A}$:

$$A_s = \sum_{n=1} \frac{1}{n} [(b_n + ia_n)(x + iy)^n]$$

➤ $n = 1$ refers to dipole (not always true !)

➤ For a large machine ($x \ll \rho$) we expand the root and sort the variables →



Hamiltonian (for large machine) ..

$$\mathcal{H} = \overbrace{\frac{p_x^2 + p_y^2}{2(1 + \delta)}}^{\text{kinematic}} - \underbrace{\left(\underbrace{\frac{x\delta}{\rho}}_{\text{dispersive}} + \underbrace{\frac{x^2}{2\rho^2}}_{\text{focusing}} \right)}_{\text{dipole}} + \overbrace{\frac{k_1}{2}(x^2 - y^2)}^{\text{quadrupole}} + \overbrace{\frac{k_2}{6}(x^3 - 3xy^2)}^{\text{sextupole}}$$

$$\left(\text{using (MAD convention)} : k_n = \frac{1}{B\rho} \frac{\partial^n B_y}{\partial x^n} \right)$$

- The Hamiltonian describes exactly the motion of a particle through a magnet
- Basis to extend the linear to a non-linear formalism

But how do we use it ??



Poisson brackets

Introduce Poisson bracket for a differential operator:

$$[f, g] = \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i} \frac{\partial g}{\partial p_i} - \frac{\partial f}{\partial p_i} \frac{\partial g}{\partial x_i} \right)$$

Here the variables x_i, p_i are canonical variables, f and g are functions of x_i and p_i .

We can now write (using the Hamiltonian H for g):

$$f(x_i, p_i) = x_i \Rightarrow [x_i, H] = \frac{\partial H}{\partial p_i} = \frac{dx_i}{dt}$$

$$f(x_i, p_i) = p_i \Rightarrow [p_i, H] = -\frac{\partial H}{\partial x_i} = \frac{dp_i}{dt}$$

Poisson brackets encode Hamilton's equations



Poisson brackets

- Poisson bracket $[f, H]$ describes the **time evolution** of the system (the function f)

It is a special case of:

$$\frac{df}{dt} = [f, H] + \frac{\partial f}{\partial t}$$

If H does not explicitly depend on time and:

$$[f, H] = 0$$

implies that f is an invariant of the motion !

Poisson brackets determine invariants



Lie transformations

We can define:

$$: f : g = [f, g]$$

where $: f :$ is an operator acting on the function g :

$$: f := [f,]$$

The operator $: f :$ is called a **Lie Operator**

It acts on functions $g(x, p)$, special cases:

$$g(x, p) = x \quad \rightarrow \quad : f : x$$

$$g(x, p) = p \quad \rightarrow \quad : f : p$$

Lie operators are Poisson brackets in waiting



Useful formulae for calculations

With x coordinate, p momentum, try special cases for f :

$$: x : = \frac{\partial}{\partial p} \qquad : p : = - \frac{\partial}{\partial x}$$

$$: x :^2 = \frac{\partial^2}{\partial p^2} \qquad : p :^2 = \frac{\partial^2}{\partial x^2}$$

$$: x^2 : = 2x \frac{\partial}{\partial p} \qquad : p^2 : = -2p \frac{\partial}{\partial x}$$

$$: xp : = p \frac{\partial}{\partial p} - x \frac{\partial}{\partial x} \qquad : x :: p : = : p :: x : = - \frac{\partial^2}{\partial x \partial p}$$



More useful formulae for calculations

With x coordinate, p momentum, as usual:

$$: p^2 : x = \frac{\partial p^2}{\partial x} \frac{\partial x}{\partial p} - \frac{\partial p^2}{\partial p} \frac{\partial x}{\partial x} = -2p$$

$$: p^2 : p = \frac{\partial p^2}{\partial x} \frac{\partial p}{\partial p} - \frac{\partial p^2}{\partial p} \frac{\partial p}{\partial x} = 0$$

$$(: p^2 :)^2 x = : p^2 : (: p^2 : x) = : p^2 : (-2p) = 0$$

$$(: p^2 :)^2 p = : p^2 : (: p^2 : p) = : p^2 : (0) = 0$$



Lie transformations

We can define powers as:

$$(: f :)^2 g =: f : (: f : g) = [f, [f, g]] \quad \text{etc.}$$

in particular:

$$e^{:f:} = \sum_{i=0}^{\infty} \frac{1}{i!} (: f :)^i$$

$$e^{:f:} = 1 + : f : + \frac{1}{2!} (: f :)^2 + \frac{1}{3!} (: f :)^3 + \dots$$

The operator $e^{:f:}$ is call an **Lie Transformation**

Lie transformations - example

Lie operators act on functions like x, p (canonical momentum, instead of x'), for example:

$$:p^2 :x = -2p \quad :p^2 :p = 0$$

or as a Lie transformation with $f = -Lp^2/2$:

$$e^{:-Lp^2/2:} x = x - \frac{1}{2}L \underbrace{:p^2 :x}_{=-2p} + \frac{1}{8}L^2 \underbrace{(:p^2 :)^2 x}_{=0} + ..$$

$$= \mathbf{x + Lp}$$

$$e^{:-Lp^2/2:} p = p - \frac{1}{2}L \underbrace{:p^2 :p}_{=0} + ...$$

$$= \mathbf{p}$$

Lie transformations - example

Lie operators act on functions like x, p (canonical momentum, instead of x'), for example:

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$$= \mathbf{x + Lp}$$

$$e^{:-Lp^2/2:} p = p - \frac{1}{2}L \underbrace{:p^2 :p}_{=0} + ...$$

$$= \mathbf{p}$$

This is the transformation of a drift space of length L !!

Lie transformations - general

Acting on the phase space coordinates:

$$e^{i f} (x, p)_0 = (x, p)_1$$

that is:

$$e^{i f} x_0 = x_1$$

$$e^{i f} p_0 = p_1$$

- Lie transforms describe how to go from one point to another [AC1, AD].
- Through a machine element (drift, magnet ...) fully described by f
- But what is f ?



Lie transformations

- The generator f is the Hamiltonian of the element !
- We use the Hamiltonian to describe the motion through an individual element
- Inside a single element the motion is "smooth" (in the full machine it is not !)
- Can track "thick" elements (and still symplectic !)



Some formulae for Lie transforms

With a constant, f, g, k arbitrary functions:

$$: a : = 0 \quad e^{:a:} = 1$$

$$: f : a = 0 \quad e^{:f:} a = a$$

$$e^{:f:} g(x) = g(e^{:f:} x)$$

$$e^{:f:} G(: g :) e^{-:f:} = G(: e^{:f:} g :)$$

$$e^{:f:} [g, h] = [e^{:f:} g, e^{:f:} h]$$

$$(e^{:f:})^{-1} = e^{-:f:}$$

and very important:

$$e^{:f:} e^{:g:} e^{-:f:} = e^{:e^{:f:} g:}$$



More examples (1D):

For:

$$f = -\frac{L}{2}p^2$$

we obtained:

$$e^{:f:}x = x + Lp$$

$$e^{:f:}p = p$$

➤ Drift space, seen that already



More examples (1D):

For:

$$f = -\frac{L}{2}(k^2 x^2 + p^2)$$

we would get (try it !):

$$e^{:f:} x = e^{:-\frac{L}{2}(k^2 x^2 + p^2):} x$$

$$e^{:f:} p = e^{:-\frac{L}{2}(k^2 x^2 + p^2):} p$$

Remember:

$$e^{:f:} g = \sum_{n=0}^{\infty} \frac{:f:^n}{n!} g$$



More examples (1D):

For:

$$f = -\frac{L}{2}(k^2 x^2 + p^2)$$

we would get (try it !):

$$e^{:-\frac{L}{2}(k^2 x^2 + p^2):} x = \sum_{n=0}^{\infty} \left(\frac{(-k^2 L^2)^{2n}}{(2n)!} \cdot x + L \frac{(-k^2 L^2)^{2n+1}}{(2n+1)!} \cdot p \right)$$

$$e^{:-\frac{L}{2}(k^2 x^2 + p^2):} p = \sum_{n=0}^{\infty} \left(\frac{(-k^2 L^2)^{2n}}{(2n)!} \cdot p - k \frac{(-k^2 L^2)^{2n+1}}{(2n+1)!} \cdot x \right)$$



More examples (1D):

For:

$$f = -\frac{L}{2}(k^2 x^2 + p^2)$$

we would get (try it !):

$$e^{i f} x = x \cos(kL) + \frac{p}{k} \sin(kL)$$

$$e^{i f} p = -kx \sin(kL) + p \cos(kL)$$

➤ Thick, focusing quadrupole !



Hamiltonians of some **thick** machine elements (3D)

dipole:

$$H = -\frac{-x\delta}{\rho} + \frac{x^2}{2\rho^2} + \frac{p_x^2 + p_y^2}{2(1 + \delta)}$$

quadrupole:

$$H = \frac{1}{2}k_1(x^2 - y^2) + \frac{p_x^2 + p_y^2}{2(1 + \delta)}$$

sextupole:

$$H = \frac{1}{6}k_2(x^3 - 3xy^2) + \frac{p_x^2 + p_y^2}{2(1 + \delta)}$$

octupole:

$$H = \frac{1}{4}k_3(x^4 - 6x^2y^2 + y^4) + \frac{p_x^2 + p_y^2}{2(1 + \delta)}$$



Remark:

In many cases the non-linear effects by the kinematic term is negligible and

$$H = \frac{1}{2}k_1(x^2 - y^2) + \frac{p_x^2 + p_y^2}{2(1 + \delta)}$$

is written as:

$$H = \frac{1}{2}k_1(x^2 - y^2) + \frac{p_x^2 + p_y^2}{2}$$

In 1D it reduces to previous example



Why all that ???

If we know the Hamiltonian of a machine elements (magnet) then:

$$e^{iH} x_0 = x_1$$

$$e^{iH} p_0 = p_1$$

This is also true for functions of x and p :

$$e^{iH} f_0(x, p) = f_1(x, p)$$

The miracles:

- Poisson brackets create **symplectic** maps
- Exponential form e^{iH} is **always** symplectic
- Better: the exponent is directly connected to the invariant of the transfer map !!



Many machine elements

- We can combine many machine elements f_n by applying one transformation after the other:

$$e^{i h} = e^{i f_1} e^{i f_2} \dots e^{i f_N}$$

- Not restricted to matrices, i.e. linear elements ...
- And arrive at a transformation for the full ring
➔ a one turn map
- The one turn map is the exponential of the effective Hamiltonian:

$$\mathcal{M}_{ring} = e^{-C \mathcal{H}_{eff}}$$



Why all that ???

concatenation very easy:

$$e^{:h:} = e^{:f:} e^{:g:} = e^{:f+g:}$$

when f and g commute (i.e. $[f, g] = [g, f] = 0$)

■ Otherwise formalism exist



Concatenation

To combine:

$$e^{:h:} = e^{:f:} e^{:g:}$$

We can use the formula (Baker-Campbell-Hausdorff (BCH)):

$$\begin{aligned} h = f &+ g + \frac{1}{2}[f, g] + \frac{1}{12}[f, [f, g]] + \frac{1}{12}[g, [g, f]] \\ &+ \frac{1}{24}[f, [g, [g, f]]] - \frac{1}{720}[g, [g, [g, [g, f]]]] \\ &- \frac{1}{720}[f, [f, [f, [f, g]]]] + \frac{1}{360}[g, [f, [f, [f, g]]]] + \dots \end{aligned}$$

or :

$$\begin{aligned} h = f &+ g + \frac{1}{2} : f : g + \frac{1}{12} : f :^2 g + \frac{1}{12} : g :^2 f \\ &+ \frac{1}{24} : f :: g :^2 f - \frac{1}{720} : g :^4 f \\ &- \frac{1}{720} : f :^4 g + \frac{1}{360} : g :: f :^3 g + \dots \end{aligned}$$



Concatenation

To combine:

$$e^{:h:} = e^{:f:} e^{:g:}$$

if one of them (f or g) is small, can truncate the series and get a very useful formula. Assume g is small:

$$e^{:f:} e^{:g:} = e^{:h:} = \exp \left[: f + \left(\frac{: f :}{1 - e^{-:f:}} \right) g + \mathcal{O}(g^2) : \right]$$



Non-linear kicks

General thin lens kick $f(x)$:

$$e: \int_0^x f(x') dx':$$

gives for the map:

$$x = x_0$$

$$p = p_0 + f(x)$$

Example: thin lens multipole of order n ($f(x) = a \cdot x^n$):

$$e: \frac{a}{n+1} \cdot x^{n+1}:$$

gives for the map:

$$x = x_0$$

$$p = p_0 + ax^n$$



Extension: general monomials

Monomials in x and p of orders n and m ($x^n p^m$)

$$e^{ax^n p^m}:$$

gives for the map (for $n \neq m$):

$$e^{ax^n p^m}:x = x \cdot [1 + a(n - m)x^{n-1}p^{m-1}]^{m/(m-n)}$$

$$e^{ax^n p^m}:p = p \cdot [1 + a(n - m)x^{n-1}p^{m-1}]^{n/(n-m)}$$

gives for the map (for $n = m$):

$$e^{ax^n p^n}:x = x \cdot e^{-anx^{n-1}p^{n-1}}$$

$$e^{ax^n p^n}:p = p \cdot e^{anx^{n-1}p^{n-1}}$$



From the Hamiltonian to the map

We have seen that given the Hamiltonian f of a machine element is known, the Lie operator becomes:

$$f \rightarrow :f:$$

the corresponding map is then:

$$e^{:f:} \quad (e^{:-Lf:})$$

This map is always symplectic and we have (in 1D):

$$e^{:f:} x_0 = x_1$$

$$e^{:f:} p_0 = p_1$$

or using $Z = (x, p_x, y, p_y, \dots)$ (in 2D):

$$e^{:f:} Z_0 = Z_1$$



From the map to the Hamiltonian

The other question \rightarrow assuming we do not have the Hamiltonian, but a matrix \mathcal{M} (from somewhere):

$$\mathcal{M} \equiv \begin{pmatrix} \cos(\mu) + \alpha \sin(\mu) & \beta \sin(\mu) \\ -\gamma \sin(\mu) & \cos(\mu) - \alpha \sin(\mu) \end{pmatrix}$$

i.e.:

$$\mathcal{M}Z_0 = Z_1$$

how do we find the corresponding form for f ?

$$\mathcal{M} \leftrightarrow e^{if} \quad (e^{-\mu f})$$



From the map to the Hamiltonian

For the linear matrix we know that f must be a **quadratic** form in (x, p, \dots) .

