Sloppy and Genetic Algorithms for Low-Emittance Tuning at CESR

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CESR



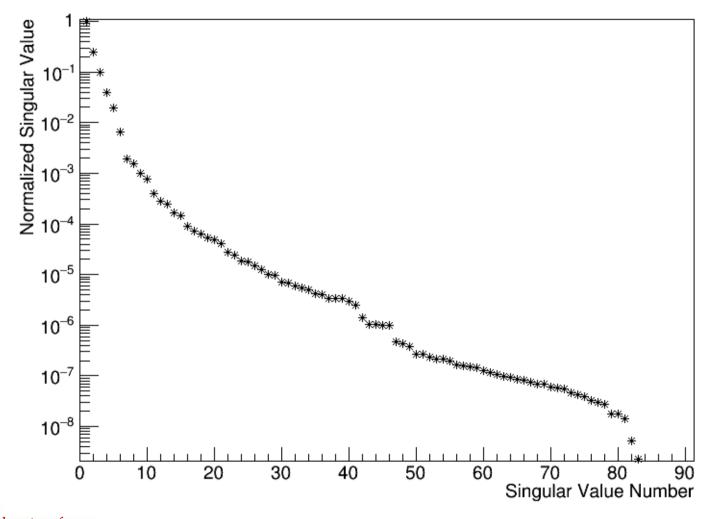


Sloppy Models Problem Statement

- Minimize one objective (vertical emittance/beam size)
- Large number of decision variables (independent magnets)
- No reliable auxiliary information (dispersion, coupling)
- Reliable model of machine responses (BMAD simulation)



Sloppy Models

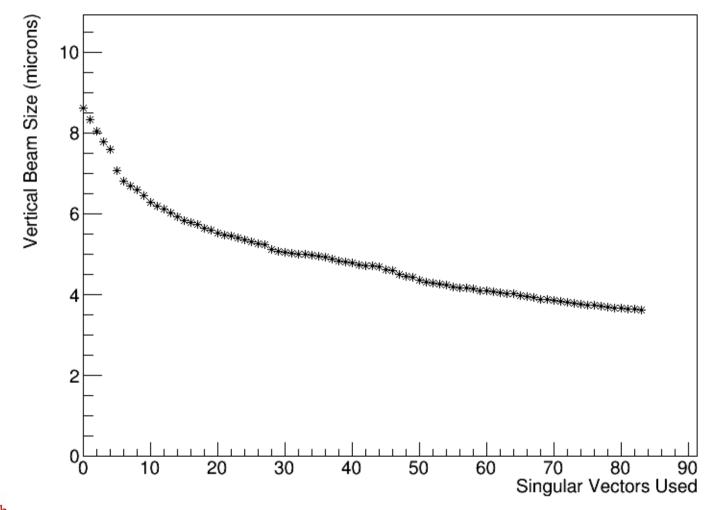




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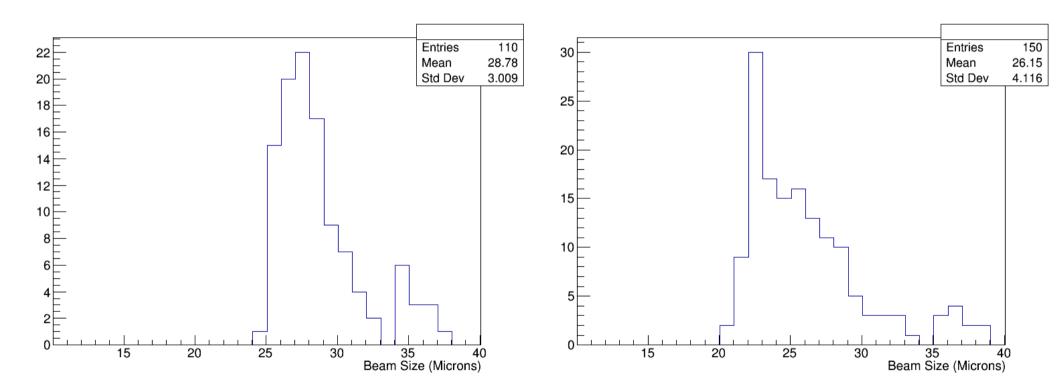
Phys. Rev. E 68 (2003) 021904.

Simulated Results





Experimental Results



RCDS - Nucl. Instr. Meth. 726 (2013) 77.



Conclusions

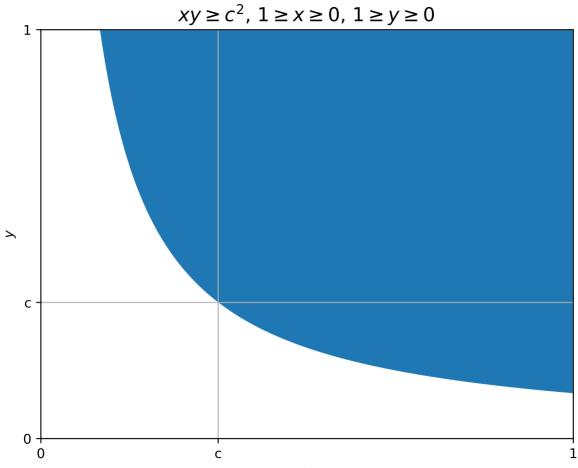
- Knobs provide some improvements
- Still far from quantum limit
- Something missing from models?

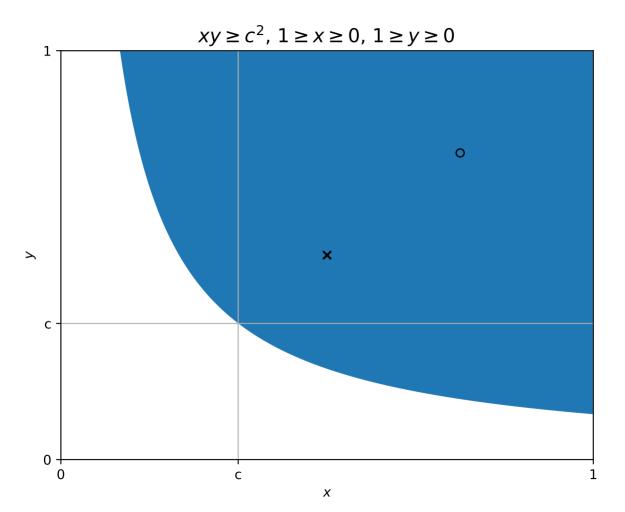


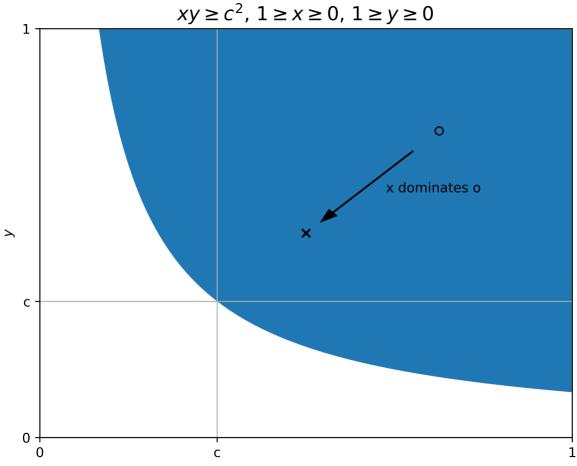
Multi-Objective Genetic Algorithm Problem Statement

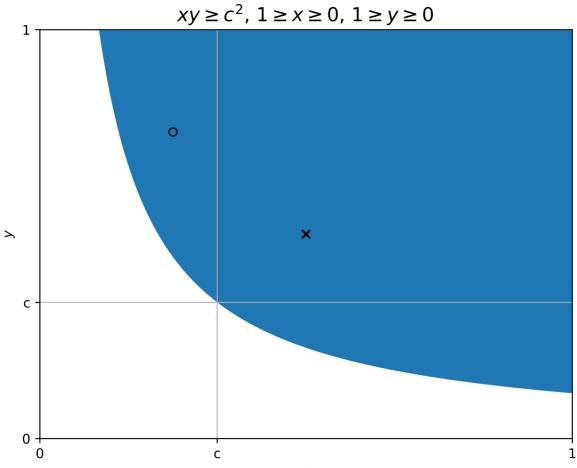
- What if:
 - Competing criteria of optimal machine performance
 - In regime where model of machine responses is unreliable
- Needed: a model-agnostic search for optimal performance *trade-offs*

