

Overview and Status of Linac Simulations with SelaV

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semi-Lagrangian Vlasov simulation

- Vlasov codes solve the Vlasov equation, i.e. they find a solution for the phase space density (PSD) $\Psi(t, z): \mathbb{R} \times \mathbb{R}^{2n} \rightarrow \mathbb{R}$

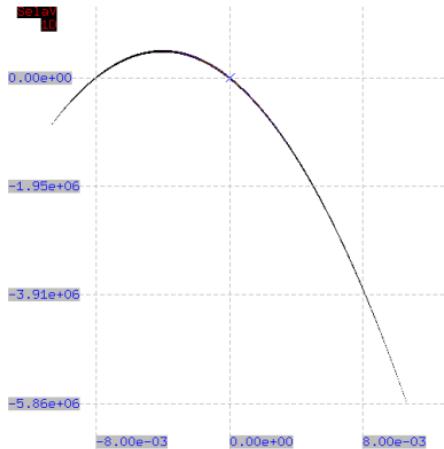
$$\partial_t \Psi - \{H[\Psi], \Psi\} = 0$$

- semi-Lagrangian codes use the method of characteristics to construct the solution using the flow $\phi: \mathbb{R} \times \mathbb{R}^{2n} \rightarrow \mathbb{R}^{2n}$ of the Hamiltonian $H: \mathbb{R}^{2n} \rightarrow \mathbb{R}$ (i.e. $\partial_t \phi = \{H[\Psi], \phi\}$)

$$\Psi(t_1, z) = \Psi(t_0, \phi_{t_0 \leftarrow t_1}(z))$$

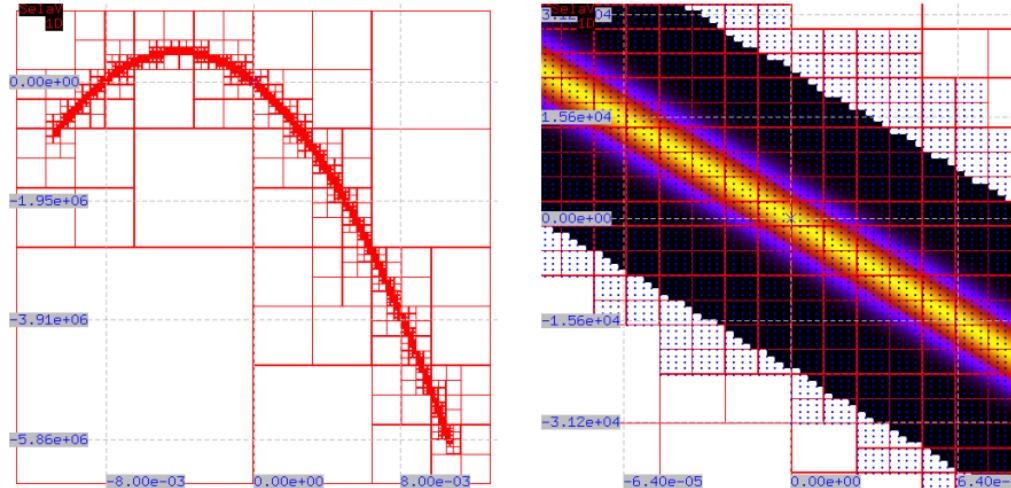
- implementation in a simulation code: store PSD values on a grid, update by back-tracking PS-coordinates and interpolate old PSD (a.k.a. Perron-Frobenius step)

challenge: exotic densities in FELs



- longitudinal PSDs in FELs are exotic:
 - local energy variance is small compared to the energy range
 - curvature due to non-linearities from RF, collective effects
- “ Ψ is exotic” $\iff \text{supp}(\Psi)$ is much smaller than its minimum bounding rectangle (MBR)
- problem for simulation codes: sampling Ψ on a homogeneous grid covering its MBR, includes huge empty areas
 - \implies too wasteful for high resolutions

solution: tree-based domain decomposition



- simulation domain is subdivided into a hierarchy of nested rectangles
- PSD values are only stored on grids (blue dots) covering the smallest rectangles ("leafs")
 \Rightarrow minimizes sampling of $\text{supp } \Psi$
- fundamental idea behind SelaV

the SelaV project

`libselav`

- C++ library
- implements arbitrary dimensional hypercube-tree data structure
 - dimension being a template parameter
- various interpolation routines (up to a C^2 third-order polynomial scheme)
- propagation routines implementing the Perron-Frobenius step
- collection of typical maps/flows
- physics-agnostic (except for some accelerator-physics specific maps/flows)