Any quadratic form can be written as:

$$f = -\frac{1}{2}Z^* F Z \quad [= -\frac{1}{2}(a \cdot x^2 + b \cdot xp + c \cdot p^2)]$$

where F is a symmetric, positive definite (why ?) matrix. Then we can write (without proof, see e.g. Dragt):

$$: f : Z = S F Z$$

where S is the "symplecticity" matrix.

Therefore we get for the Lie transformation:

$$e^{:f:} Z \leftrightarrow e^{S F} Z$$




From the map to the Hamiltonian

Since we have $n = 2$, we get (using *Hamilton – Cayley theorem*):

$$e^{SF} = \exp \begin{pmatrix} b & c \\ -a & -b \end{pmatrix} = a_0 + a_1 \begin{pmatrix} b & c \\ -a & -b \end{pmatrix}$$

We now have to find a_0 and a_1 !

The eigenvalues of SF are:

$$\lambda_{\pm} = \pm i\sqrt{ac - b^2}$$


From the map to the Hamiltonian

This tells us for the coefficients the conditions:

$$e^{\lambda_+} = a_0 + a_1 \cdot \lambda_+$$

$$e^{\lambda_-} = a_0 + a_1 \cdot \lambda_-$$

and therefore:

$$a_0 = \cos(\sqrt{ac - b^2})$$

$$a_1 = \frac{\sin(\sqrt{ac - b^2})}{\sqrt{ac - b^2}}$$

and

$$e^{SF} = \cos(\sqrt{ac - b^2}) + \frac{\sin(\sqrt{ac - b^2})}{\sqrt{ac - b^2}} \begin{pmatrix} b & c \\ -a & -b \end{pmatrix}$$

From the map to the Hamiltonian

For a general 2×2 matrix:

$$M = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix}$$


we get by comparison:

$$\cos(\sqrt{ac - b^2}) = \frac{1}{2} \text{tr}(M)$$

and

$$\frac{a}{-m_{21}} = \frac{2b}{m_{11} - m_{22}} = \frac{c}{m_{12}} = \frac{\sqrt{ac - b^2}}{\sin(\sqrt{ac - b^2})}$$

for the quadratic form of f :

$$f = -\frac{1}{2}(a \cdot x^2 + b \cdot xp + c \cdot p^2)$$


From the map to the Hamiltonian

For the example of a drift:

$$\mathcal{M} \equiv \begin{pmatrix} 1 & L \\ 0 & 1 \end{pmatrix}$$

we find:

$$a = 0, \quad b = 0, \quad c = L$$

and for the generator:

$$f = -\frac{1}{2}(Lp^2)$$



From the map to the Hamiltonian

For the example of a thin quadrupole:

$$\mathcal{M} \equiv \begin{pmatrix} 1 & 0 \\ -\frac{1}{f} & 1 \end{pmatrix}$$

we find:

$$a = \frac{1}{f}, \quad b = 0, \quad c = 0$$

and for the generator:

$$f = -\frac{1}{2f}(x^2)$$



A very important example ...

$$\mathcal{M} \equiv \begin{pmatrix} \cos \mu + \alpha \sin(\mu) & \beta \sin \mu \\ -\gamma \sin \mu & \cos \mu - \alpha \sin(\mu) \end{pmatrix}$$

corresponds to:

$$e^{i h} = e^{i f_2} = e^{-i \mu \frac{1}{2} (\gamma x^2 + 2 \alpha x p + \beta p^2)}:$$

In this form f is: $-\mu \cdot$ **(Courant-Snyder invariant)**

$$e^{i h} = e^{i f_2} = e^{-i \mu \epsilon}:$$

➤ We have standard $(e^{i f_2})$ for the linear one-turn-matrix (a rotation)...



A very important example ...

With our linear transformation to normalized variables:

$$\begin{pmatrix} \cos \mu + \alpha \sin(\mu) & \beta \sin \mu \\ -\gamma \sin \mu & \cos \mu - \alpha \sin(\mu) \end{pmatrix} \Rightarrow \begin{pmatrix} \cos \mu & \sin \mu \\ -\sin \mu & \cos \mu \end{pmatrix}$$

therefore:

$$e^{:-\mu \frac{1}{2}(\gamma x^2 + 2\alpha xp + \beta p^2)}: \Rightarrow e^{:-\mu \frac{1}{2}(x^2 + p^2)}:$$

and for a **3D** linear system we have for f_2 :

$$f_2 = -\frac{\mu_x}{2}(x^2 + p_x^2) - \frac{\mu_y}{2}(y^2 + p_y^2) - \frac{1}{2}\alpha_c \delta^2$$

or in action variables J :

$$f_2 = -\mu_x J_x - \mu_y J_y - \frac{1}{2}\alpha_c \delta^2$$

➤ A standard ($e^{:f_2:}$) transformation in 3D

First summary: Lie transforms and integrators

- We have powerful tools to describe non-linear elements
- They are always symplectic !
- Can be combined to form a ring (and therefore a non-linear One-Turn-Map)
- Tools and programs are available for their manipulation and computation
- How do we analyse the maps ? Guess: Normal Forms



Normal forms non-linear case

Normal form transformations can be generalized for non-linear maps (i.e. not matrices). If \mathcal{M} is our usual one-turn-map, we try to find a transformation:

$$\mathcal{N} = \mathcal{A}\mathcal{M}\mathcal{A}^{-1}$$

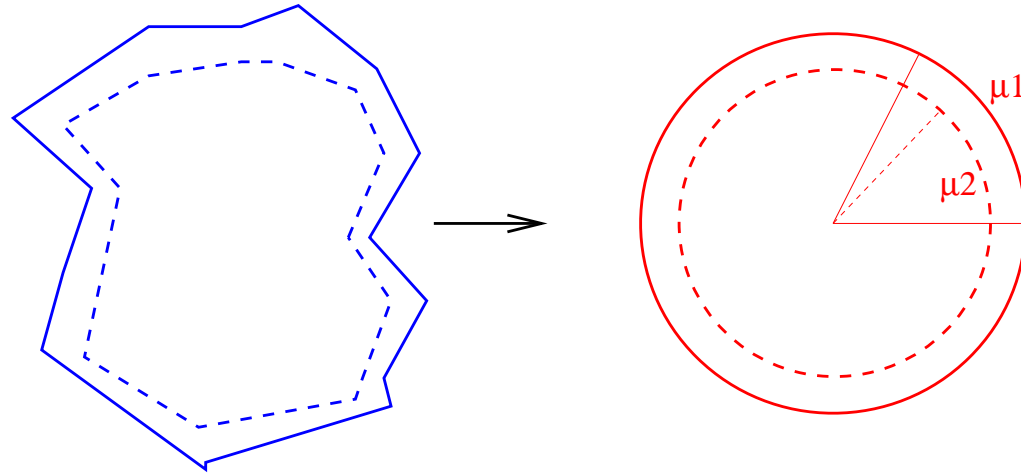
➤ where \mathcal{N} is a simple form (like the rotation we had before)

Of course we now do not have matrices, we use a Lie transform F to describe the transform \mathcal{A} :

$$\mathcal{N} = e^{-:h:} = \mathcal{A}\mathcal{M}\mathcal{A}^{-1} = e{:F:}\mathcal{M}e^{-:F:}$$



Normal forms - non-linear case



- More complicated transformation F required
- Transform to coordinates where map is just a rotation
- In general better done in action angle variables: J, Ψ
- Rotation angle may be amplitude dependent: $\mu \rightarrow \mu(J)$



Normal forms - non-linear case

The canonical transformation A :

$$\mathcal{N} = AMA^{-1} \Rightarrow A = e^{iF}$$

should be the transformation to produce our simple form

➤ $h(J_x, \Psi_x, J_y, \Psi_y, z, \delta) \Rightarrow h(J_x, J_y, \delta) = h_{eff}(J_x, J_y, \delta)$

➤ Should work for any kind of local perturbation

➤ Formalism and software tools exist to find F (see e.g. Chao¹⁾ or E.Forest, M. Berz, J. Irwin, SSC-166)

➤ Once we know $h_{eff}(J_x, J_y, \delta)$ we can derive everything !

 ¹⁾ A. Chao, Lecture Notes on Topics in Accelerator Physics, 2001 

Normal forms - non-linear case

Once we can write the map as (now example in 3D):

$$\mathcal{N} = e^{-i h_{eff}(J_x, J_y, \delta)}$$

where h_{eff} depends only on J_x, J_y , and δ , then we have the tunes:

$$Q_x(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_x}$$

$$Q_y(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_y}$$

and the change of path length:

$$\Delta z = - \frac{\partial h_{eff}}{\partial \delta}$$

Particles with different J_x, J_y and δ have different tunes:

→ Dependence on J is amplitude detuning, dependence on δ are the chromaticities !



How does h_{eff} look like ?


The effective Hamiltonian can be written (here to 3rd order) (see e.g. E. Forest, M. Berz, J. Irwin, SSC-166) as:

$$\begin{aligned} h_{eff} = & + \mu_x J_x + \mu_y J_y + \frac{1}{2} \alpha_c \delta^2 \\ & + c_{x1} J_x \delta + c_{y1} J_y \delta + c_3 \delta^3 \\ & + c_{xx} J_x^2 + c_{xy} J_x J_y + c_{yy} J_y^2 + c_{x2} J_x \delta^2 + c_{y2} J_y \delta^2 + c_4 \delta^4 \end{aligned}$$

and then:

$$Q_x(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_x} = \frac{1}{2\pi} (\mu_x + 2c_{xx} J_x + c_{xy} J_y + c_{x1} \delta + c_{x2} \delta^2)$$

$$Q_y(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_y} = \frac{1}{2\pi} (\mu_y + 2c_{yy} J_y + c_{xy} J_x + c_{y1} \delta + c_{y2} \delta^2)$$

$$\Delta z = -\frac{\partial h_{eff}}{\partial \delta} = \alpha_c \delta + 3c_3 \delta^2 + 4c_4 \delta^3 + c_{x1} J_x + c_{y1} J_y + 2c_{x2} J_x \delta + 2c_{y2} J_y \delta$$


What's the meaning of it ?