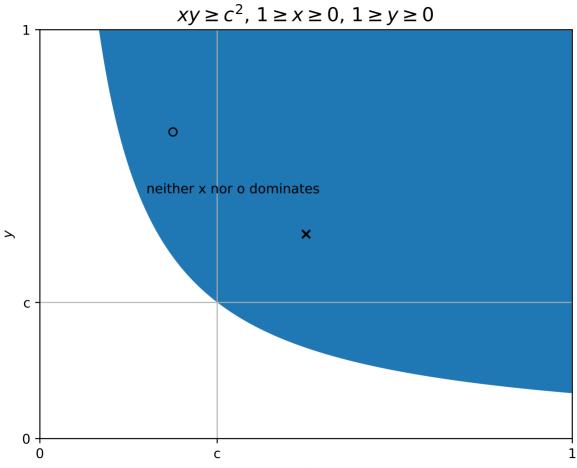


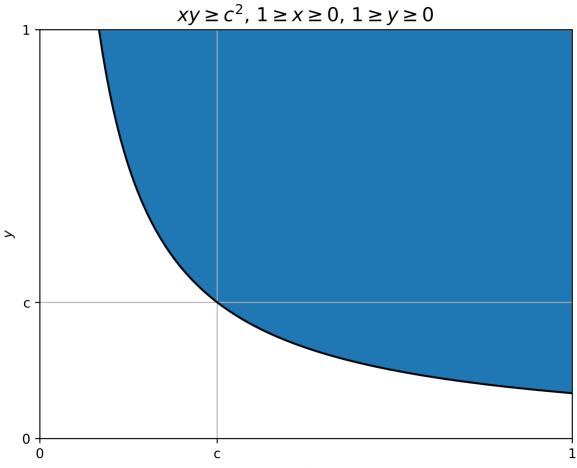


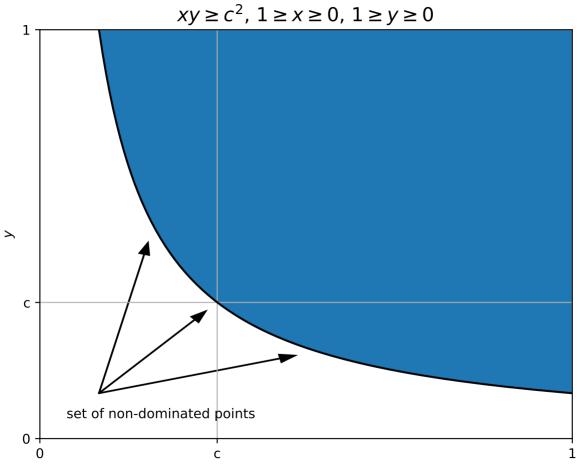


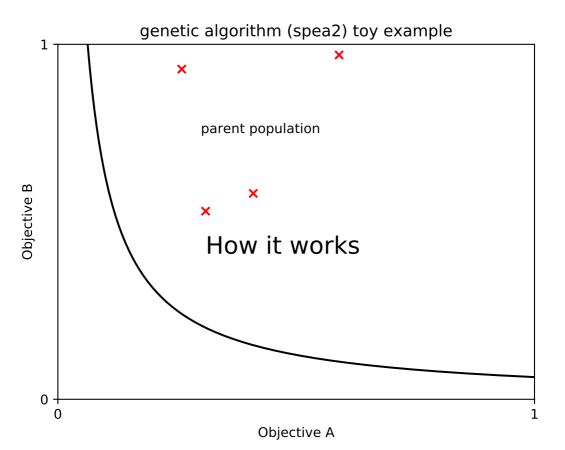


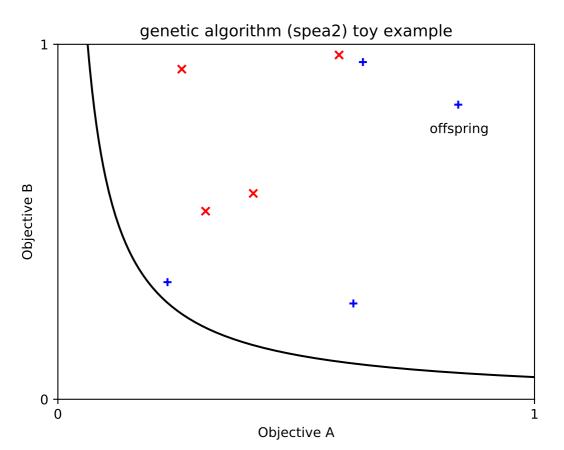


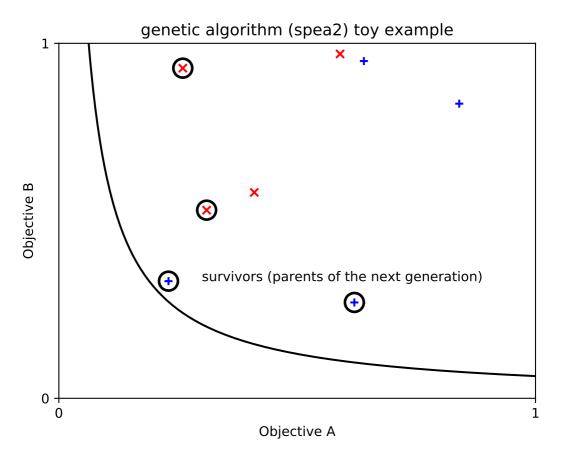


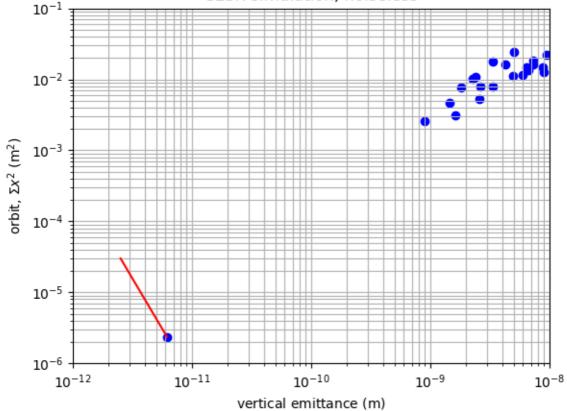


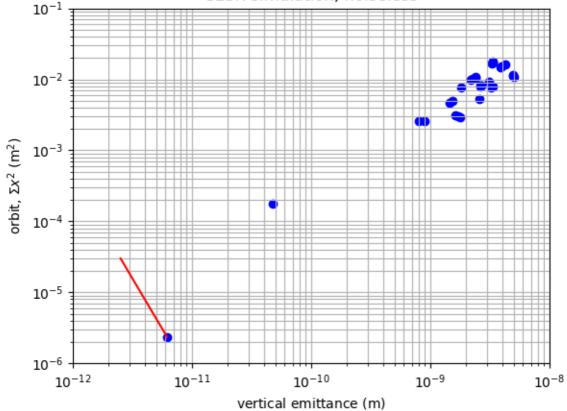


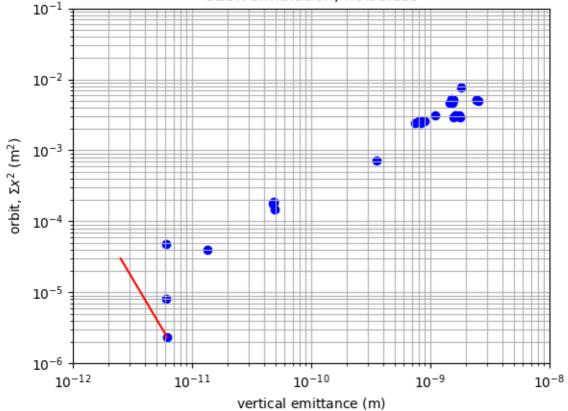


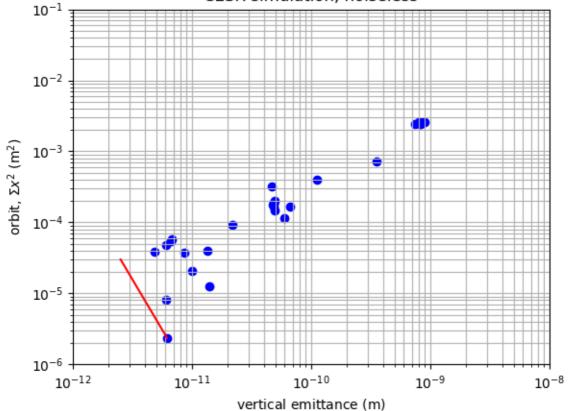


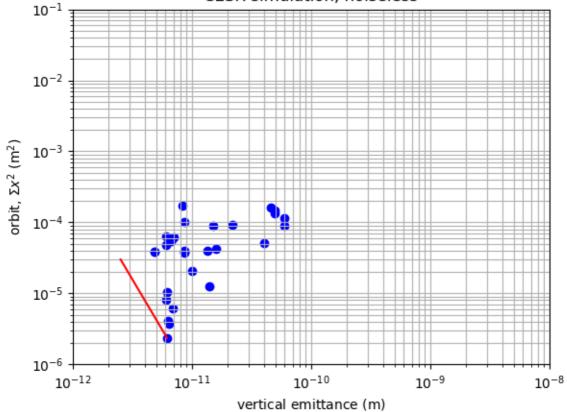