`selav1d`

- builds upon `libselav`
- implements a 1D vlasov code intended for (but not limited to) FEL linac simulations
- various PSD initialization routines
 - Gaussian, flat-top, analytic function, particle ensemble (e.g. Astra)
- flexible input language allows handling of PSDs, maps as objects

example: Vlasov-Poisson Equation / Landau damping

```

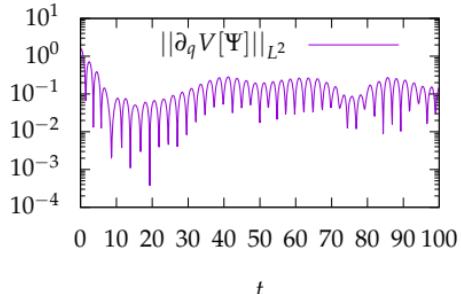
1   psi = psd_analytic(          /* generate psd from analytic func. */
2     "(1-0.5*cos(0.5*q))*exp(-0.5*p*p)/sqrt(2*pi)",
3     limits=[-2*pi, -5, 2*pi,5],      /* simulation domain limits      */
4     topology=1,                      /* C x R                         */
5     depth=4,                        /* depth of quad-tree             */
6     nexp=5,                         /* log2(sample points / dimension) */
7     interpolation=3                /* 5th order C^2 interpolation   */
8 );
9
10 normalize(psi);               /* psi <- psi / int_{R^2} psi dz  */
11
12 T=100;                       /* simulation time                 */
13 DT=1e-1;                      /* time step                       */
14
15 D = map_driftl(DT);          /* define linear drift map         */
16
17 i=0; do(T/DT) {              /* loop over time steps           */
18   print(i);
19
20   /* generate output */
21   show(psi, file=format("%05g.ppm",i)); /* write image of psi            */
22   write_localmoments(psi,           /* write energy-centroid, -variance */
23     format("moments-%05g.dat",i), /* -skewness, etc to file       */
24     1);
25 }
26
27 /* calculate maps */
28 K = map_poissonid(psi,factor=DT,    /* solve 1d poisson eq for psi   */
29   file=format("field-%05g.dat",i), /* also write solution to file   */
30 );
31 M = map_compose(K,D);             /* M(z) = D(K(z))                */
32
33 /* propagate psi */
34 psi = propagate(psi, M,           /* psi <- psi( M^-1(z) )        */
35   box="KEEP", center="KEEP");    /* keep simulation domain         */
36
37 i=i+1;                          /* increment loop counter         */
38
39 };

```

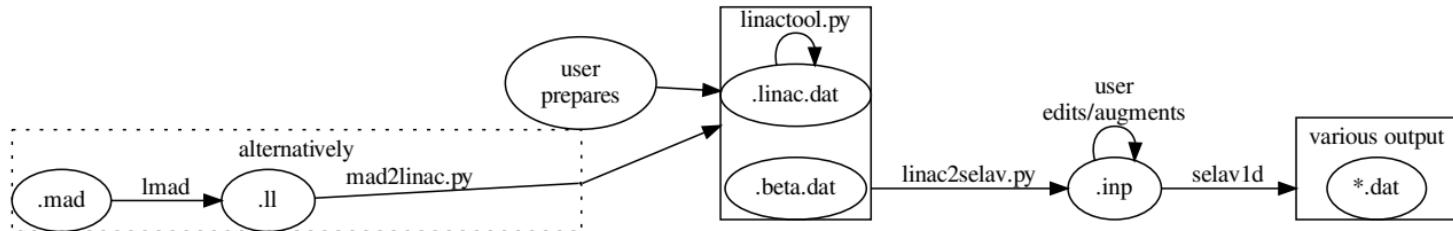
$$\partial_t \Psi - \{ p^2/2 + V[\Psi], \Psi \} = 0$$

$$-\partial_q^2 V[\Psi] = \int_{\mathbb{R}} \Psi dp - 1$$

movie landau



workflow for preparing a linac simulation



- basic idea: generate SelaV input file from simple linac description `.linac.dat` (plus beta function data `.beta.dat`)
- `.linac.dat` are easy to write manually or generate from existing data
 - syntax on next slide
 - optionally w/ some clean-up (remove/rename/merge elements) using `linactool.py`
- crucially, all lattice parameters (except basic geometry) are variables in the SelaV input file
 ⇒ available for scans
- resulting SelaV input file can then be edited/augmented as needed