- μ_x, μ_y : tunes
- $\frac{1}{2}\alpha_c, c_3, c_4$: linear and non-linear "momentum compaction"
- c_{x1}, c_{y1} : first order chromaticities
- c_{x2}, c_{y2} : second order chromaticities
- c_{xx}, c_{xy}, c_{yy} : detuning with amplitude



Example: sextupole

A linear map followed by a single (weak) sextupole:

$$\mathcal{M} = e^{-:\frac{\mu}{2}x^2+p^2 + \frac{1}{2}\alpha_c\delta^2:} e{:f_3:} = e^{-:\mu J_x + \frac{1}{2}\alpha_c\delta^2:} e{:kx^3 + \frac{p^2}{2(1+\delta)}:}$$

we get for h_{eff} (see e.g. [AC1, EF]):

$$h_{eff} = \mu_x J_x + \frac{1}{2}\alpha_c\delta^2 - kD^3\delta^3 - 3k\beta_x J_x D\delta$$

or in 3D:

$$h_{eff} = \mu_x J_x + \mu_y J_y + \frac{1}{2}\alpha_c\delta^2 - kD^3\delta^3 - 3k\beta_x J_x D\delta + 3k\beta_y J_y D\delta$$



Example: sextupole

When we have h_{eff} in **3D** we obtain:

$$Q_x(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_x} = \frac{1}{2\pi} (\mu_x - 3k\beta_x D\delta)$$

$$Q_y(J_x, J_y, \delta) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_y} = \frac{1}{2\pi} (\mu_y + 3k\beta_y D\delta)$$

and the change of path length:

$$\Delta s = -\frac{\partial h_{eff}}{\partial \delta} = \alpha_c \delta - 3kD^3 \delta^2 - 3kD(\beta_x J_x - \beta_y J_y)$$



Normal forms - non-linear case

Assume a linear rotation (as always) followed by an octupole, the Hamiltonian is (1D to keep it simple):

$$\mathcal{H} = \frac{\mu}{2}(x^2 + p^2) + k_3 \cdot \frac{x^4}{4} \quad (p = p_x)$$

▣ With the Hamilton's equation leading to:

$$\dot{x} = \frac{\partial \mathcal{H}}{\partial p} = \mu p$$

$$\dot{p} = -\frac{\partial \mathcal{H}}{\partial x} = -\mu x - k_3 \cdot x^3$$



Normal forms - non-linear case

The map, written in Lie representation is:

$$\mathcal{M} = e^{-\frac{\mu}{2}:x^2+p^2:} e^{:k_3 \cdot \frac{x^4}{4}:} = Re^{:k_3 \cdot \frac{x^4}{4}:}$$


we transform by applying:

$$\begin{aligned} \mathcal{N} &= \mathcal{A}\mathcal{M}\mathcal{A}^{-1} = e^{:F:} Re^{:k_3 \cdot \frac{x^4}{4}:} e^{-:F:} = RR^{-1} e^{:F:} Re^{:k_3 \cdot \frac{x^4}{4}:} e^{-:F:} \\ &= Re^{:R^{-1}F+k_3 \cdot \frac{x^4}{4}-F:+O(\epsilon^2)} = Re^{:(R^{-1}-1)F+k_3 \cdot \frac{x^4}{4}:+O(\epsilon^2)} \end{aligned}$$

we have now to choose F to simplify the expression:

$$= (R^{-1} - 1)F + k_3 \cdot \frac{x^4}{4}$$

and get [EF1, AW]:

$$F = -\frac{1}{64} \{-5x^4 + 3p^4 + 6x^2p^2 + x^3p \cdot (8\cot(\mu) + 4\cot(2\mu)) + xp^3(8\cot(\mu) - 4\cot(2\mu))\}$$


Normal forms - non-linear case

We go back to x and p coordinates and with:

$$J = (x^2 + p^2)/2$$

we can write the map:


$$M = e^{-:F:} e^{-\mu J + \frac{3}{8} k_3 \cdot J^2} e{:F:}$$

the term $\frac{3}{8} k_3 \cdot J^2$ produces the tune shift with amplitude we know for an octupole ($\cdot < \beta^2 >$ in real space)

Note: the normalized map (our most simple map):

$$R = e^{-\mu J + \frac{3}{8} k_3 \cdot J^2}$$

is again a rotation in phase space, but the rotation angle now depends on the amplitude **J**



For the tune shift: octupole (1D)

When we have h_{eff} in 1D for a single octupole (see before):

$$h_{eff} = -\mu J + \frac{3}{8}k_3 \cdot J^2$$

$$Q_x(J_x, J_y) = \frac{1}{2\pi} \frac{\partial h_{eff}}{\partial J_x} = -\frac{1}{2\pi} \mu_x + \frac{3}{8 \cdot 2\pi} k_3 J_x$$

and with normalization in real space (i.e. $\beta \neq 1$):

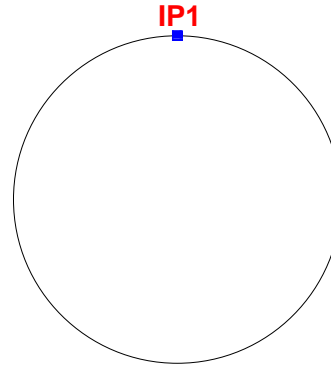
$$\Delta Q_x(J_x, J_y) = \frac{3}{8 \cdot 2\pi} k_3 \langle \beta^2 \rangle J_x$$

Example: $\beta = 300\text{m}$, $k_3 = 0.01$

$$\Delta Q_x(J_x, J_y) = 53.7 \cdot J_x$$



A real life example: beam-beam interaction



- Localized distortion, very strong non-linearity
- Standard perturbation theory not appropriate



Effect on invariants - start with single IP

➤ Look for invariant h


➤ Linear transfer $e^{i f_2}$ and beam-beam interaction $e^{i F}$, i.e.:

$$e^{i f_2} \cdot e^{i F} = e^{i h}$$

with (see before):

$$f_2 = -\frac{\mu}{2} \left(\frac{x^2}{\beta} + \beta p_x^2 \right)$$

and (see before):

$$F = \int_0^x dx' f(x')$$


Effect on invariants

For a Gaussian beam we have for $f(x)$ (see lecture on beam-beam effects):

$$f(x) = \frac{2}{x} \left(1 - e^{-\frac{x^2}{2\sigma^2}}\right)$$

as usual go to action angle variables Ψ, J :

$$x = \sqrt{2J\beta} \sin \Psi, \quad p = \sqrt{\frac{2J}{\beta}} \cos \Psi$$

and write $F(x)$ as Fourier series:

$$F(x) = \sum_{n=-\infty}^{\infty} c_n(J) e^{in\Psi}$$



We need:

REMEMBER: with this transform f_2 becomes very simple:

$$f_2 = -\mu J$$

and useful properties of Lie operators (any textbook²⁾):

$$: f_2 : g(J) = 0, \quad : f_2 : e^{in\Psi} = in\mu e^{in\Psi}, \quad g(: f_2 :) e^{in\Psi} = g(in\mu) e^{in\Psi}$$

and the formula (any textbook²⁾):

$$e{:f_2}:} e{:F}:} = e{:h}:} = \exp \left[: f_2 + \left(\frac{: f_2 :}{1 - e^{-:f_2:}} \right) F + \mathcal{O}(F^2) : \right]$$

²⁾ E. Forest, "Beam Dynamics, A New Attitude and Framework", 1998

Single IP

gives immediately for h :

$$h = -\mu J + \sum_n c_n(J) \frac{in\mu}{1 - e^{-in\mu}} e^{in\Psi}$$

$$h = -\mu J + \sum_n c_n(J) \frac{n\mu}{2\sin(\frac{n\mu}{2})} e^{(in\Psi + i\frac{n\mu}{2})}$$

away from resonance normal form transformation gives:

$$h_n = -\mu J + c_0(J) = \text{const.}$$

$$\left[\text{homework : } \frac{dc_0(J)}{dJ} \right]$$


Single IP - analysis of h

$$h = -\mu J + \sum_n c_n(J) \frac{n\mu}{2\sin(\frac{n\mu}{2})} e^{(in\Psi + i\frac{n\mu}{2})}$$

On resonance:

$$Q = \frac{p}{n} = \frac{\mu}{2\pi}$$

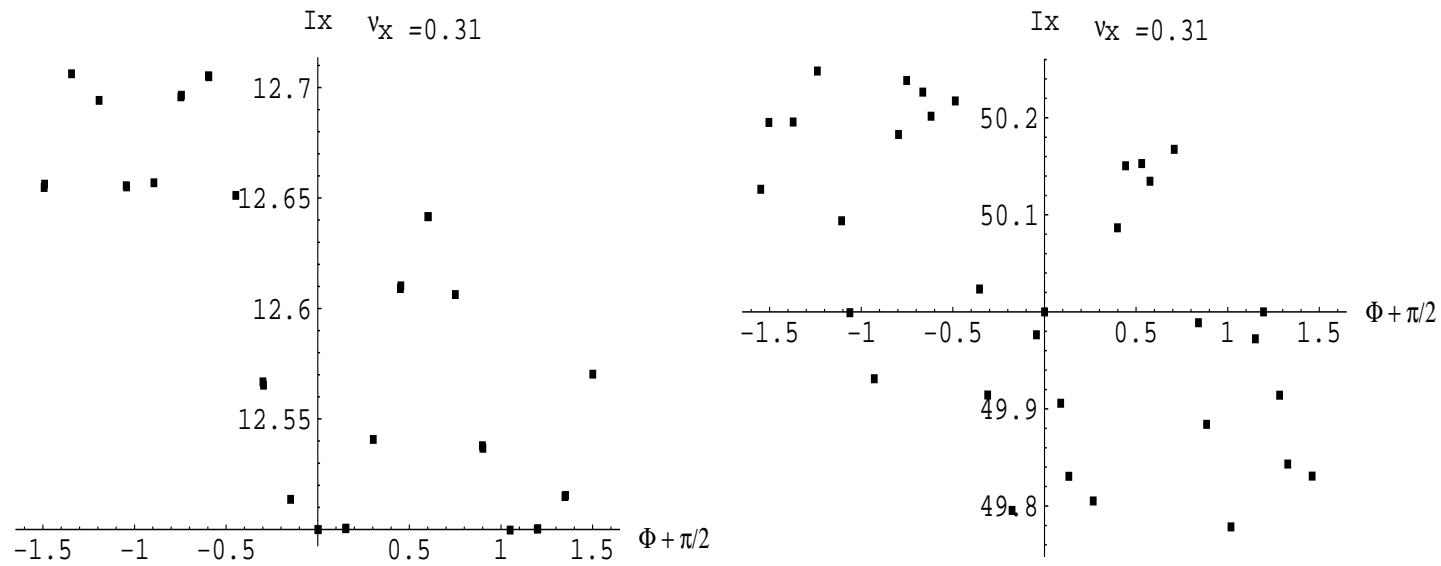
with $c_n \neq 0$:

$$\sin\left(\frac{n\pi p}{n}\right) = \sin(p\pi) \equiv 0 \quad \forall \text{ integer } p$$

and h diverges, find automatically all resonance conditions



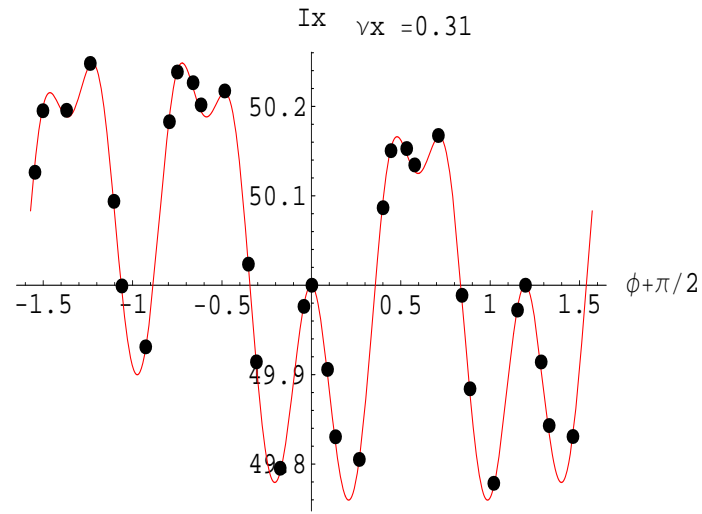
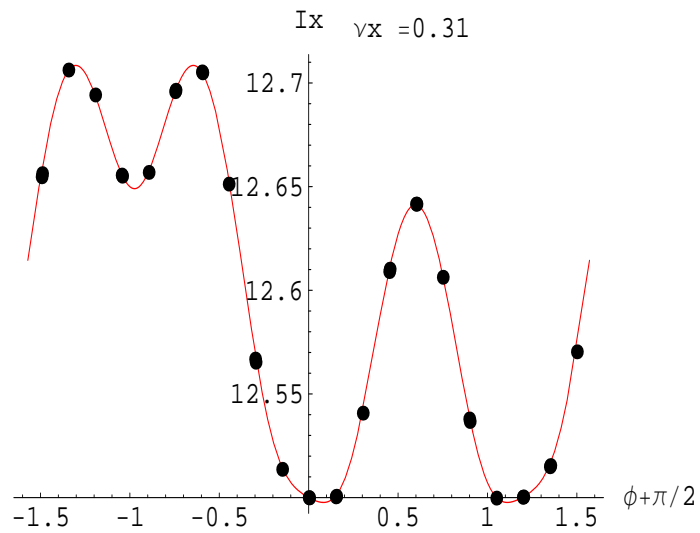
Invariant from tracking: one IP