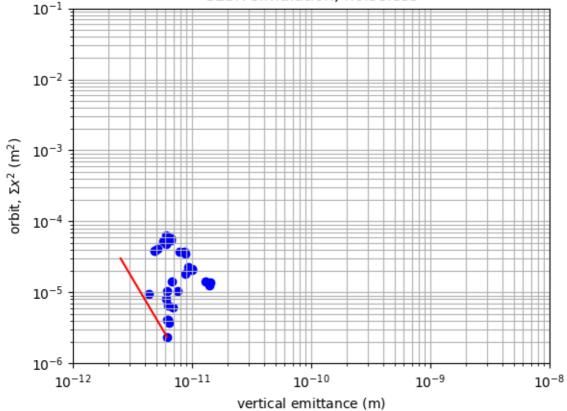


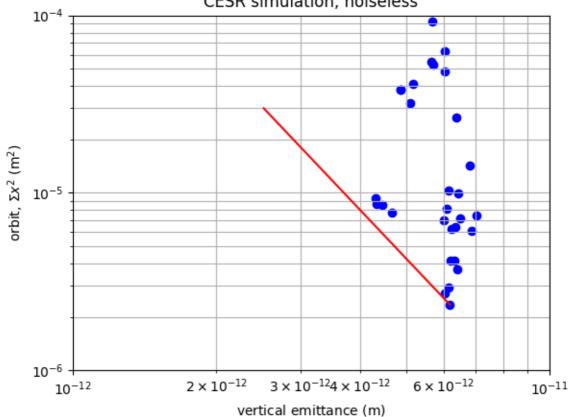


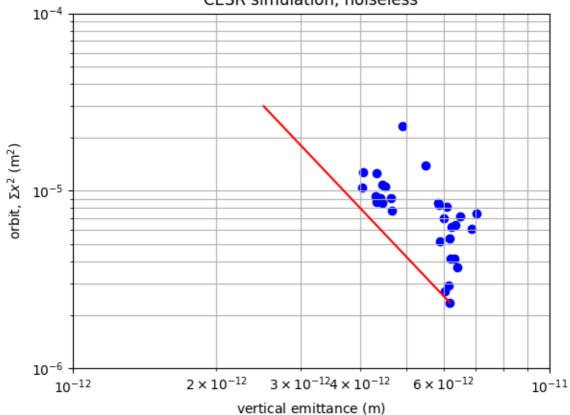


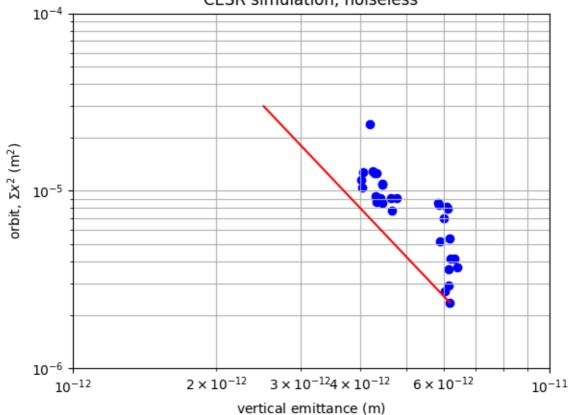


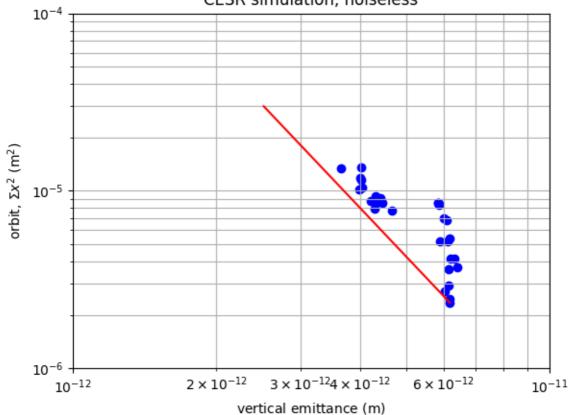


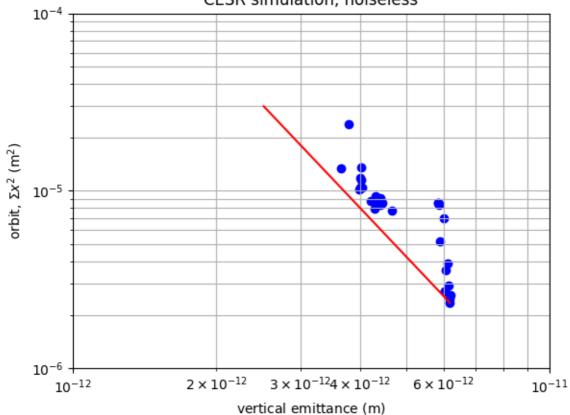


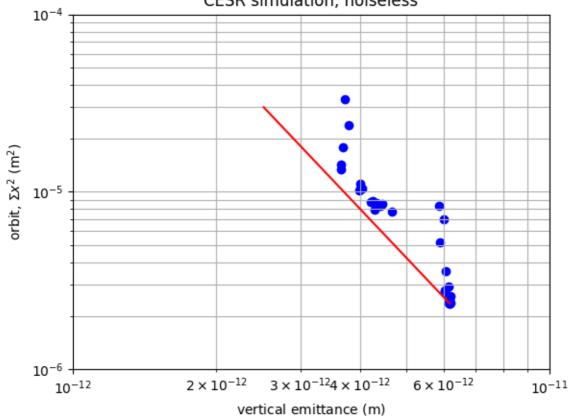


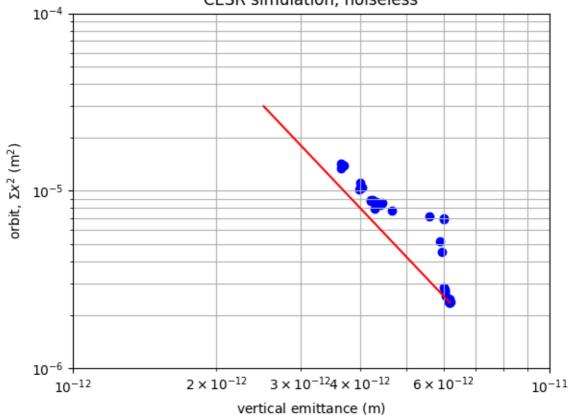


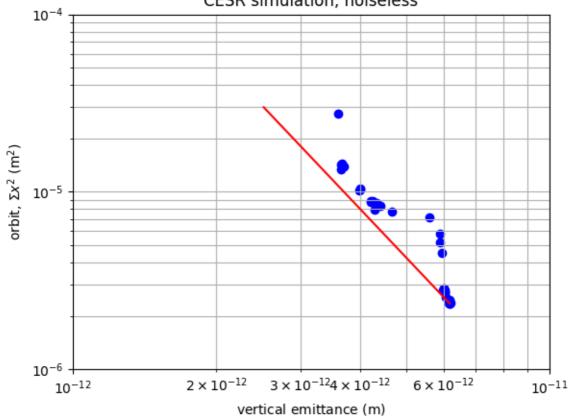


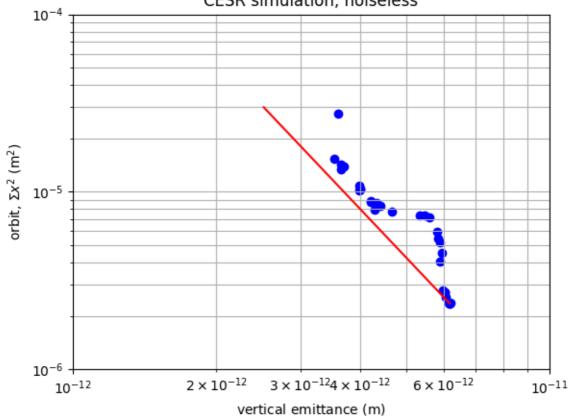