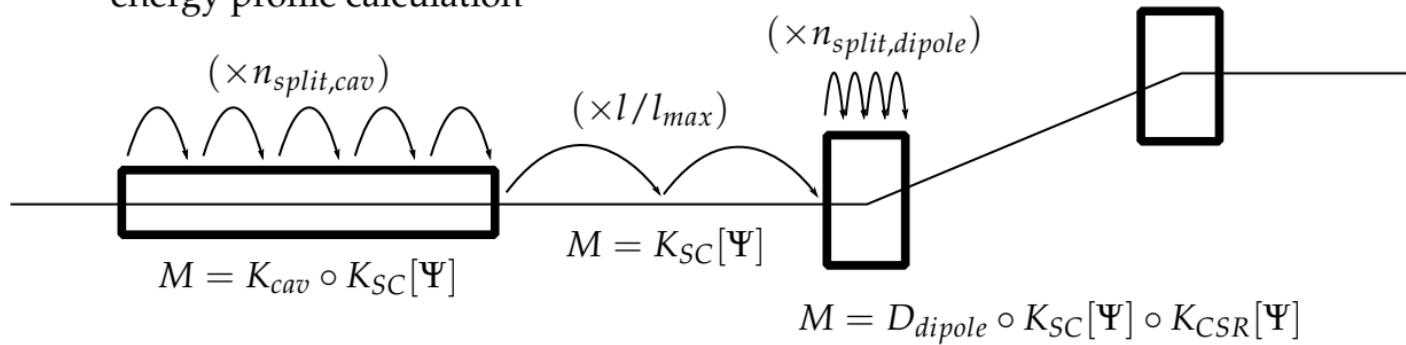
linac2selav

- linac description syntax for linac2selav.py:

```
cavity <name> <z0> <z1> <energy gain [eV]> <frequency [Hz]> <phase [rad]>
| dipole <name> <z0> <z1> <strength [1/m] or rel. to 1st w/ same name>
| quadrupole <name> <z0> <z1> <strength [1/m^2]>
```

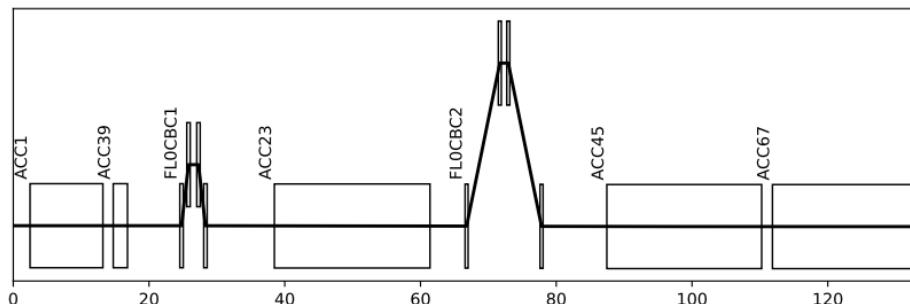
- resulting SelaV input file takes care of:

- dispersion calculations
- spacecharge kicks in non-dispersive regions
- operator splitting (CSR and spacecharge) in dispersive regions
- energy profile calculation