→ Shown for $5\sigma_x$ and $10\sigma_x$



Invariant versus tracking: one IP

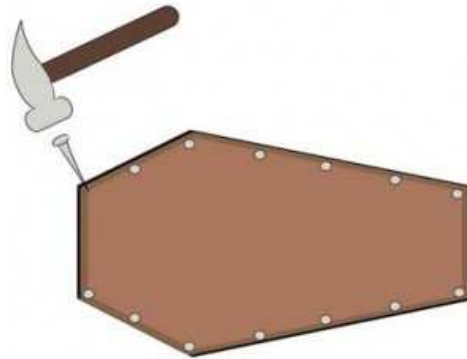


→ Shown for $5\sigma_x$ and $10\sigma_x$

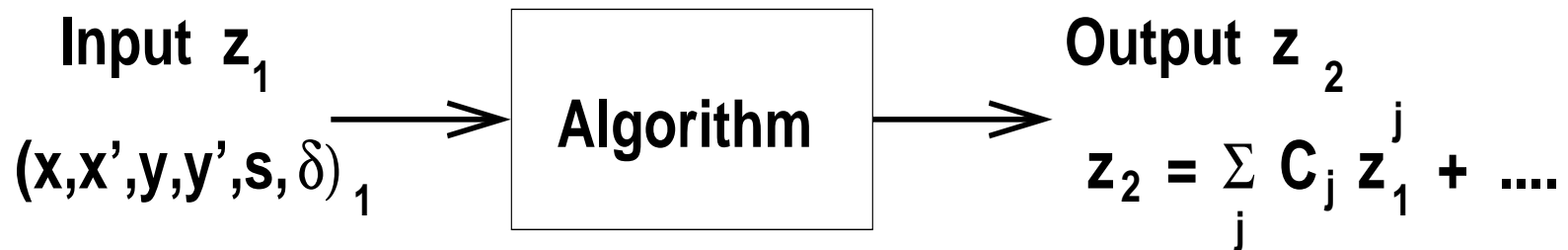


Truncated Power Series Algebra (TPSA)

- Tracking particles is very reliable method
- Simulation can produce maps for complicated configurations
- How can we analyse the map produced by a tracking code ?
- Now we put the final nail into the coffin of any other approach ...



Truncated Power Series Algebra



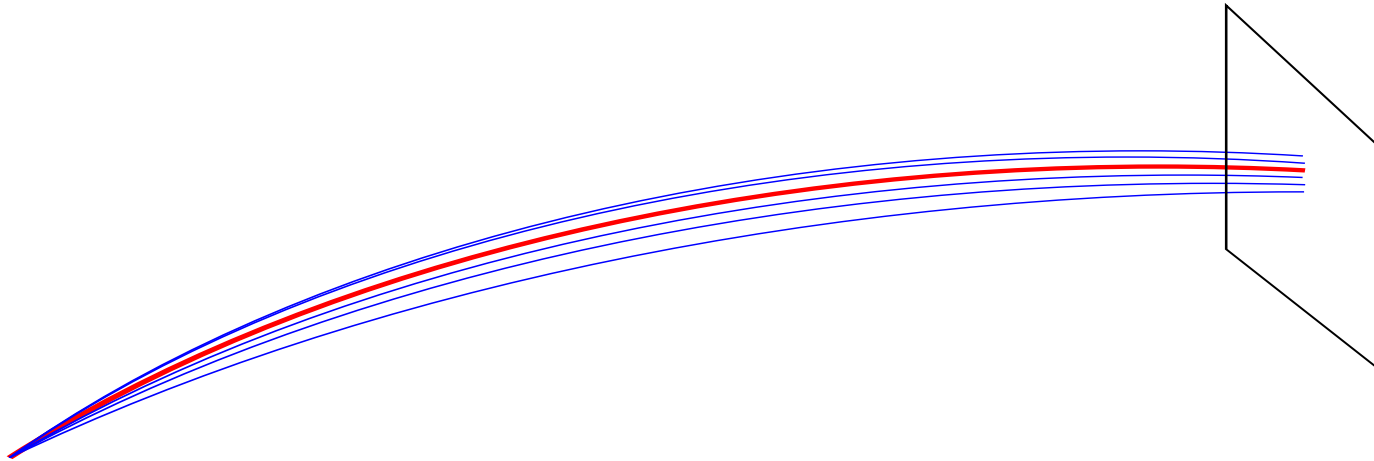
- The tracking of a complicated system relates the output **numerically** to the input
- Could we imagine something that relates the output **algebraically** to the input ?
- For example a Taylor series ?

$$z_2 = \sum C_j z_1^j = \sum f^{(n)} z_1^j$$



Why are Taylor series useful ?

Let us study the paraxial behaviour:



- Red line is the ideal orbit
- Blue lines are small deviations
- If we understand how small deviations behave, we understand the system much better



Why are Taylor series useful ?

Now remember the definition of the Taylor series:

$$f(a + \Delta x) = f(a) + \sum_{n=1}^{\infty} \frac{f^{(n)}(a)}{n!} \Delta x^n$$

$$f(a + \Delta x) = f(a) + \frac{f'(a)}{1!} \Delta x^1 + \frac{f''(a)}{2!} \Delta x^2 + \frac{f'''(a)}{3!} \Delta x^3 + \dots$$

- The coefficients determine the behaviour of small deviations Δx from the ideal orbit x
- The Taylor expansion does a paraxial analysis of the system


Why are Taylor series useful ?

If the function $f(x)$ is represented by a Taylor series:

$$f(a + \Delta x) = f(a) + \sum_{n=1}^{\infty} \frac{f^{(n)}(a)}{n!} \Delta x^n$$

if it is truncated to the m -th order:

$$f(a + \Delta x) = f(a) + \sum_{n=1}^m \frac{f^{(n)}(a)}{n!} \Delta x^n$$

- There is an equivalence between the function $f(x)$ and the vector $(f(a), f'(a), f''(a), \dots, f^{(m)}(a))$
 - This vector is a **Truncated Power Series Algebra (TPSA)** representation of $f(x)$ around a
 - How to get these coefficients without extra work ?
- 

Numerical differentiation

The problem getting the derivatives $f^{(n)}(a)$ of $f(x)$ at a :

$$f'(a) = \frac{f(a + \epsilon) - f(a)}{\epsilon}$$

- Need to subtract almost equal numbers and divide by small number.
- For higher orders f'' , f''' .., accuracy hopeless !
- We can use Differential Algebra (DA)
(M. Berz, 1988 and [\[MB\]](#))



Differential Algebra

1. Define a pair (q_0, q_1) , with q_0, q_1 real numbers



Differential Algebra

1. Define a pair (q_0, q_1) , with q_0, q_1 real numbers
2. Define operations on a pair like:

$$(q_0, q_1) + (r_0, r_1) = (q_0 + r_0, q_1 + r_1)$$

$$c \cdot (q_0, q_1) = (c \cdot q_0, c \cdot q_1)$$

$$(q_0, q_1) \cdot (r_0, r_1) = (q_0 \cdot r_0, q_0 \cdot r_1 + q_1 \cdot r_0)$$



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$$(q_0, q_1) \cdot (r_0, r_1) = (q_0 \cdot r_0, q_0 \cdot r_1 + q_1 \cdot r_0)$$

3. And some ordering:

$$(q_0, q_1) < (r_0, r_1) \quad \text{if} \quad q_0 < r_0 \quad \text{or} \quad (q_0 = r_0 \quad \text{and} \quad q_1 < r_1)$$

$$(q_0, q_1) > (r_0, r_1) \quad \text{if} \quad q_0 > r_0 \quad \text{or} \quad (q_0 = r_0 \quad \text{and} \quad q_1 > r_1)$$



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4. This implies something strange:

$$(0, 0) < (0, 1) < (r, 0) \quad (\text{for any pos. } r)$$

$$(0, 1) \cdot (0, 1) = (0, 0) \quad \rightarrow \quad (0, 1) = \sqrt{(0, 0)}!!$$



Differential Algebra

This means that $(0,1)$ is between 0 and ANY real number
→ infinitely small !!!

We call this therefore "differential unit" $d = (0, 1) = \delta$.

Of course $(q, 0)$ is just the real number q and we define "real part" and "differential part" (a bit like complex numbers..):

$$q_0 = \mathcal{R}(q_0, q_1) \quad \text{and} \quad q_1 = \mathcal{D}(q_0, q_1)$$

With our rules we can further see that:

$$(1, 0) \cdot (q_0, q_1) = (q_0, q_1)$$

$$(q_0, q_1)^{-1} = \left(\frac{1}{q_0}, -\frac{q_1}{q_0^2} \right)$$



Differential Algebra

Of course can let a function f act on the pair (or vector) using our rules.

For example:

$$f(x) \rightarrow f(x, 0)$$

acts like the function f on the real variable x :

$$f(x) = \mathcal{R}[f(x, 0)]$$

What about the differential part \mathcal{D} ?



Differential Algebra

For a function $f(x)$ without proof:

$$\mathcal{D}[f(x + d)] = \mathcal{D}[f((x, 0) + (0, 1))] = \mathcal{D}[f(x, 1)] = f'(x)$$

An example instead:

$$f(x) = x^2 + \frac{1}{x}$$

then using school calculus:

$$f'(x) = 2x - \frac{1}{x^2}$$

For $x = 2$ we get then:

$$f(2) = \frac{9}{2}, f'(2) = \frac{15}{4}$$



Differential Algebra


For x in:

$$f(x) = x^2 + \frac{1}{x}$$

we substitute: $x \rightarrow (x, 1) = (2, 1)$ and use our rules:

$$\begin{aligned} f[(2, 1)] &= (2, 1)^2 + (2, 1)^{-1} \\ &= (4, 4) + \left(\frac{1}{2}, -\frac{1}{4}\right) \\ &= \left(\frac{9}{2}, \frac{15}{4}\right) = (f(2), f'(2)) \quad !!! \end{aligned}$$

The computation of derivatives becomes an algebraic problem, no need for small numbers, exact !



Differential Algebra - higher orders

1. The pair $(q_0, 1)$, becomes a vector of order N :

$$(q_0, 1) \rightarrow (q_0, 1, 0, 0, \dots, 0) \quad \delta = (0, 1, 0, 0, 0, \dots)$$

2. $(q_0, q_1, q_2, \dots, q_N) + (r_0, r_1, r_2, \dots, r_N) = (s_0, s_1, s_2, \dots, s_N)$

with: $s_i = q_i + r_i$

3. $c \cdot (q_0, q_1, q_2, \dots, q_N) = (c \cdot q_0, c \cdot q_1, c \cdot q_2, \dots, c \cdot q_N)$

4. $(q_0, q_1, q_2, \dots, q_N) \cdot (r_0, r_1, r_2, \dots, r_N) = (s_0, s_1, s_2, \dots, s_N)$

with:

$$s_i = \sum_{k=0}^i \frac{i!}{k!(i-k)!} q_k r_{i-k}$$



Differential Algebra

If we had started with:

$$x = (a, 1, 0, 0, 0\dots)$$

we would get:


$$f(x) = (f(a), f'(a), f''(a), f'''(a), \dots f^{(n)}(a))$$

can be extended to more variables x, y :

$$x = (a, 1, 0, 0, 0\dots) \quad dx = (0, 1, 0, 0, 0, \dots)$$

$$y = (b, 0, 1, 0, 0\dots) \quad dy = (0, 0, 1, 0, 0, \dots)$$

and get (with more complicated multiplication rules):

$$f((x + dx), y + dy) = \left(f, \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial^2 f}{\partial x^2}, \frac{\partial^2 f}{\partial x \partial y}, \dots \right) (x, y)$$


What is the use of that:



Can extract a truncated Taylor map of a beam line or ring by pushing the identity map $f(x) = (a, 1, 0, 0, 0\dots)$ through the algorithm as if it is a vector in phase space !

The maps are provided with the desired accuracy and to any order.



What is the use of that:



- "Algorithm" can be a mathematical function
- "Algorithm" can be a complex computer code
- Easy using polymorphism of modern languages (see example)
- Normal form analysis on Taylor series is much easier !!
- We get a Taylor map for a computer code !!!