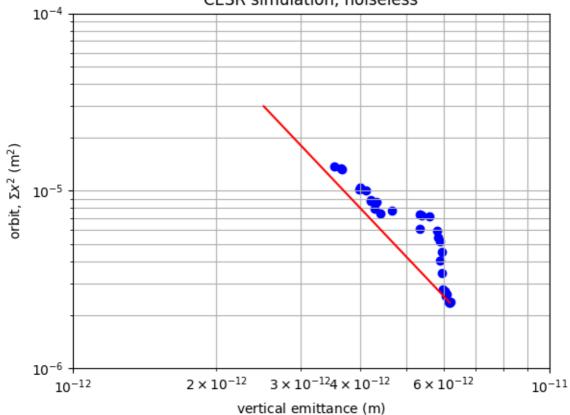


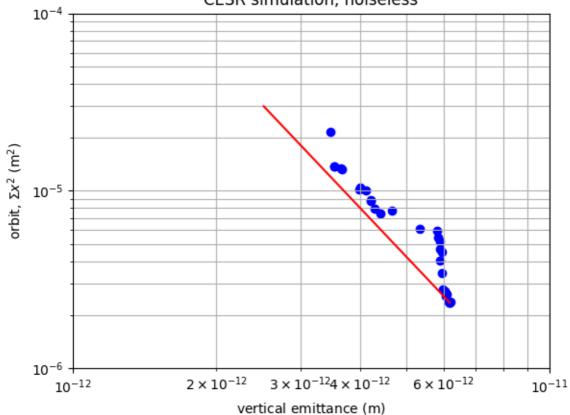


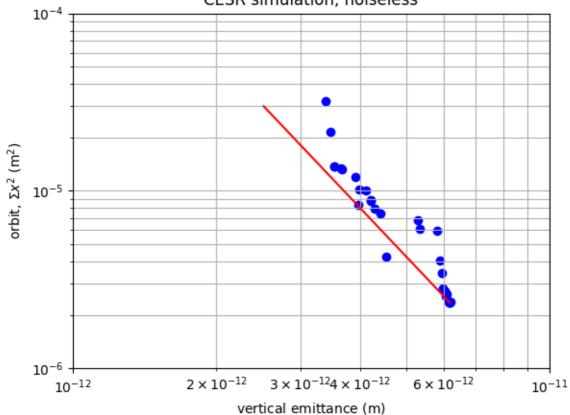


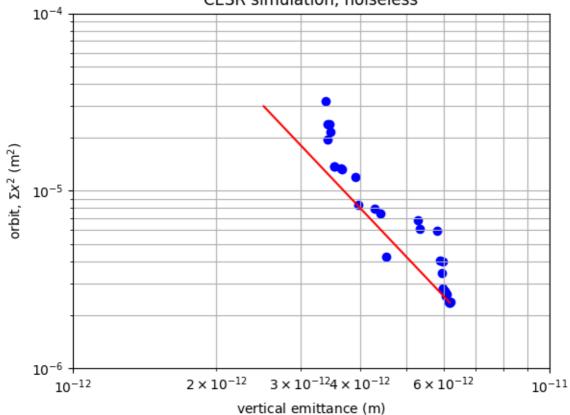


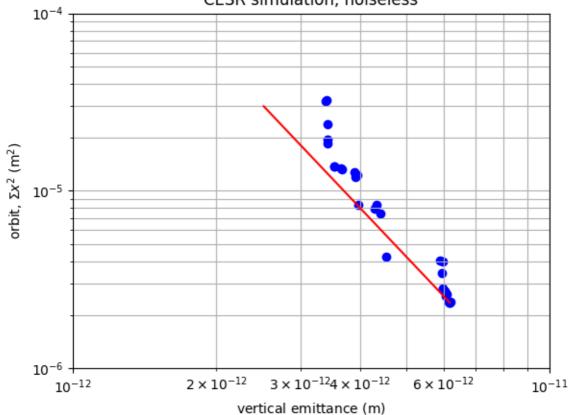


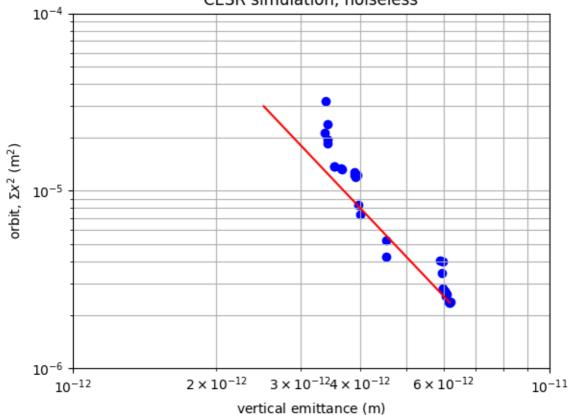


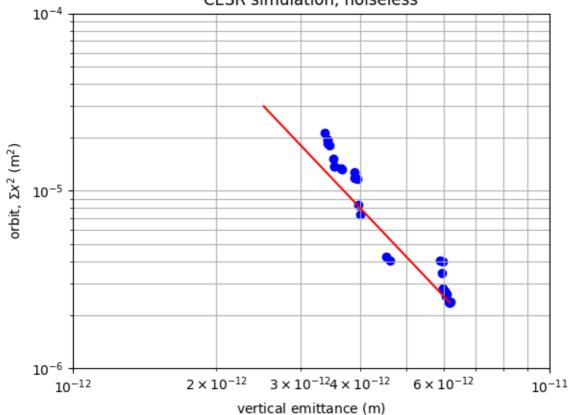


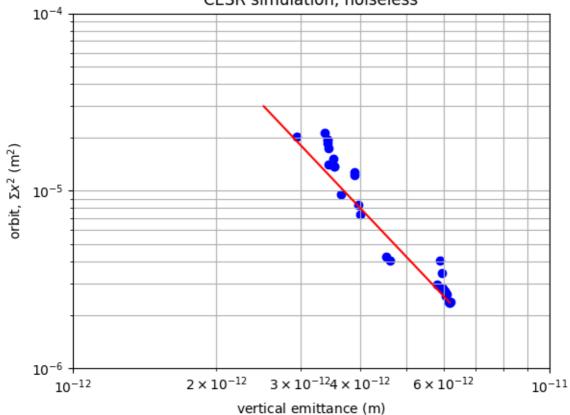


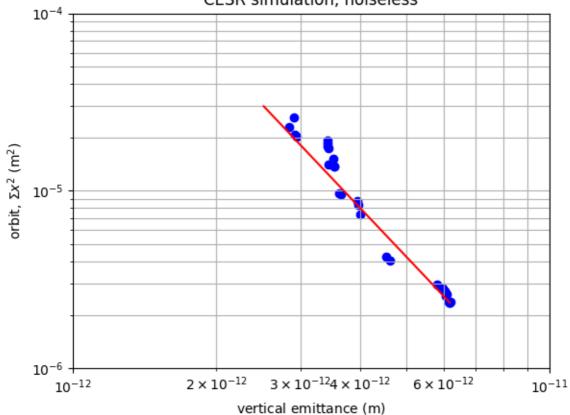


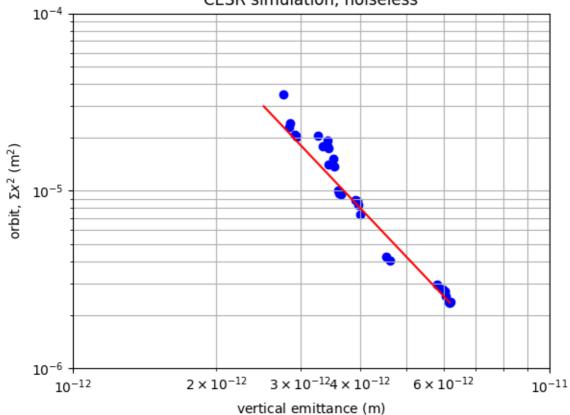


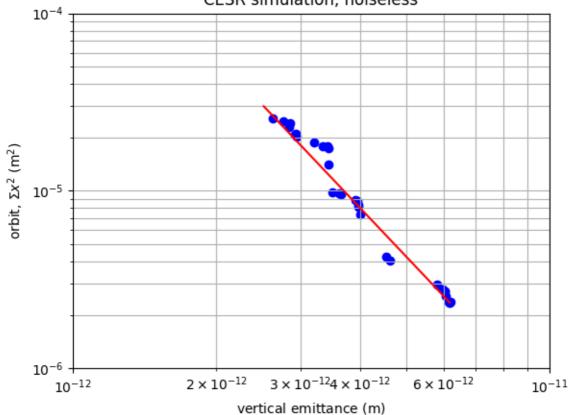


