## **pyTRAIN** a modern TRAIN implementation

### Michi Hostettler, Xavier Buffat, Tobias Persson, Tatiana Pieloni, Jorg Wenninger



### a brief history TRAIN

#### iteratively find self-consistent closed orbits

- under the presence of beam-beam effects Ο
- in the many-bunch case ("trains") Ο
- using second-order sectormaps from MAD-X  $\bigcirc$
- "soft-Gaussian" approach Ο

### pioneered by E. Keil, F.C. Iselin for LEP

### applied to LHC design by W. Herr, H. Grote

- showed clear advantage of V-H alternating Ο crossing scheme over H-H crossing
- showed "PACMAN" effects (bunches missing Ο long-range encounters due to kicker gaps)
- later further extended by T. Pieloni, Ο
  - A. Gorzawski, M. Hostettler, X. Buffat, A. Ribes





W. Herr, "Features and implications of different LHC crossing schemes", LHC project report 628, 2003.

### **TRAIN** in a nutshell



### historic TRAIN limitations

#### • ltrain.f, one flat file, 12145 lines of FORTAN-77

- local implementation of all numeric primitives (linear algebra, ...)
  - partially copied from historic MAD-8 or MAD-X code
  - not always numerically stable or the most efficient
- no version control, no changelog (ktrain, ltrain, mtrain, ...)

### historic input formats

- flat files of records/numbers
   read(..., \*) in
- MAD-X maps: historic scripts

# 1=full\_coll 2=nturn 3=debug 4=# of out\_bunches 5=out\_pos 6=write\_norm
# 7=xi\_fact 8=hofact 9=amp\_bunch (0=all,- every..) 10= amp\_fac (see below)
# 11=lumi\_hist 12=beam\_2 offset (half-buckets)
# 1 2 3 4 5 6 7 8 9 10 11 12
1 1 1 0 16 0 1 1 0 0 0 0
# list of out\_bunches
10 16 22 0 0 0 0 0 0 0 0 0 0

### limited extensibility & scriptability

- today most analysis is done in python / Jupyter
- first version of pyTRAIN: TRAIN python interface
  - limited to running TRAIN in full, no control over the process, output only for BB points



# the FORTRAN TRAIN code served us well for many years

### now it is time to move on





### pyTRAIN: a modern re-implementation

- complete re-implementation in python
- using numpy and scipy primitives
  - linear algebra, Faddeeva function, ...

#### • interface to MAD-X via cpymad

- reading MAD-X output from files also possible
- total 1050 lines of python code
  - 44 lines of Cython: concatenation of second-order maps

### • performance similar to FORTAN-77 code

- slightly slower not the first priority
- few minutes to solve full LHC with ~2400 bunches





https://gitlab.cern.ch/mihostet/pytrain/-/tree/pure-python



### pyTRAIN - basic usage

from pytrain.fileio import read\_train\_files
from pytrain.machine import FillingScheme
from pytrain.solver import solve train

# read survey, twiss & sector map input files (alternatively use included cpymad utils)
machine, twiss\_b1, twiss\_b2, maps\_b1, maps\_b2 = read\_train\_files('train-output')

# construct a "filling scheme": bunch intensities & normalized emittances
filling\_scheme = FillingScheme(int\_b1, int\_b2, emit\_b1x, emit\_b1y, emit\_b2x, emit\_b2y)

# solve self-consistent orbits with BBLR interactions
result = solve\_train(machine, filling\_scheme, twiss\_b1, maps\_b1, twiss\_b2, maps\_b2)

# bunch-by-bunch closed orbit at any element co\_b1\_x, co\_b1\_y = result.bunch\_positions\_b1('MKIP5') co\_b2\_x, co\_b2\_y = result.bunch\_positions\_b2('MKIP5')



### benchmarking: TRAIN vs pyTRAIN









### beam-beam long-range effects in LHC

#### • ~2400 bunches spaced by 25ns

longer gaps for kicker rise times

# • 4 interaction regions with common vacuum chamber

- long-range beam-beam encounters
- "pacman" effects due to kicker gaps (missing LR encounters)
- "super-pacman" effects as IRs not symmetric (missing head-on colls.)
- luminosity levelling by beta\* and separation: changing optics