What is the use of that:

■ Demonstrate with simple examples (FORTRAN 95):

➤ First show the concept

➤ Simple FODO cell

➤ Normal form analysis of the FODO cell with octupoles

■ All examples and all source code in:

Website: <http://cern.ch/Werner.Herr/CAS2013/DA>

Small DA package provided by E. Forest

Look at this small example:

```
PROGRAM DATEST1
use my_own_da
real(8) x,z, dx
my_order=3
dx=0.0
x=3.141592653/6.0 + dx
call track(x, z)
call print(z,6)
END PROGRAM DATEST1
```

```
SUBROUTINE TRACK(a, b)
use my_own_da
real(8) a,b
b = sin(a)
END SUBROUTINE TRACK
```

```
PROGRAM DATEST2
use my_own_da
type(my_taylor) x,z, dx
my_order=3
dx=1.0.mono.1 ! this is our (0,1)
x=3.141592653/6.0 + dx
call track(x, z)
call print(z,6)
END PROGRAM DATEST2
```

```
SUBROUTINE TRACK(a, b)
use my_own_da
type(my_taylor) a,b
b = sin(a)
END SUBROUTINE TRACK
```

Look at the results:

(0,0)	0.500000000000000E+00
(1,0)	0.8660254037844E+00
(0,1)	0.000000000000000E+00
(2,0)	-0.250000000000000E+00
(0,2)	0.000000000000000E+00
(1,1)	0.000000000000000E+00
(3,0)	-0.1443375672974E+00
(0,3)	0.000000000000000E+00
(2,1)	0.000000000000000E+00
(1,2)	0.000000000000000E+00

(0,0) 0.500000000000000E+00


We have $\sin(\frac{\pi}{6}) = 0.5$ all right, but what is the rest ??



Look at the results:

(0,0) 0.500000000000000E+00
(1,0) 0.8660254037844E+00
(0,1) 0.000000000000000E+00
(2,0) -0.250000000000000E+00
(0,2) 0.000000000000000E+00
(1,1) 0.000000000000000E+00
(3,0) -0.1443375672974E+00
(0,3) 0.000000000000000E+00
(2,1) 0.000000000000000E+00
(1,2) 0.000000000000000E+00

(0,0) 0.500000000000000E+00

$$\sin\left(\frac{\pi}{6} + \Delta x\right) = \sin\left(\frac{\pi}{6}\right) + \cos\left(\frac{\pi}{6}\right)\Delta x^1 - \frac{1}{2}\sin\left(\frac{\pi}{6}\right)\Delta x^2 - \frac{1}{6}\cos\left(\frac{\pi}{6}\right)\Delta x^3$$


What is the use of that:

- We have used a simple algorithm here (*sin*) but it can be **anything** very complex
- We can compute nonlinear maps as a Taylor expansion of **anything** the program computes
- Simply by:
 - Replacing regular (e.g. REAL) types by TPSA types (*my_taylor*) i.e. variables x, p are automatically replaced by $(x, 1, 0, ..)$ and $(p, 0, 1, 0, ..)$ etc.
 - Operators and functions ($+, -, *, =, \dots, exp, sin, \dots$) automatically overloaded, i.e. behave according to new type

What is the use of that:

Assume the *Algorithm* describes one turn, then:

■ Normal tracking:

➤ $X_n = (x, p_x, y, p_y, s, \delta)_n \rightarrow X_{n+1} = (x, p_x, y, p_y, s, \delta)_{n+1}$

➤ Coordinates after one completed turn

■ TPSA tracking:

➤ $X_n = (x, p_x, y, p_y, s, \delta)_n \rightarrow X_{n+1} = \sum C_j X_n^j$

➤ Taylor expansion after one completed turn

➤ Automatically all X_{n+1} where it converges

➤ The C_j contain useful information about behaviour

➤ Taylor map directly used for normal form analysis



Another example:

➤ Track through a FODO lattice:

QF - DRIFT - QD

Integrate 100 steps in the quadrupoles

Now we use **three** variables:

$x, p, \Delta p = (z(1), z(2), z(3))$

Another example:

```
program fodol
use my_own_da
use my_analysis
type(my_taylor) z(3)
type(normalform) NORMAL
type(my_map) M,id

real(dp) L,DL,k1,k3,fix(3)
integer i,nstep



my_order=4 ! maximum order 4
fix=0.0 ! fixed point
id=1
z=fix+id

LC=62.5 ! half cell length
L=3.0 ! quadrupole length
nstep=100
DL=L/nstep
k1=0.003 ! quadrupole strength

do i=1,nstep ! track through quadrupole
z(1)=z(1)+DL/2*z(2)
z(2)=z(2)-k1*DL*z(1)/(1 + z(3))
z(1)=z(1)+DL/2*z(2)
enddo
z(1)=z(1)+LC*z(2) ! drift of half cell
length

do i=1,nstep ! track through quadrupole
z(1)=z(1)+DL/2*z(2)
z(2)=z(2)-k1*DL*z(1)/(1 + z(3))
z(1)=z(1)+DL/2*z(2)
enddo
z(1)=z(1)+LC*z(2) ! drift of half cell
length

call print(z(1),6)
call print(z(2),6)
M=z
NORMAL=M
write(6,*) normal%tune, normal%dtune_da
end program fodol
```

 2) Courtesy E. Forest for the small DA package used here ... 

The result is:

Only linear elements in the Taylor expansion, the result for the matrix per cell:

(0,0,0) 0.9442511679729E-01
(0,0,1) -0.9729519276183E-01

(1,0,0) 0.6972061935061E-01
(0,1,0) 0.1677727932585E+03
(1,0,1) 0.1266775134236E+01
(0,1,1) -0.3643444875882E+02
(1,0,2) -0.1603248617779E+01
(0,1,2) 0.3609522079691E+02
(1,0,3) 0.1939697138318E+01
(0,1,3) -0.3575511053483E+02

(1,0,0) -0.5300319873866E-02
(0,1,0) 0.1588490329398E+01
(1,0,1) 0.1060055415702E-01
(0,1,1) -0.5832024543075E+00
(1,0,2) -0.1590066005419E-01
(0,1,2) 0.5779004431627E+00
(1,0,3) 0.2120059477024E-01
(0,1,3) -0.5725843143370E+00

$$\Delta x_f = 0.06972 \Delta x_i + 167.77 \Delta p_i$$

$$\Delta p_f = -0.00530 \Delta x_i + 1.5885 \Delta p_i$$

The output from the normal form analysis are (per cell !):

Tune = 0.094425

Chromaticity = -0.097295

Modified previous example (with one octupole):

```
program fodo3
use my_own_da
use my_analysis
type(my_taylor) z(3)
type(normalform) NORMAL
type(my_map) M,id

real(dp) L,DL,k1,k3,fix(3)
integer i,nstep

my_order=4 ! maximum order 4
fix=0.0 ! fixed point
id=1
z=fix+id

LC=62.5 ! half cell length
L=3.0 ! quadrupole length
nstep=100
DL=L/nstep
k1=0.003 ! quadrupole strength
k3=0.01 ! octupole strength

do i=1,nstep ! track through quadrupole
z(1)=z(1)+DL/2*z(2)
z(2)=z(2)-k1*DL*z(1)/(1 + z(3))
z(1)=z(1)+DL/2*z(2)
enddo
z(2)=z(2)-k3*z(1)**3/(1 + z(3)) ! octupole
kick !!!
z(1)=z(1)+LC*z(2) ! drift of half cell
length

do i=1,nstep ! track through quadrupole
z(1)=z(1)+DL/2*z(2)
z(2)=z(2)-k1*DL*z(1)/(1 + z(3))
z(1)=z(1)+DL/2*z(2)
enddo

z(1)=z(1)+LC*z(2) ! drift of half cell
length

call print(z(1),6)
call print(z(2),6)
M=z
NORMAL=M
write(6,*) normal%tune, normal%dtune_da
end program fodo3
```

The result is:

```
(0,0,0) 0.9442511679729E-01
(0,0,1) -0.9729519276183E-01

(2,0,0) 0.5374370086899E+02
(0,2,0) 0.5374370086899E+02
(0,0,2) 0.1018391758451E+00
(2,0,1) 0.2035776281196E+02
.....

(1,0,0) 0.6972061935061E-01
(0,1,0) 0.1677727932585E+03
(1,0,1) 0.1266775134236E+01
(0,1,1)-0.3643444875882E+02
(3,0,0)-0.1586519461687E+01
(2,1,0)-0.1440953324752E+02
(1,2,0)-0.4362477179879E+02

.....

(1,0,0)-0.5300319873866E-02
(0,1,0) 0.1588490329398E+01
(1,0,1) 0.1060055415702E-01
(0,1,1)-0.5832024543075E+00
(3,0,0)-0.1519218878892E-01
```

Now non-linear elements in the Taylor expansion,

The output from the normal form analysis are (per cell !):

Tune = **0.094425**

Chromaticity= **-0.097295**

The detuning with amplitude is **53.74 !**

Modified previous example (with octupole):

Remember the normal form transformation:

$$A\mathcal{M}A^{-1} = \mathcal{R}$$

The type **normalform** in the demonstration package also contains the maps A and \mathcal{R} !

$$j2 = (x^{**2} + p^{**2}) * \text{NORMAL} \% A^{**}(-1)$$

(remember: $x^{**2} + p^{**2}$ is the tilted ellipse

Can get the optical functions out because

■ β : coefficient of p^{**2} of invariant $j2$

■ α : coefficient of x^*p of invariant $j2$

■ γ : coefficient of x^{**2} of invariant $j2$

Modified previous example (with octupole):

In our code use like :

$$\beta = \text{j2.sub.beta}$$

$$\alpha = 0.5 * \text{j2.sub.twoalpha}$$

$$\gamma = \text{j2.sub.gamma}$$

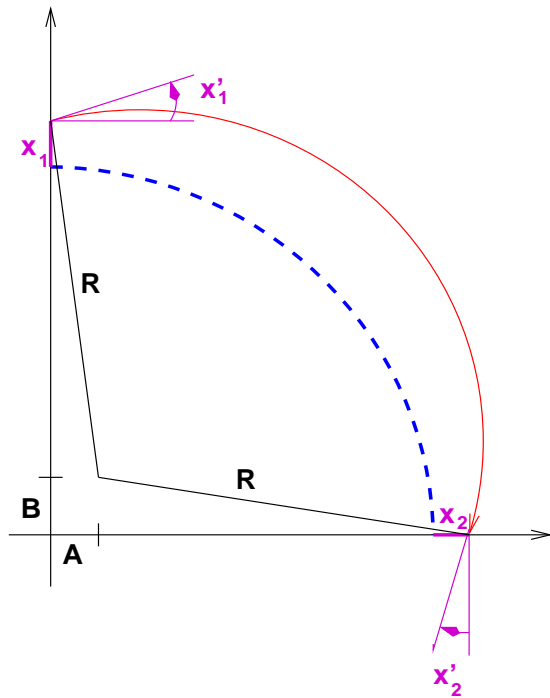
we obtain (here at the end of the cell):

beta, alpha, gamma

300.080714 -1.358246 9.480224E-003

This was trivial - now a (normally) hard one

The exact map:



$$p_2 = \sin(x'_2) = -\frac{B}{R}$$

$$x_2 = A - R(1 - \cos(x'_2)) = A - R(1 - \sqrt{1 - p_2^2})$$

$$A = R \cdot p_1 = R \cdot \sin(x'_1)$$

$$B = R(1 - \cos(x'_1)) + x_1 = R(1 - \sqrt{1 - p_1^2}) + x_1$$

A 90° bending magnet ..



How to apply Differential Algebra here ...

➤ Start with initial coordinates in DA style:

$$x_1 = (0, 1, 0, \dots)$$

$$p_1 = (0, 0, 1, \dots) \quad \text{and have:}$$

$$A = (0, 0, R, 0, \dots)$$

$$B = (0, 1, 0, 0, 0, R, 0, \dots)$$

➤ After pushing them through the algorithm:

$$\rightarrow x_2 = (0, 0, R, -\frac{1}{R}, 0, 0, 0\dots) = (0, \frac{\partial x_2}{\partial x_1}, \frac{\partial x_2}{\partial p_1}, \frac{\partial^2 x_2}{\partial x_1^2}, \frac{\partial^2 x_2}{\partial x_1 \partial p_1}, \dots)$$

$$\rightarrow p_2 = (0, -\frac{1}{R}, 0, 0, 0, -1, 0\dots) = (0, \frac{\partial p_2}{\partial x_1}, \frac{\partial p_2}{\partial p_1}, \frac{\partial^2 p_2}{\partial x_1^2}, \frac{\partial^2 p_2}{\partial x_1 \partial p_1}, \dots)$$

➤ Automatically evaluates all non-linearities to any desired order ..



How to apply Differential Algebra here ...

➤ Start with initial coordinates in DA style:

$$x_1 = (0, 1, 0, \dots)$$

$$p_1 = (0, 0, 1, \dots) \quad \text{and have:}$$

$$A = (0, 0, R, 0, \dots)$$

$$B = (0, 1, 0, 0, 0, R, 0, \dots)$$

➤ After pushing them through the algorithm:

$$\rightarrow x_2 = (0, 0, R, -\frac{1}{R}, 0, 0, 0\dots) = (0, \frac{\partial x_2}{\partial x_1}, \frac{\partial x_2}{\partial p_1}, \frac{\partial^2 x_2}{\partial x_1^2}, \frac{\partial^2 x_2}{\partial x_1 \partial p_1}, \dots)$$

$$\rightarrow p_2 = (0, -\frac{1}{R}, 0, 0, 0, -1, 0\dots) = (0, \frac{\partial p_2}{\partial x_1}, \frac{\partial p_2}{\partial p_1}, \frac{\partial^2 p_2}{\partial x_1^2}, \frac{\partial^2 p_2}{\partial x_1 \partial p_1}, \dots)$$

➤ Automatically evaluates all non-linearities to any desired order ..