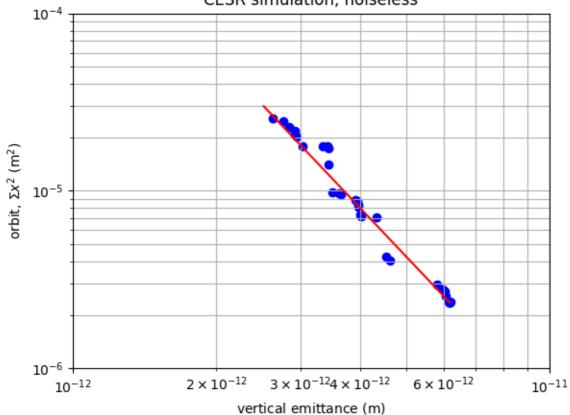


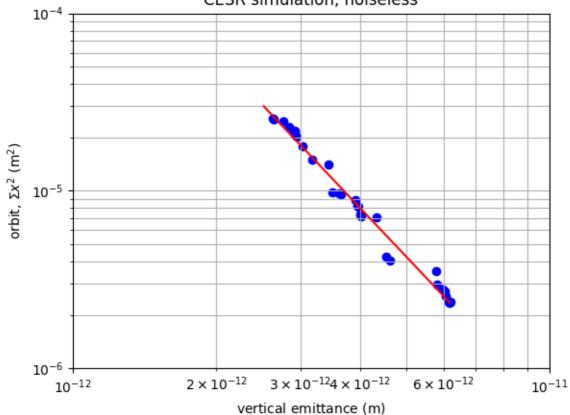


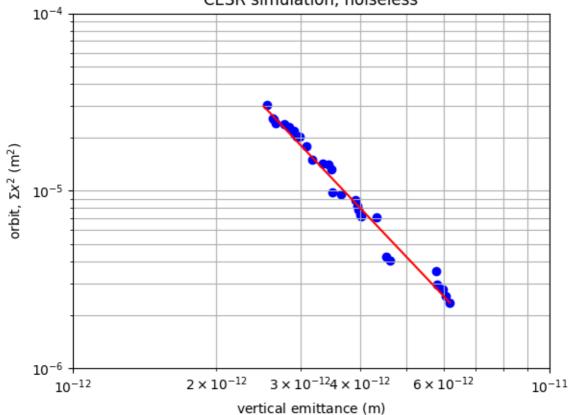


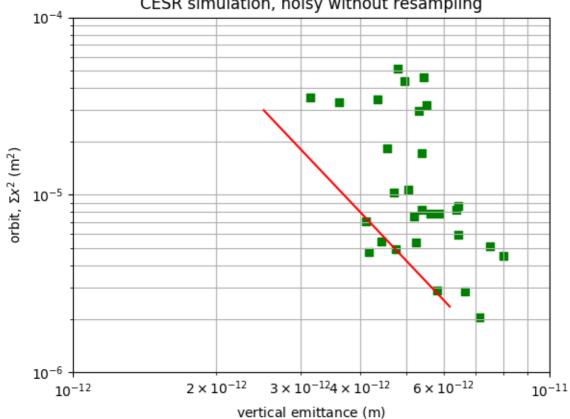


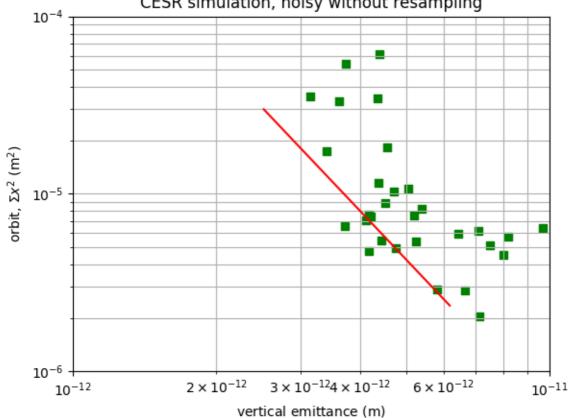


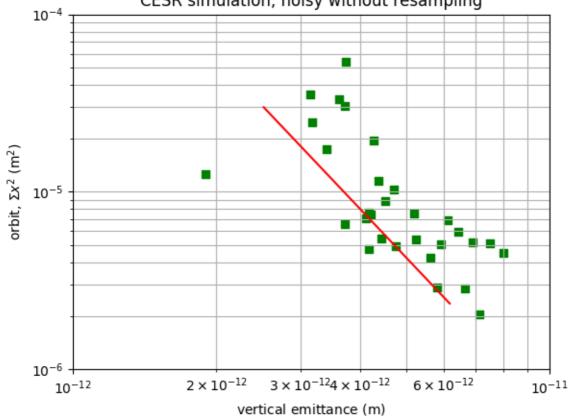


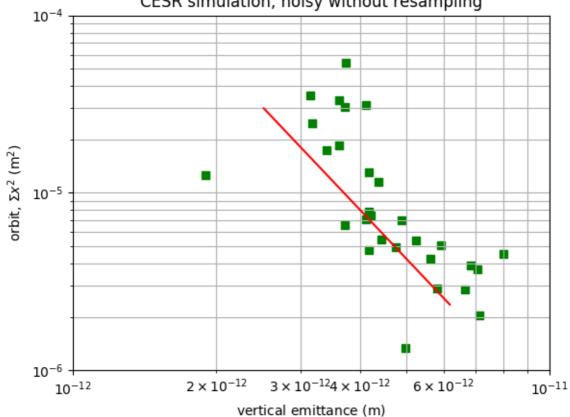


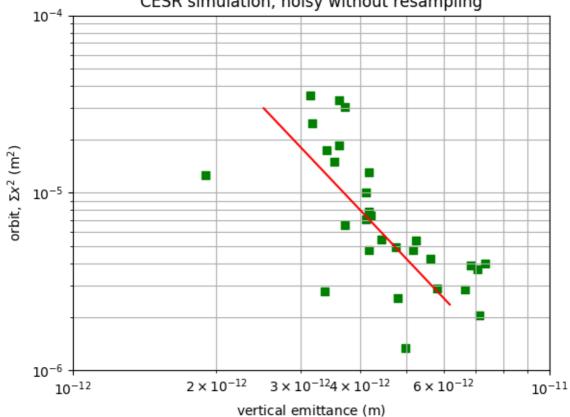


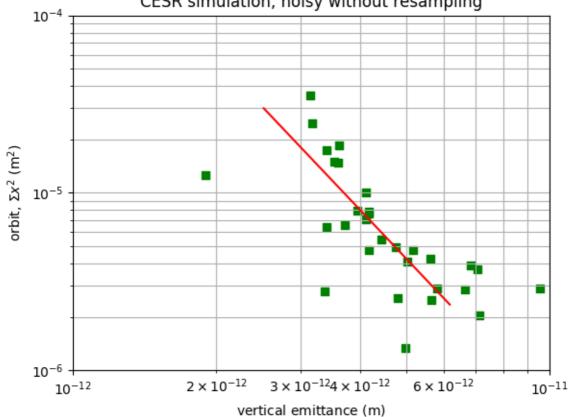


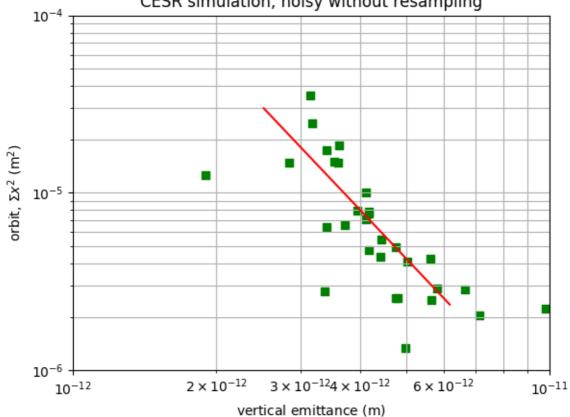