example linac description: FLASH2020+ WP1

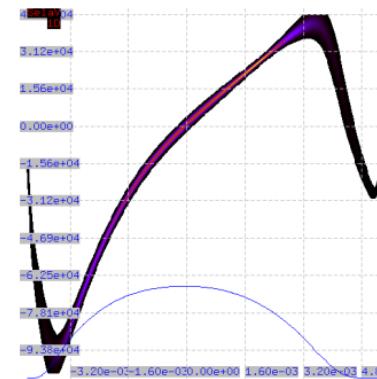
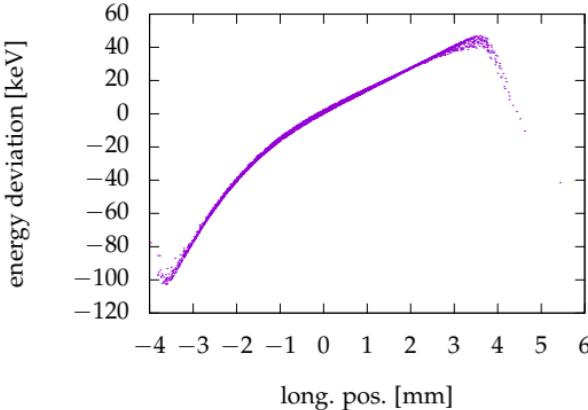
F2020P-WP1.linac.dat						
cavity	ACC1	2.4779	13.2078	159.624e6	1.3e9	0.1102
cavity	ACC39	14.7330	16.8310	-19.224e6	3.9e9	2.9431
dipole	FLOCBC1	24.5403	25.0468	0.5513		
dipole	FLOCBC1	25.5670	26.0736	-1		
dipole	FLOCBC1	27.0366	27.5431	-1		
dipole	FLOCBC1	28.0633	28.5698	+1		
cavity	ACC23	38.4707	61.3960	404.000e6	1.3e9	0.2276
dipole	FLOCBC2	66.5668	67.0137	0.1953		
dipole	FLOCBC2	71.4611	71.9080	-1		
dipole	FLOCBC2	72.7117	73.1586	-1		
dipole	FLOCBC2	77.6060	78.0529	+1		
cavity	ACC45	87.4489	110.2589	380.000e6	1.3e9	0.0
cavity	ACC67	111.8489	134.6494	420.000e6	1.3e9	0.0



import/export particle ensembles

- densities can be imported from ensembles using the “local moments” method
 - divide particles in longitudinal slices around q_i
 - determine energy-centroids μ_i and -variances σ_i^2 of particles in slices
 - use interpolants μ and σ to calculate

$$\Psi(q, p) = \frac{1}{\sqrt{2\pi\sigma(q)^2}} \exp\left(\frac{1}{2}\left(\frac{p - \mu(q)}{\sigma(q)}\right)^2\right)$$



- (N.B.: smooth kernel density estimation was not able to get rid of shot noise)
- export: efficient sampling via weighted tree descent

FLASH2020+ examples (WP1 w/ $\sigma_E = 10\text{keV}$)

homogeneous

seeded

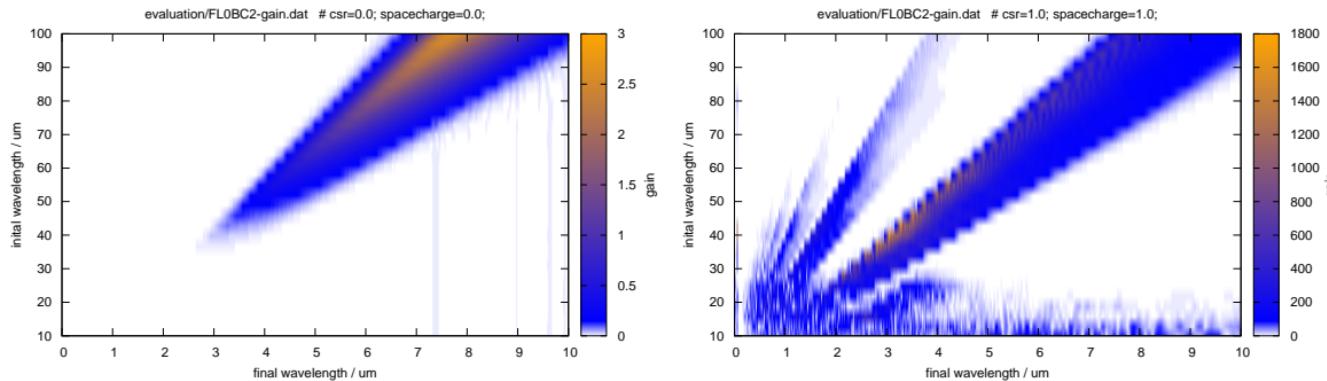
shot-noise

movie smooth

movie seed

movie noise

possible studies: 2D gain functions (old data)



- imprint inhomogeneity on the initially homogeneous PSD

$$\Psi_{A,k_i}^{(\text{ini})}(q, p) = (1 + A \cos(k_i q)) \Psi_0(q, p)$$

- run simulations for different k_i and determine 2D gain function

$$G_A(k_f, k_i) := \frac{\tilde{\rho}_{A,k_i}^{(\text{fin})}(k_f)}{A}, \quad \text{with } \tilde{\rho} = \mathcal{F} \int_{\mathbb{R}} \Psi(\cdot, p) dp$$

- linear gain function: $G(k) = \lim_{A \rightarrow 0} G_A(C k, k)$

Thanks!