Some we know ...

Transfer matrix of a dipole:

$$M_{dipole} = \begin{pmatrix} \cos(\frac{L}{R}) & R\sin(\frac{L}{R}) \\ -\frac{1}{R}\sin(\frac{L}{R}) & \cos(\frac{L}{R}) \end{pmatrix} = \begin{pmatrix} \frac{\partial x_2}{\partial x_1} & \frac{\partial x_2}{\partial p_1} \\ \frac{\partial p_2}{\partial x_1} & \frac{\partial p_2}{\partial p_1} \end{pmatrix}$$

For a 90° bending angle we get:

$$M_{dipole} = \begin{pmatrix} 0 & R \\ -\frac{1}{R} & 0 \end{pmatrix}$$

as computed, but we also have **all** derivatives and non-linear effects !



What is the use of that:

- Although not strictly an analytic method in the traditional sense:
 - TPSA provide analytic expression (Taylor series) for the one turn map
 - Can be used for tracking
 - Can be analysed for dynamic behaviour of the system
 - Typical use: Normal Form Analysis discussed earlier, rather straightforward from a Taylor expansion



Is there a summary ?

$$m = z$$

$$\text{NORMAL} = m$$

- Get the map m somehow (no matter how)
- Analyse this map (Normal form)



And another summary

- Perturbation treatment limited to:
 - Small perturbations (not in real machines)
 - Pedagogical purpose
- For realistic machines symplectic, iterative mapping is appropriate, using:
 - Symplectic integration
 - Lie transformations and normal form analysis
 - Differential algebra



Back up

- backup slides -



Example: sextupole (1D)

Given the Hamiltonian h :

$$h = -\mu J - \frac{3}{8}k(2\beta J)^{3/2} \cdot \left(\frac{\sin(3\Psi + \frac{3\mu}{2})}{\sin\frac{3\mu}{2}} - \frac{\sin(\Psi + \frac{\mu}{2})}{\sin\frac{\mu}{2}} \right)$$

particles move in phase space along constant h .

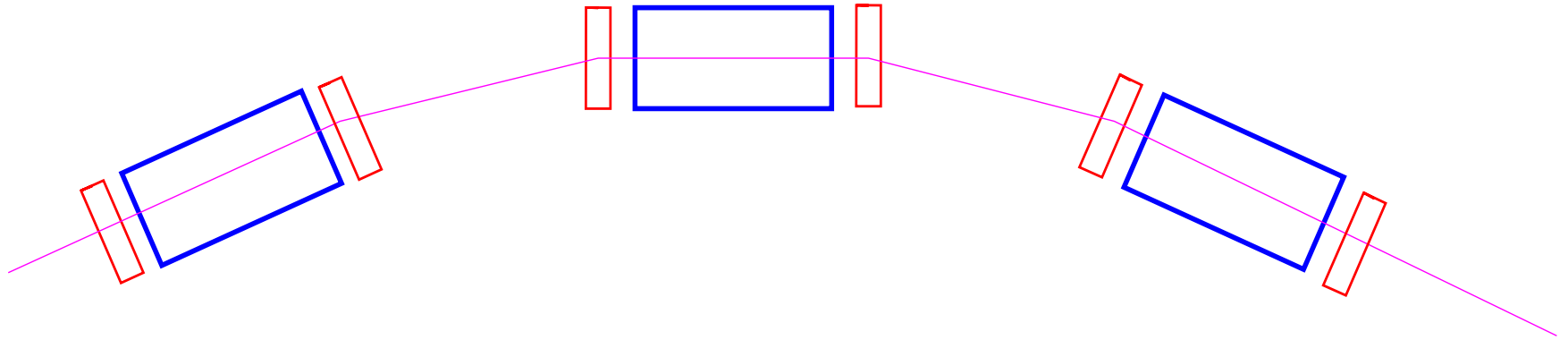
Back to Cartesian coordinates we get for h :

$$h = -\frac{\mu}{2}(x^2 + x'^2) \frac{3}{8}\mu\beta^{3/2}x[(3x'^2 - x^2)\cot\frac{3\mu}{2} - (x^2 + x'^2)\cot\frac{\mu}{2} - 4xx']$$

Constant h defines the trajectory in phase space !



Where to put the elements in an accelerator ?



$$\frac{d^2x}{ds^2} + K(s)x = 0$$

- Usually use **s** (pathlength) along "reference path"
- "Reference path" defined geometrically by straight sections and bending magnets

Second order MAPS concatenation

Assume now 2 maps of second order:


$$\mathcal{A}_2 = [R^A, T^A] \quad \text{and} \quad \mathcal{B}_2 = [R^B, T^B]$$

the combined second order map

$$\mathcal{C}_2 = \mathcal{A}_2 \circ \mathcal{B}_2 \quad \text{is} \quad \mathcal{C}_2 = [R^C, T^C] \quad \text{with:}$$

$$R^C = R^A \cdot R^B$$

and (after truncation of higher order terms !!):

$$T_{ijk}^C = \sum_{l=1}^4 R_{il}^B T_{ljk}^A + \sum_{l=1}^4 \sum_{m=1}^4 T_{ilm}^B R_{lj}^A R_{mk}^A$$


Symplecticity for higher order MAPS

try truncated Taylor map in 2D, second order:

$$\begin{pmatrix} x \\ x' \end{pmatrix} = \begin{pmatrix} R_{11}x_0 + R_{12}x'_0 + T_{111}x_0^2 + T_{112}x_0x'_0 + T_{122}x_0'^2 \\ R_{21}x_0 + R_{22}x'_0 + T_{211}x_0^2 + T_{212}x_0x'_0 + T_{222}x_0'^2 \end{pmatrix}$$

The Jacobian becomes:

$$\mathcal{J} = \begin{bmatrix} R_{11} + 2T_{111}x_0 + T_{112}x'_0 & R_{12} + T_{112}x_0 + 2T_{122}x'_0 \\ R_{21} + 2T_{211}x_0 + T_{212}x'_0 & R_{22} + T_{212}x_0 + 2T_{222}x'_0 \end{bmatrix}$$

symplecticity condition requires that:

$\det \mathcal{J} = 1$ for all x_0 and all x'_0



Symplecticity for higher order MAPS

This is only possible for the conditions:

$$\begin{pmatrix} R_{11}R_{22} - R_{12}R_{21} = 1 \\ R_{11}T_{212} + 2R_{22}T_{111} - 2R_{12}T_{211} - R_{21}T_{112} = 0 \\ 2R_{11}T_{222} + R_{22}T_{112} - R_{12}T_{212} - 2R_{21}T_{122} = 0 \end{pmatrix}$$

- 10 coefficients, but 3 conditions
- number of **independent** coefficients only 7 !
- Taylor map requires more coefficients than necessary
- e.g. 4D, order 4: coefficients 276 instead of 121



Canonical transformations

- With Hamiltonian's equations, still have to solve $(2n)$ differential equations
- Not necessarily easy, but:
 - More freedom to choose the variables q and p (because they have now "equal" status)
 - Try to find variables where they are easy to solve
- Change of variables through "canonical transformations"



Why canonical transformations ?

■ Hamiltonian have one advantage over Lagrangians:

- If the system has a symmetry, i.e. a coordinate q_i does not occur in H (i.e. $\frac{\partial H}{\partial q_i} = 0 \rightarrow \frac{dp_i}{dt} = 0$) → the corresponding momentum p_i is conserved (and the coordinate q_i can be ignored in the other equations of the set).
- Comes also from Lagrangian, but the velocities still occur in \mathcal{L} !



Canonical transformations

Starting with $H(q, p, t)$ get new coordinates:

$$Q_i = Q_i(q, p, t)$$

$$P_i = P_i(q, p, t)$$

and new Hamiltonian $K(Q, P, t)$ with:

$$\frac{\partial K}{\partial Q_j} = -\dot{P}_j = -\frac{dP_j}{dt}, \quad \frac{\partial K}{\partial P_j} = \dot{Q}_j = \frac{dQ_j}{dt}$$

■ We can two types of canonical transformations



Canonical transformations - type 1

- Ideally one would like a Hamiltonian H and coordinates with:

$$\frac{\partial H}{\partial q_j} = -\dot{p}_j = -\frac{dp_j}{dt} = 0$$

- Coordinate q_j not explicit in H
- p_j is a constant of the motion (!) and:

$$\frac{dq_j}{dt} = \frac{\partial H(p_1, p_2, \dots, p_n)}{\partial p_j} = F_j(p_1, p_2, \dots, p_n)$$

which can be directly integrated to get $q_j(t)$



Canonical transformations - type 1, example

Harmonic oscillator:

$$H = T + V = \frac{1}{2}mv^2 + \frac{m\omega^2}{2}x^2 = \frac{p^2}{2m} + \frac{m\omega^2}{2}x^2$$

try: $x = \sqrt{\frac{2P}{m\omega}} \cdot \sin(X)$ and $p = \sqrt{2m\omega P} \cdot \cos(X)$ and we get:

$$K = \omega P \cos^2(X) + \omega P \sin^2(X) = \omega P$$

then:

$$\frac{dX}{dt} = \frac{\partial K}{\partial P} = \omega \quad \rightarrow \quad X = \omega t + \alpha$$

back transformation to x, p :

$$x = \sqrt{\frac{2E}{m\omega^2}} \sin(\omega t + \alpha)$$



Canonical transformations - type 2

- Find a transformation of q, p at time t to values q_0, p_0 at time $t = 0$.

$$q = q(q_0, p_0, t)$$

$$p = p(q_0, p_0, t)$$

- The transformations **ARE** the solution of the problem !

For both types: how to find the transformation ?

- Without details: Hamilton-Jacobi equation ...



Extension: general monomials

Monomials in x and p of orders n and m ($x^n p^m$)

$$e^{:ax^n p^m:}$$

gives for the map (for $n \neq m$):

$$e^{:ax^n p^m:} x = x \cdot [1 + a(n - m)x^{n-1}p^{m-1}]^{m/(m-n)}$$

$$e^{:ax^n p^m:} p = p \cdot [1 + a(n - m)x^{n-1}p^{m-1}]^{n/(n-m)}$$

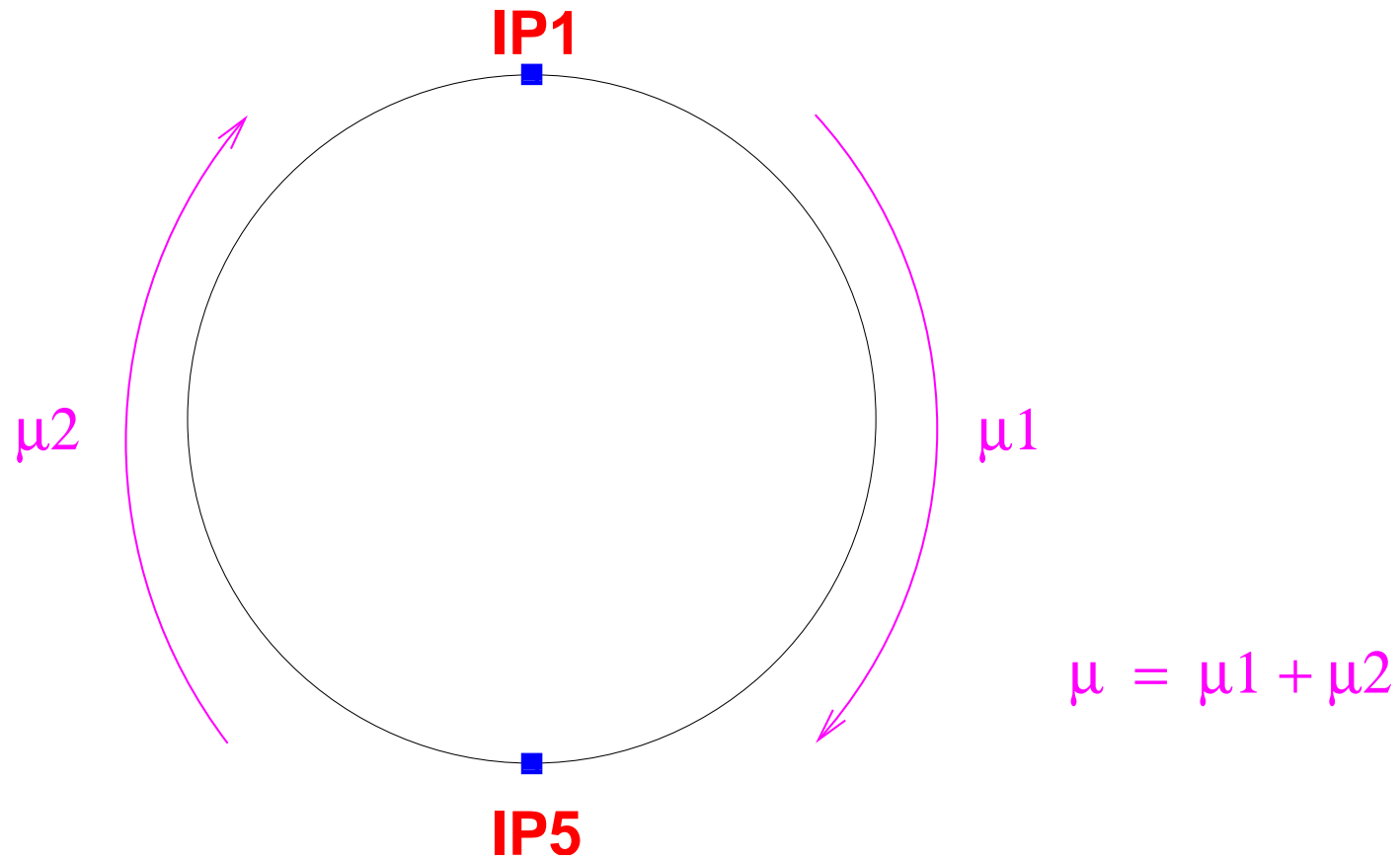
gives for the map (for $n = m$):

