

1

# **Multiobjective Optimization of Injectors**

# **Christos F. Papadopoulos, Lawrence Berkeley National Laboratory**



- I. V. Bazarov, C. K. Sinclair, Phys. Rev. ST Accel. Beams 8, 034202 (2005): Excellent resource on injector dynamics & multiobjective optimization
- K. Deb, Multi-Objective Optimization Using Evolutionary Algorithms, (2002): Thorough discussion of different optimization strategies



## Designing an injector that takes advantage of laser shaping, velocity and ballistic bunching, emittance compensation etc is challenging



Used for emittance compensation

Multicell Accel. Cavities Control Phase, Amplitude Used for compression, acceleration

Laser Pulse: Control transverse and longitudinal profile Defines initial bunch length and emittance Problem 2: Components quickly add up. In order to take advantage of the flexibility of the system, we need to optimize > 10 parameters.

Problem 3: Analytic models offer good qualitative understanding, but not quantitative design parameters, since they do not include nonlinear SC, detailed fields etc



## **Definition of our 1<sup>st</sup> problem: Maximize** f(x, x)

$$f_m(x_{1,}x_{2,...,x_n}), m=1,...,M;$$

Subject to the constraints

$$g_{j}(x_{1,}x_{2,.}.,x_{n}) \ge 0, j=1,...,J;$$
  
 $x_{i}^{(L)} \le x_{i} \le x_{i}^{(U)}, i=1,...,N;$ 

In practical terms:

The objective functions f: The most important quantities of the beam, such as emittance, bunch length, energy spread. Ultimately, in start to end simulations, the goal is to optimize the quantities users want.

The constraints xu, xl: upper and lower values allowed in our "knobs", such as peak gradients of RF fields, minimum and maximum laser pulse length, etc The constraints g: Constraints on calculated quantities, such as transverse beam size, number of particles lost, etc

The difference between the f's and the g's can be subtle, and deciding which is which depends on whether we have a gradual or a threshold effect.



# **Pareto Optimality**

LA2: Lecture Title (C. F. Papadopoulos)

#### **Definition of dominance:**

Solution A is said to dominate solution C if A is not worse than C in all the objectives f, and is strictly better than C in at least one objective. **Definition of non-dominated subset:** 

The subset P' of solutions within a solution set P that are not dominated by another member of P. **Definition of Pareto-optimal set:** 

The non-dominated subset P' when P is the set of all f1 the allowable search space



Vilfredo Pareto 1848-1923





#### A multi-stage process

- **1. Initialize population**
- 2. Evaluate objective functions / constraints
- 3. Assign fitness to all individuals (convergence & diversity), nondominated solutions are preferred.
- 4.Stochastically choose a subset for mating pool (higher fitness being preferred)
- 5. Apply crossing and mutation operators to generate offspring
  - Crossing: combine solutions to (hopefully) find a better one
  - Mutation: Introduce randomness to investigate larger volumes of parameter space
- 6. Evaluate objectives / constraints for the offspring
- 7. Repeat from step 3.

#### **Addresses problem 2:**

Global and efficient search in the multidimensional parameter space. The random aspect helps avoid local minima.

#### **One caveat:**

We can only approximate the Pareto optimal set, as we are not guaranteed that another solution which dominates all or some of the current ones does not exist.



Particle-in-Cell code, includes trans. and long. space charge, widely used and benchmarked for photoinjectors

**ASTRA** 

- 300 pC charge
- **10k-100k particles**
- Variable step size, Variable grid
- Not enough to resolve microbunching
- No CSR or wakefields
- Good enough for core properties, emittance growth FAST (10 mins-1hr for a single run)

Longitudinal Phase-Space

z = 20.00 m





Slice Emittance

Addresses Problem 3: Nonlinear SC forces, realistic fields



٥

∆t ps

5



 $5 \times 10^{-3}$ 

-5



LA2: Lecture Title (C. F. Papadopoulos)

#### Use a cluster of linux machines

Computes 100s of solutions at once, usually corresponding to 1 generation "Embarrassingly" parallel, since each solution is calculated on 1 processor Few cycles are lost, since each solution takes comparable time, and overall computation scales well with the number of processors



## Addresses Problem 3 Allow for faster computation of multiple solutions

Electron Injectors for 4th Generation Light Sources- University of Texas at Austin, January 23-27, 2012



## **Choose objectives at the injector exit** Start with emittance and bunch length Also interested in 95% emittance to estimate halo

100's of generations, a few days running in a cluster (computer time, not people's time)







Within specs for peak current, slice emittance (95%) But the tails and and the high order momentum-time correlations were problematic during the linac transport. Specifically, after removing 1<sup>st</sup> and 2<sup>nd</sup> order t-pz correlation, the remainder rms e-spread is 12.95 keV





# **Tail management**

11





objective: Add 3<sup>rd</sup> 8 The energy spread after removing <t\*pz> (can the downstream use for this) linac and 3<sup>rd</sup> <t^2\*pz> (use 3 harmonic cavity) The pareto front is no longer a 1D curve





We lose a bit in current and emittance, but reduce the remainder energy spread by a factor of more than 3 New remainder rms e spread = 3.325 keV





16

14

12

10

8

6

4

2

0

0

1

Energy (MeV)

Run 184

Run 026

Reducing the laser tails improves the emittance In turn this allows for shorter bunches at the cathode



beamline