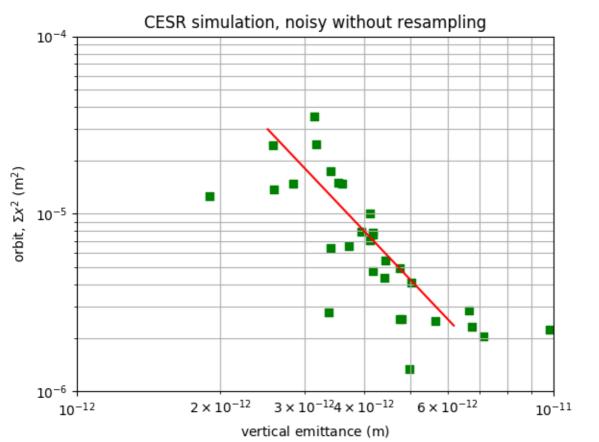


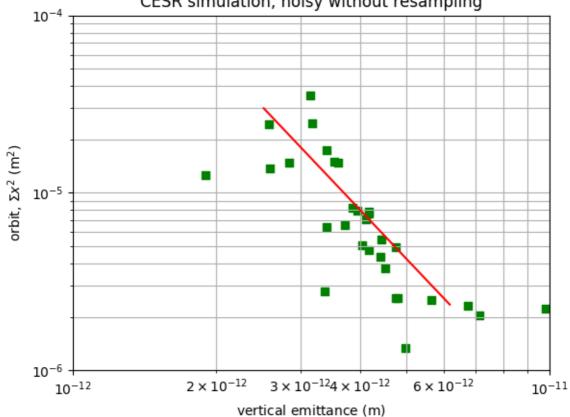


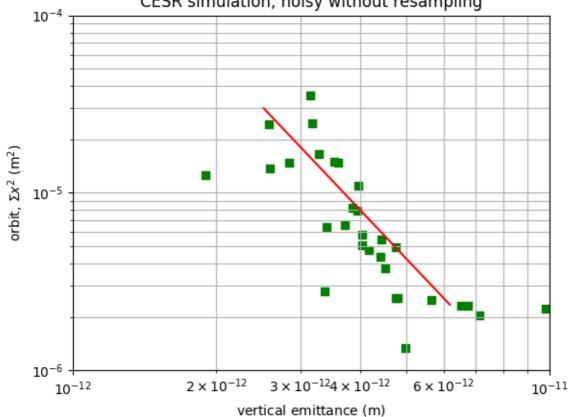


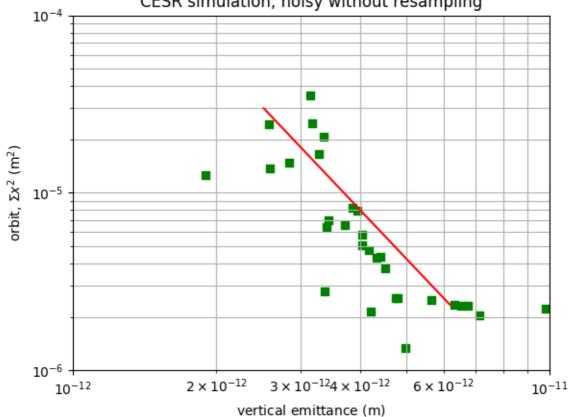


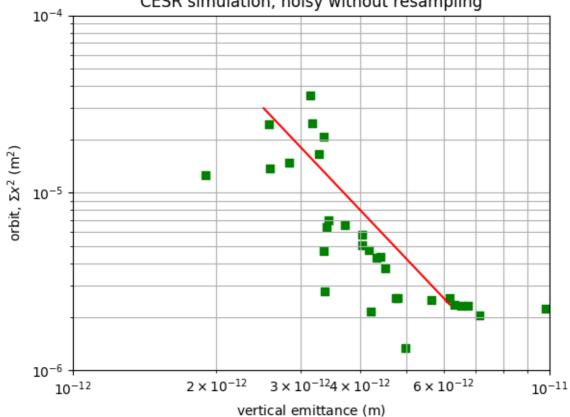


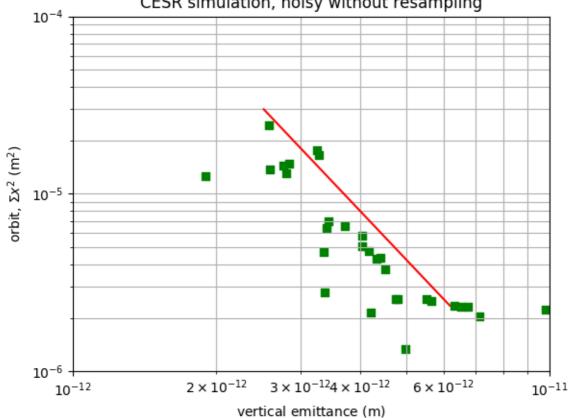


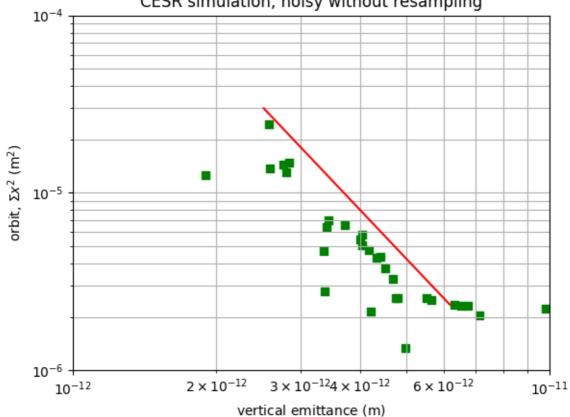


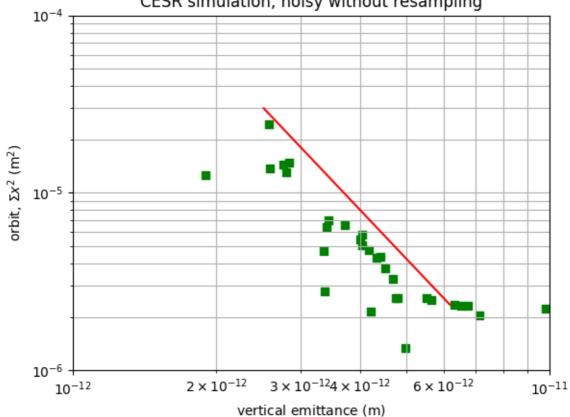


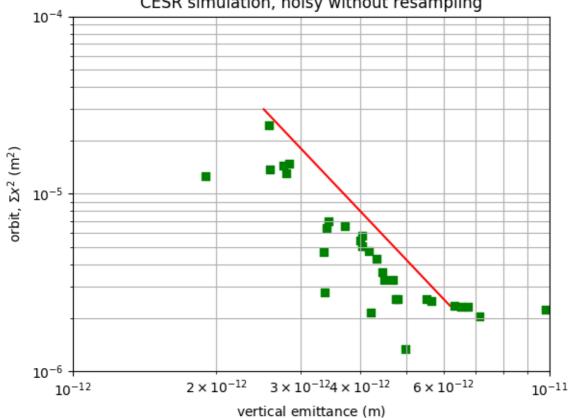


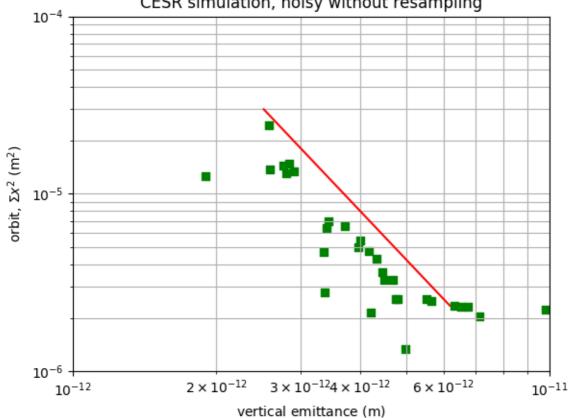


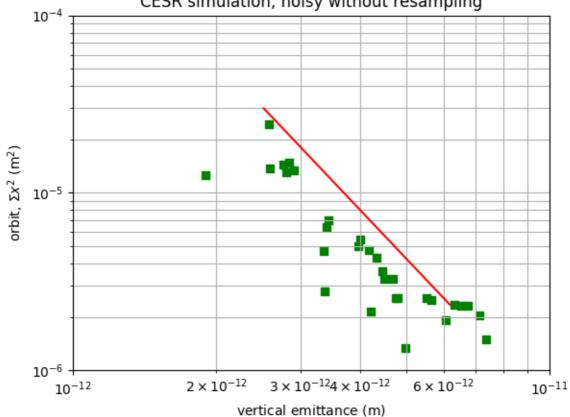