$$e^{:ax^n p^n:} x = x \cdot e^{-anx^{n-1}p^{n-1}}$$

$$e^{:ax^n p^n:} p = p \cdot e^{anx^{n-1}p^{n-1}}$$



Collision scheme - two IPs



Two IPs

→ two transfers f_2^1, f_2^2 and two beam-beam kicks F^1, F^2 ,
 first IP at μ_1 , second IP at μ :

$$\begin{aligned}
 &= e^{:f_2^1:} e^{:F^1:} e^{:f_2^2:} e^{:F^2:} = e^{:h_2:} \\
 &= e^{:f_2^1:} e^{:F^1:} e^{-:f_2^1:} e^{:f_2^1:} e^{:f_2^2:} e^{:F^2:} = e^{:h_2:} \\
 &= e^{:f_2^1:} e^{:F^1:} e^{-:f_2^1:} e^{:f_2:} e^{:F^2:} e^{-:f_2:} e^{:f_2:} = e^{:h_2:} \\
 &= e^{:e^{-:f_2^1:} F^1:} e^{:e^{-:f_2^2:} F^2:} e^{:f_2:} = e^{:h_2:}
 \end{aligned}$$

$$f_2 = -\mu A, \quad f_2^1 = -\mu_1 A, \quad \text{and} \quad f_2^2 = -\mu_2 A$$



Two IPs

here a miracle occurs (remember $g(: f_2 :)e^{in\Psi} = g(in\mu)e^{in\Psi}$):

$$e{:f_2^1}:e^{in\Psi} = e^{in\mu_1}e^{in\Psi} = e^{in(\mu_1+\Psi)}$$

i.e. the Lie transforms of the perturbations are phase shifted²). Therefore:

$$e{:e^{-:f_2^1}:F^1}:e{:e^{-:f_2}:F^2}:e{:f_2}: = e{:h_2}:$$

becomes simpler with substitutions of $\Psi_1 = \Psi + \mu_1$ and $\Psi = \Psi + \mu$ in F^1 and F :

$$e{:F^1(\Psi_1):}e{:F(\Psi):}e{:f_2}: \Rightarrow e{:F^1(\Psi_1)+F(\Psi):}e{:f_2}:$$

²) E. Forest, "Beam Dynamics, A New Attitude and Framework", 1998

Two IPs

gives for h_2 :

$$h_2 = -\mu A + \sum_{n=-\infty}^{\infty} \frac{n\mu c_n(A)}{2\sin(n\frac{\mu}{2})} e^{-in(\Psi + \mu/2 + \mu_1)} + e^{-in(\Psi + \mu/2)}$$

$$h_2 = -\mu A + 2c_0(A) + \underbrace{\sum_{n=1}^{\infty} \frac{2n\mu c_n(A)}{2\sin(n\frac{\mu}{2})} \cos(n(\Psi + \frac{\mu}{2} + \frac{\mu_1}{2})) \cos(n\frac{\mu_1}{2})}_{\text{interesting part}}$$

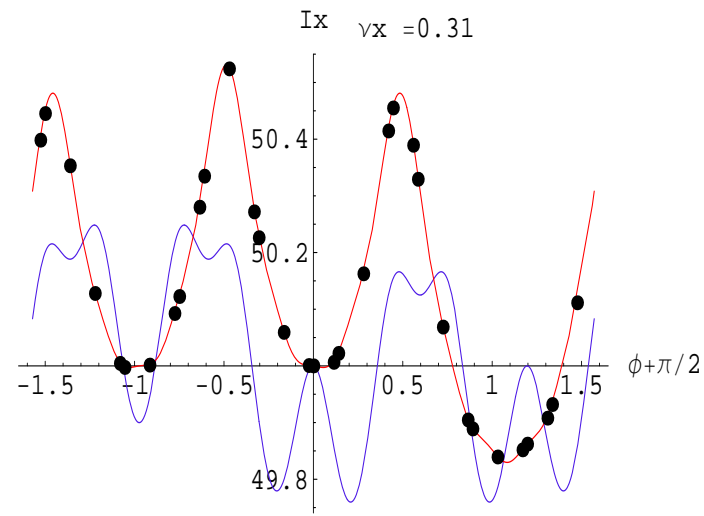
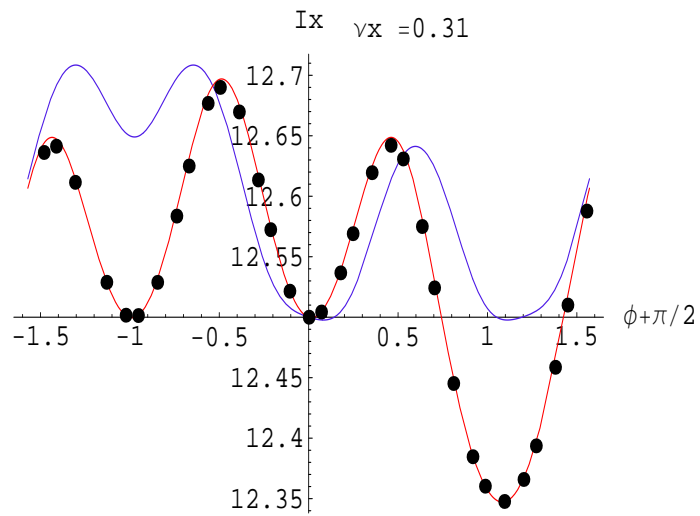
Nota bene, because of:

$$e^{iF(\Psi)} e^{if_2} \rightarrow e^{iF^1(\Psi_1) + F(\Psi)} e^{if_2}$$

can be generalized to more interaction points ...



Invariant versus tracking: two IPs



→ Shown for $5\sigma_x$ and $10\sigma_x$



Recap: Hamiltonian for a finite length element

We have from the Hamiltonian equations for the motion through an element with the Hamiltonian H for the element of length L :

$$\frac{dq}{dt} = [q, H] =: -H : q \quad (\text{from lecture 5})$$

$$\rightarrow \frac{d^k q}{dt^k} = (: -H :)^k q$$

$$\rightarrow q(t) = \sum_{k=0}^{\infty} \frac{t^k}{k!} \left(\frac{d^k q}{dt^k} \right) = \sum_{k=0}^{\infty} \frac{t^k}{k!} (- : H :)^k = e^{-tH} :$$

with independent variable s instead of t (nota bene: $s_0 = 0, t_0 = 0$):

$$\rightarrow q(s) = e^{-LH} :$$



Lie transformations on moments:

We have used Lie transformations mainly to propagate coordinates and momenta, i.e. like:

$$e^{:f:} x_0 = x_1$$

$$e^{:f:} p_0 = p_1$$

or using $Z = (x, p_x, y, p_y, \dots)$:

$$e^{:f:} Z_0 = Z_1$$

- Remember: can be applied to any function of x and p !!
- In particular to moments like x^2, xp, p^2, \dots



Lie transformations on moments

Assume a matrix M of the type:

$$M = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix}$$

described by a generator f , we have for the Lie transformation on the moment:

$$e^{:f:} x^2 = (e^{:f:} x)^2 \quad (\text{see lecture 5})$$

therefore:

$$(e^{:f:} x)^2 = (m_{11}x + m_{12}p)^2$$
$$(e^{:f:} x)^2 = m_{11}^2 x^2 + 2 m_{11} m_{12} x p + m_{12}^2 p^2$$



More on moments

To summarize the moments:

$$\begin{pmatrix} x^2 \\ xp \\ p^2 \end{pmatrix}_{s_2} = \begin{pmatrix} m_{11}^2 & 2m_{11}m_{12} & m_{12}^2 \\ m_{11}m_{21} & m_{11}m_{22} + m_{12}m_{21} & m_{12}m_{22} \\ m_{21}^2 & 2m_{21}m_{22} & m_{22}^2 \end{pmatrix} \circ \begin{pmatrix} x^2 \\ xp \\ p^2 \end{pmatrix}_{s_1}$$

This is the well known transfer matrix for optical parameters



A real life example: beam-beam interaction^{*)}

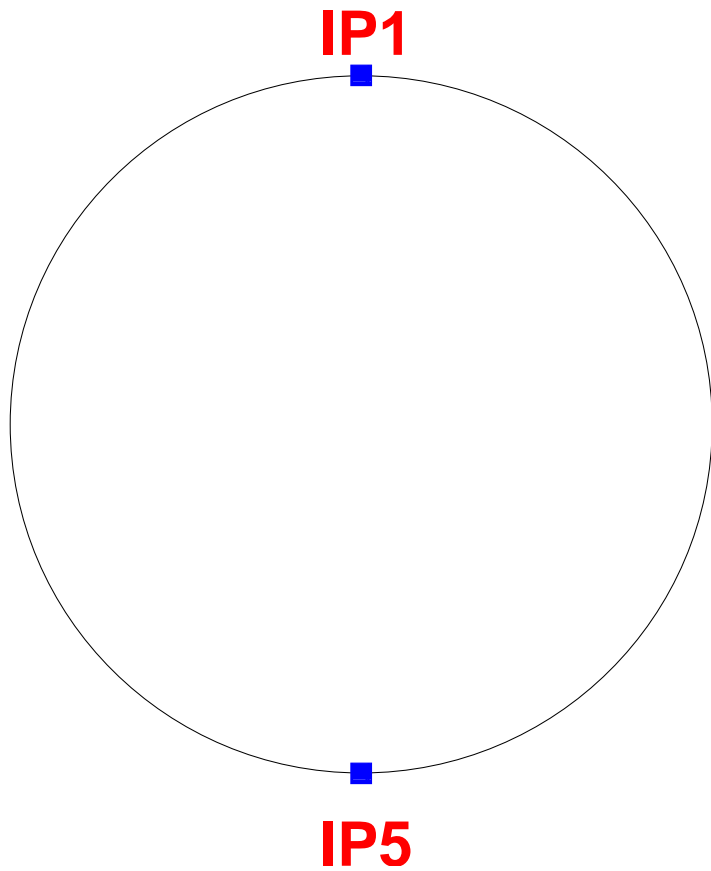
- Beam-beam interaction very non-linear
- Important to understand stability
- Non-linear effects such as amplitude detuning very important

Our questions ?

- How does the particles behave in phase space ?
- Do we have an invariant ?
- Can we calculate the invariant ?

^{*)} From: W. Herr, D. Kaltchev, LHC Project Report 1082, (2008).

Collision scheme - two IPs



Start with single IP

”Classic” (B.C.) approach:

- Interaction point at beginning (end) of the ring (very local interactions, δ -functions)
- s-dependent Hamiltonian and perturbation theory:

$$\mathcal{H} = \dots + \delta(s)\epsilon V$$

■ Disadvantages:

- for several IPs endless mathematics
- conceptually and computationally easier method



Effect on invariants - start with single IP

Look for invariants h , (see e.g. Dragt¹⁾), and evaluate for different number of interactions and phase advance.