#### more details in talk of T. Pieloni

![](_page_8_Picture_10.jpeg)

### luminous centroid position at experiments

#### experiments measure the primary vertex positions ("beam spot")

- offline reconstructed from tracker data
- "luminous region" size
- "luminous centroid" position
- high-statistics data collected during calibration sessions
  - interferes with physics data taking
- "luminous centroid": center of the overlap region
  - average position of the two beams
  - measured bunch-by-bunch

![](_page_9_Figure_10.jpeg)

![](_page_9_Picture_11.jpeg)

### **ATLAS luminous centroid position**

![](_page_10_Figure_1.jpeg)

Preliminary BeamSpot data courtesy of the ATLAS collaboration. LHC fill 9653 / ATLAS run 476033, 2024-05-28

### emittance scans - beam separations

#### • beam separation scans ("mini-VdM")

- Iuminosity vs. separation fitted with Gaussian bunch-by-bunch
- done regularly in LHC for emittances and tracking of luminosity monitor degradation

#### • fit centre gives bunch-by-bunch beam separation

![](_page_11_Figure_5.jpeg)

Luminosity [Hz/µb]

ſ

50

100

Time [s]

150

details in: M. Hostettler et al., "Luminosity scans for beam diagnostics", PRAB 21, 2018

### emittance scans - beam separations

![](_page_12_Figure_1.jpeg)

emittance scans in CMS, at beta\*=1.2m. Luminosity data courtesy of the CMS collaboration. LHC fill 10066, 2024-08-28

CÉRN

### beam positions at wire scanners

#### • wire scans regularly taken during LHC injection

- first 108 bunches only (scanner intensity limit)
- bunch positions from centre of Gaussian fit
- o per-beam, per-plane, per-bunch data

#### • no head-on collisions, but long-range encounters present!

![](_page_13_Figure_6.jpeg)

![](_page_13_Figure_7.jpeg)

### beam positions at wire scanners

![](_page_14_Figure_1.jpeg)

CERN

### from orbits to optics parameters

- classic TRAIN output: orbits, tunes, chromaticities per bunch
- internal calculation based on second-order maps (per bunch)
  - the maps contain all optics information up to second order!
  - → calculate twiss parameters & dispersion for any bunch, at any location
    - based on MAD-X code translated to python
    - coupling not yet treated future improvement

![](_page_15_Figure_7.jpeg)

![](_page_15_Figure_8.jpeg)

![](_page_15_Picture_9.jpeg)

### iteration processors

![](_page_16_Figure_1.jpeg)

- e.g. feedback corrections
- self-consistent treatment needs to hook into TRAIN iteration loop

#### iteration processors

- callback per iteration after new beam-beam maps are established
- can insert / mutate bunch maps

#### • pre-defined: mean orbit correction (SVD)

 simulates the effect of a bunch-average feedback

![](_page_16_Figure_9.jpeg)

![](_page_16_Picture_10.jpeg)

### conclusions

### • pyTRAIN: a reimplementation of TRAIN in modern python

- using numpy / scipy primitives where possible
- interface to MAD-X via cpymad
- scriptable from python

#### • results look promising

• reproduces well the BBLR patterns observed in LHC

#### allows for novel features

- orbit anywhere in the ring (not just BB interaction points)
- twiss parameters anywhere in the ring
- iteration processors
- extensible in the future!

![](_page_17_Picture_12.jpeg)

### future improvements

### • physics improvements

- finite bunch length effect on the effective beam size due to beam angles
  - A. Babaev, <u>https://arxiv.org/abs/2104.02595</u>
- fully self-consistent beam sizes, iterating on perturbed twiss parameters
- treatment of beam-beam introduced coupling

#### • validation & application to other machines

currently only tested on LHC - any collaborators welcome!

#### • integration with Xsuite

- at least for input map generation
- install beam-beam elements at final separation for tracking?

#### • performance

- nice to have: clean and readable code has higher priority
- possible synergy with Xsuite integration (fast primitives and data structures)