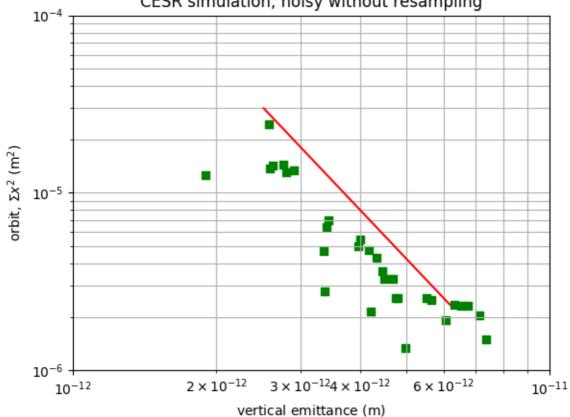


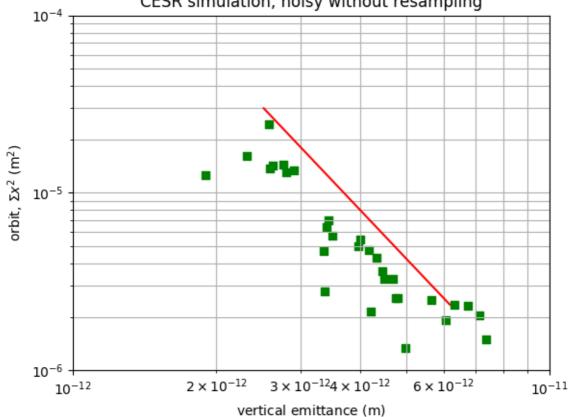


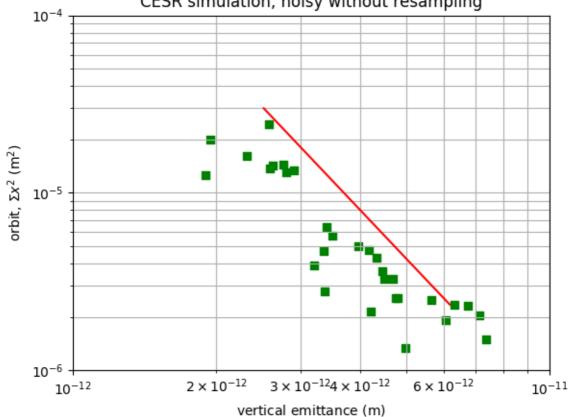


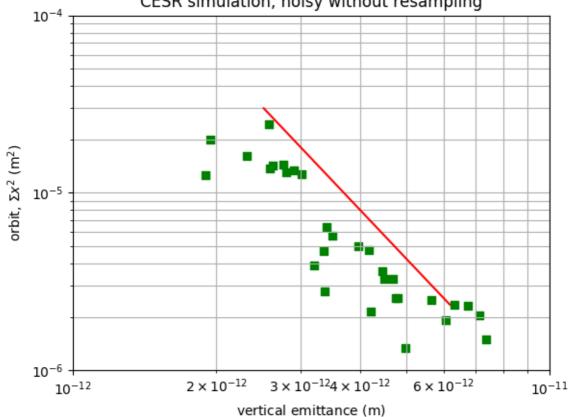


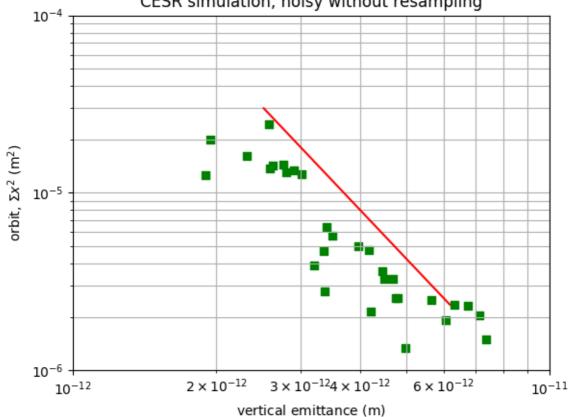


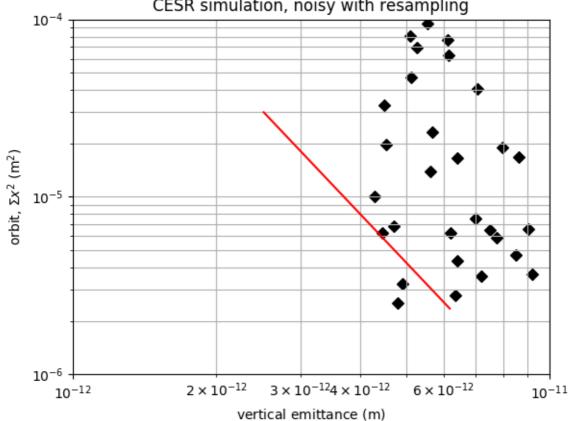


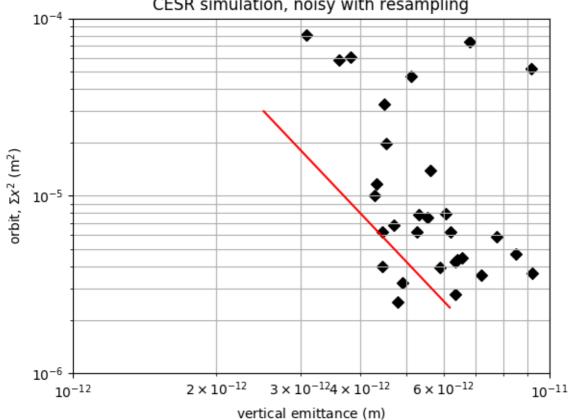


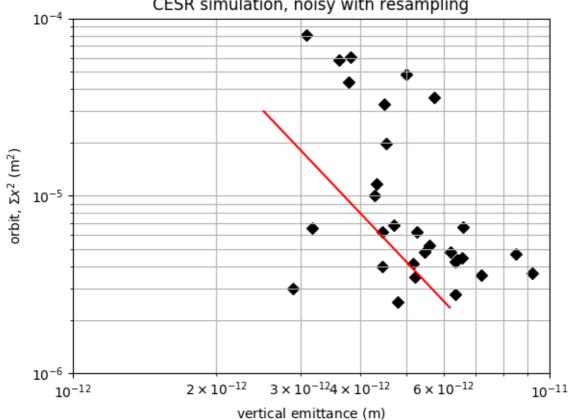


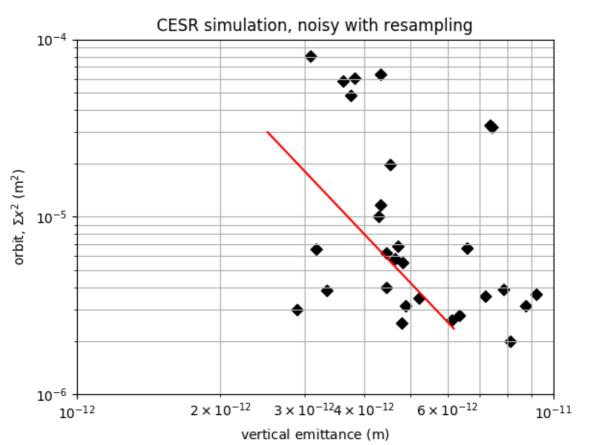


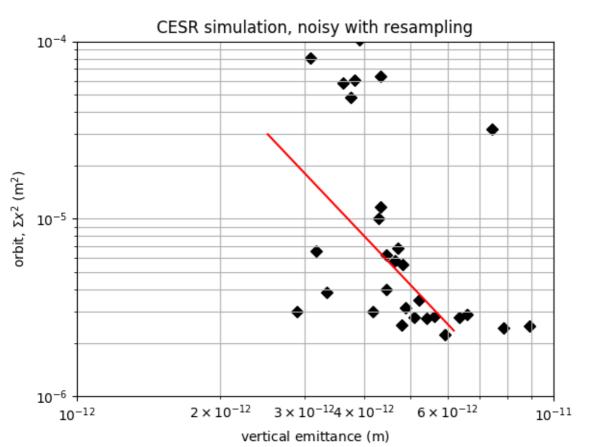