Very well suited for local distortions (e.g. beam-beam kick)

Linear transfer $e^{i f_2}$ and beam-beam interaction $e^{i F}$, i.e.:



$$e^{i f_2} \cdot e^{i F} = e^{i h}$$

with

$$f_2 = -\frac{\mu}{2} \left(\frac{x^2}{\beta} + \beta p_x^2 \right)$$

and

$$F = \int_0^x dx' f(x')$$

 ¹⁾ A. Dragt, AIP Conference proceedings, Number 57 (1979) 

Effect on invariants

using for a Gaussian beam $f(x)$:

$$f(x) = \frac{2}{x} \left(1 - e^{-\frac{x^2}{2\sigma^2}}\right)$$

as usual go to action angle variables Ψ , A :

$$x = \sqrt{2A\beta} \sin\Psi, \quad p = \sqrt{\frac{2A}{\beta}} \cos\Psi$$

and write $F(x)$ as Fourier series:

$$F(x) = \sum_{n=-\infty}^{\infty} c_n(A) e^{in\Psi} \quad \text{with :} \quad c_n(A) = \frac{1}{2\pi} \int_0^{2\pi} e^{-in\Psi} F(x) d\Psi$$

We need:

REMEMBER: with this transform:

$$f_2 = -\mu A$$

and useful properties of Lie operators (any textbook²⁾):

$$:f_2: g(A) = 0, \quad :f_2: e^{in\Psi} = in\mu e^{in\Psi}, \quad g(:f_2:) e^{in\Psi} = g(in\mu) e^{in\Psi}$$

and the formula (because the beam-beam perturbation is small !):

$$e{:f_2:} e{:F:} = e{:h:} = \exp \left[:f_2 + \left(\frac{:f_2:}{1 - e^{-:f_2:}} \right) F + \mathcal{O}(F^2) : \right]$$

²⁾ E. Forest, "Beam Dynamics, A New Attitude and Framework", 1998

Single IP

gives immediately for h :

$$h = -\mu A + \sum_n c_n(A) \frac{in\mu}{1 - e^{-in\mu}} e^{in\Psi}$$

$$h = -\mu A + \sum_n c_n(A) \frac{n\mu}{2\sin(\frac{n\mu}{2})} e^{(in\Psi + i\frac{n\mu}{2})}$$

away from resonance, a normal form transformation takes away the pure oscillatory part and we have only:

$$h = -\mu A + c_0(A) = \text{const.}$$

$$\left[\text{homework : } \frac{dc_0(A)}{dA} \right]$$


Single IP

If you are too lazy or too busy:

$$\Delta Q = \frac{-1}{2\pi} \frac{dc_0(A)}{dA}$$

is the detuning with amplitude, i.e. the amplitude dependent frequency change of the transformation we had before ...

We get:

$$\Delta Q = \frac{-1}{2\pi} \frac{Nr_0}{\gamma A} [1 - e^{-A\beta/2\sigma^2} I_0(A\beta/2\sigma^2)]$$



Single IP - analysis of h

$$h = -\mu A + \sum_n c_n(A) \frac{n\mu}{2\sin(\frac{n\mu}{2})} e^{(in\Psi + i\frac{n\mu}{2})}$$

On resonance:

$$Q = \frac{p}{n} = \frac{\mu}{2\pi}$$

with $c_n \neq 0$:

$$\sin\left(\frac{n\pi p}{n}\right) = \sin(p\pi) \equiv 0 \quad \forall \text{ integer } p$$

and h diverges

Invariant versus tracking

■ Is it useful what we obtained ?

→ Debug and compare (“benchmark”)

■ Compare to very simple tracking program:

→ linear transfer between interactions

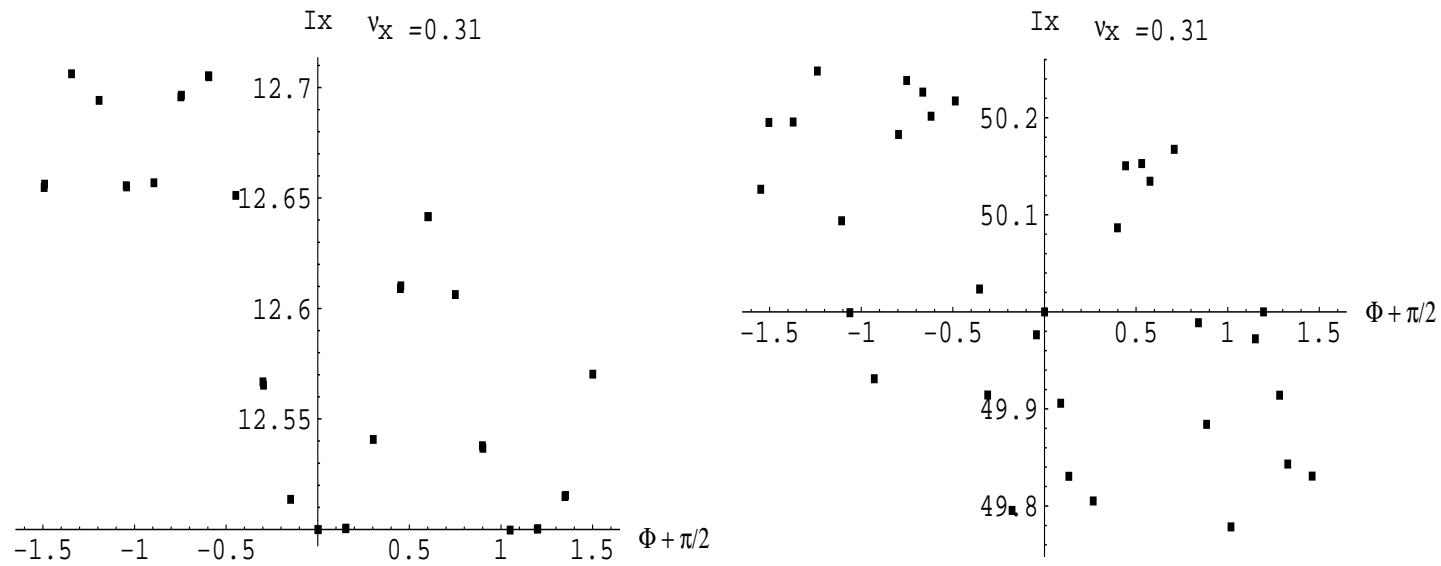
→ beam-beam kick for round beam

→ compute action $I = \frac{\beta^*}{2\sigma^2} \left(\frac{x^2}{\beta^*} + p_x^2 \beta^* \right)$

→ and phase $\Psi = \arctan\left(\frac{p_x}{x}\right)$

→ compare I with h

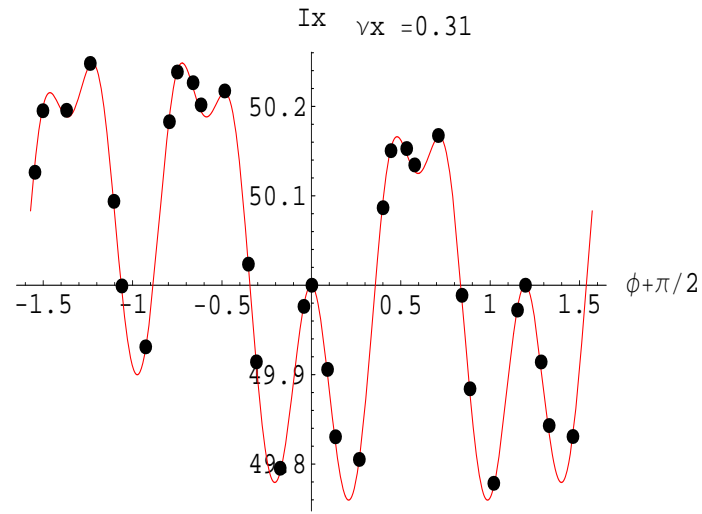
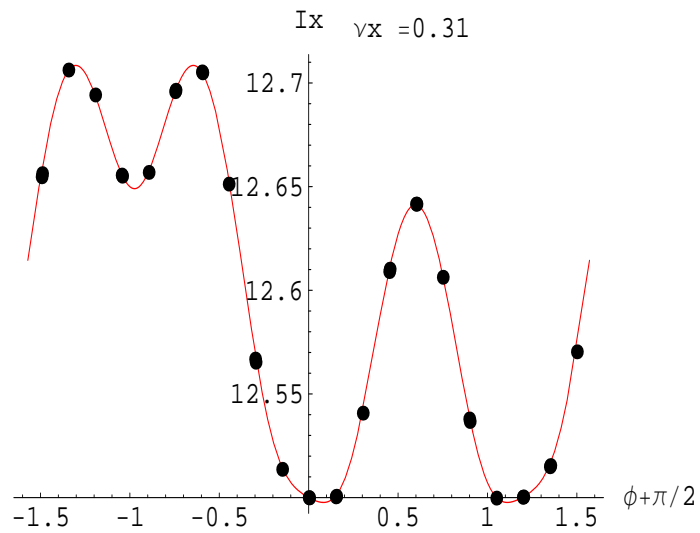
Invariant from tracking: one IP



→ Shown for $5\sigma_x$ and $10\sigma_x$



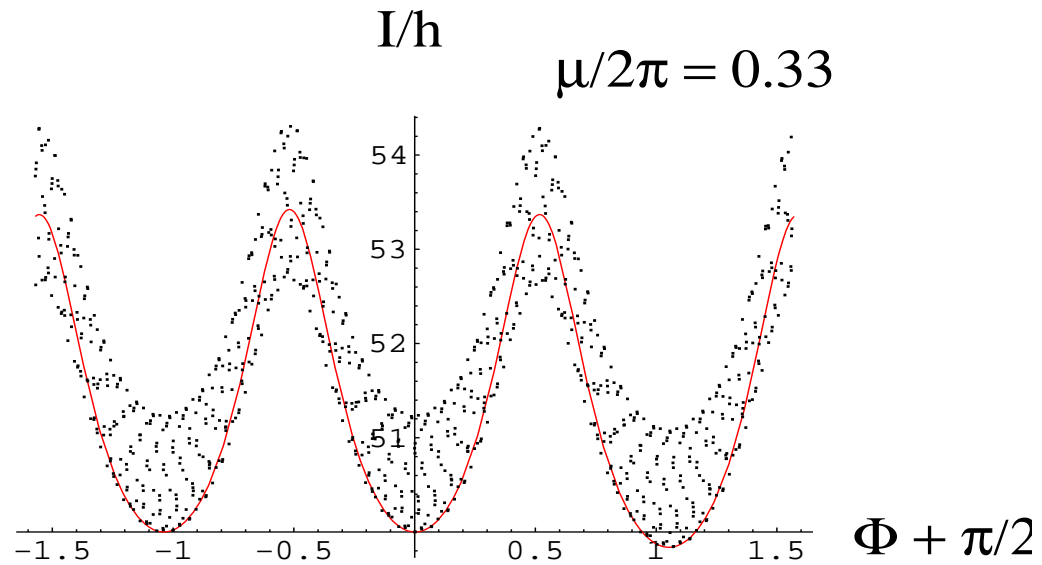
Invariant versus tracking: one IP



→ Shown for $5\sigma_x$ and $10\sigma_x$



Invariant versus tracking:



- ➔ Behaviour near a resonances: no more invariant possible
- ➔ Envelope of tracking well described

What about close to resonance ?

If we have $Q = \frac{\mu}{2\pi} \approx \frac{m}{3}$ (3rd order resonance). Using a "distance to resonance d " as:


$$Q = \frac{m + d}{3} \quad \text{where : } d \ll 1$$

The trick is to observe the motion every **3** turns:

$$\mathcal{M}^3 = (e^{-\mu J} e^{ikx^3})^3 = e^{3h}$$

We get a factor:

$$e^{-3\mu J} = e^{-2\pi d J} \quad (\text{because : } e^{-2\pi m J} \equiv 1)$$

$$d = \frac{3\mu}{2\pi}$$


What about close to resonance ?

Without proof (but like before, see e.g. Chao), we get:

$$h = -\frac{2\pi}{3}dJ - \frac{\pi}{12}dk(2J)^{3/2} \cdot \left(\frac{\sin(3\Psi + \frac{3\mu}{2})}{\sin\frac{3\mu}{2}} - \frac{\sin(\Psi + \frac{\mu}{2})}{\sin\frac{\mu}{2}} \right)$$

For small d ($\sin\frac{3\mu}{2} \approx -\pi d$) we can simplify:

$$h \approx -\frac{2\pi}{3}dJ - \frac{1}{\sqrt{2}}k(\beta J)^{3/2}\sin(3\Psi)$$

Analysis give fixed points, i.e. (back in Cartesian again):

$$\frac{\partial h}{\partial x} = -\frac{2\pi}{3}dx - \frac{1}{4}\beta^{3/2}(3x'^2 - 3x^2) = 0$$

$$\frac{\partial h}{\partial x'} = -\frac{2\pi}{3}dx' - \frac{1}{4}\beta^{3/2}3xx' = 0$$