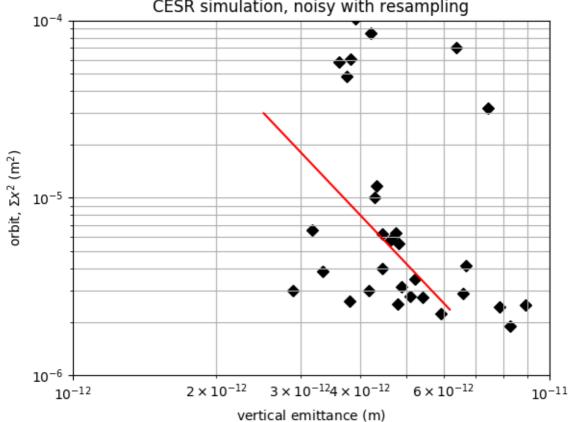


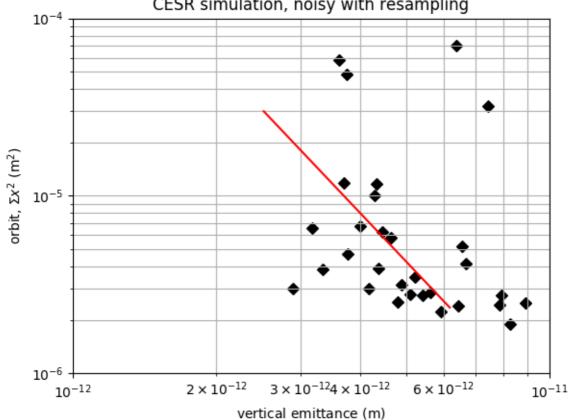


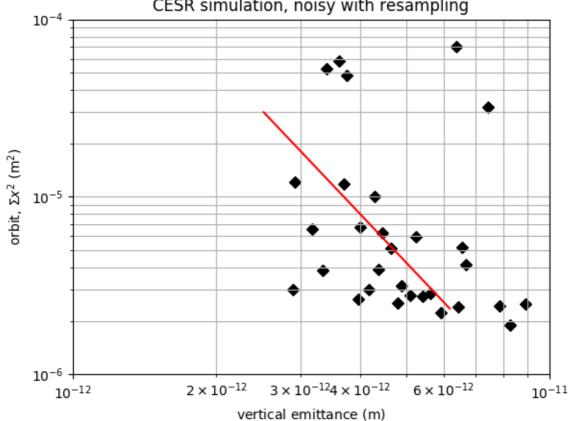


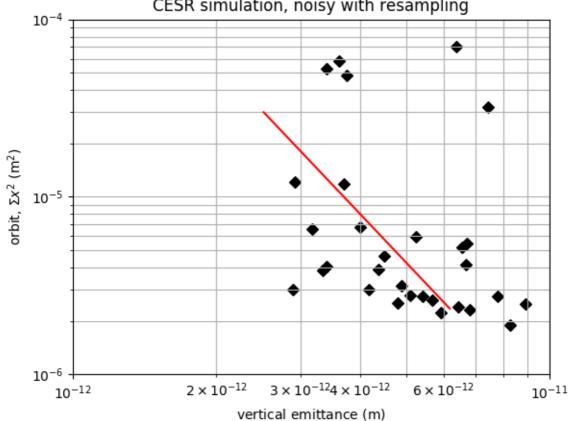


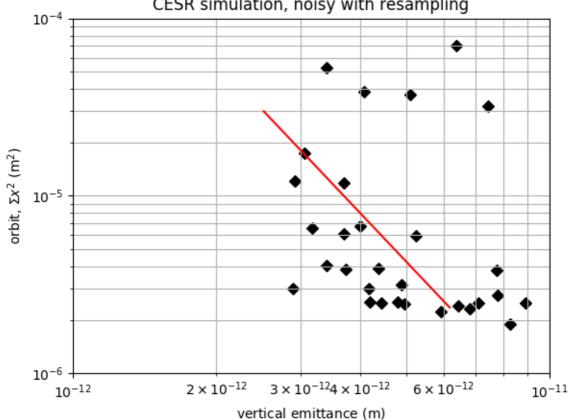


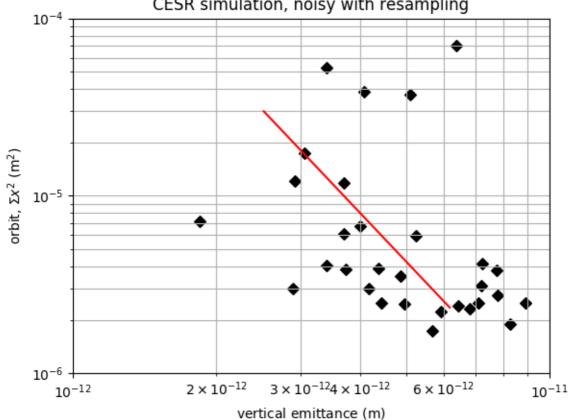


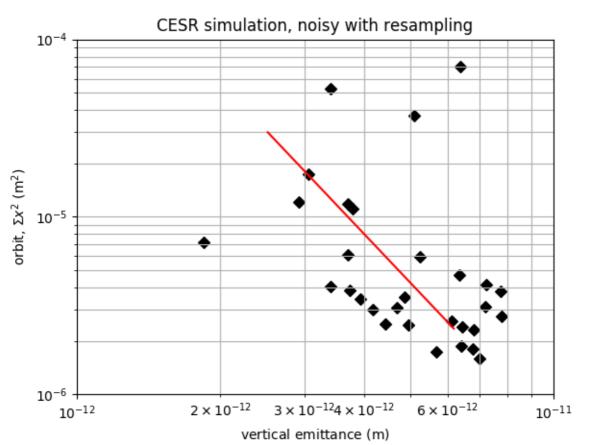


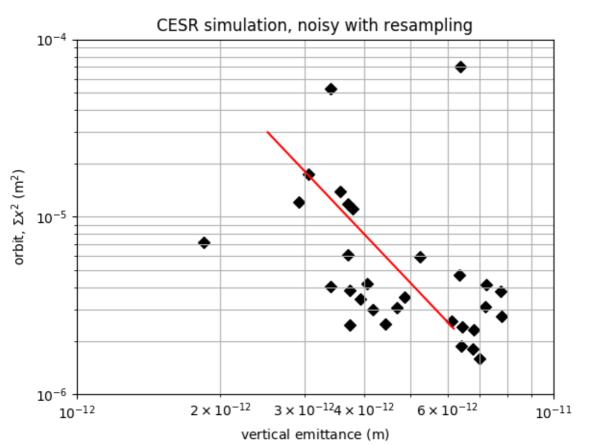


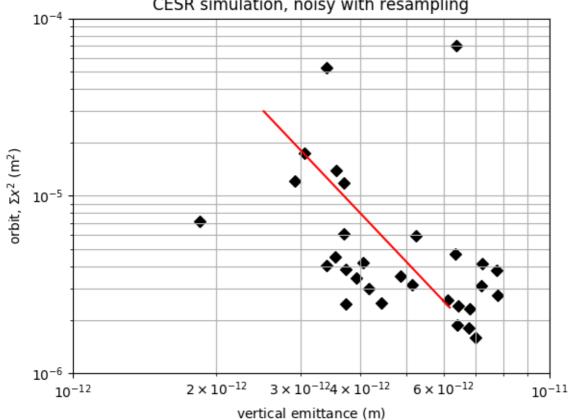


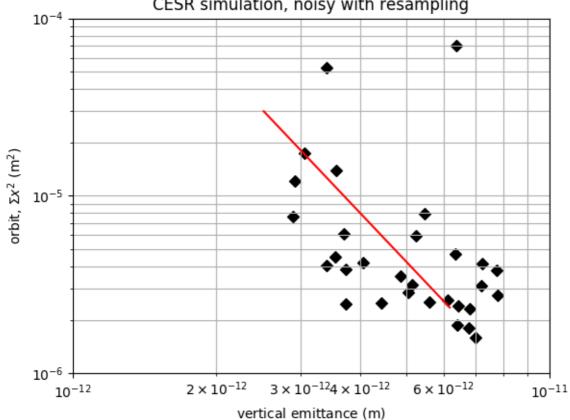


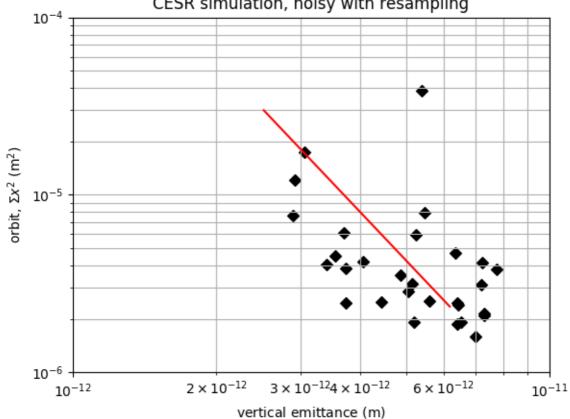


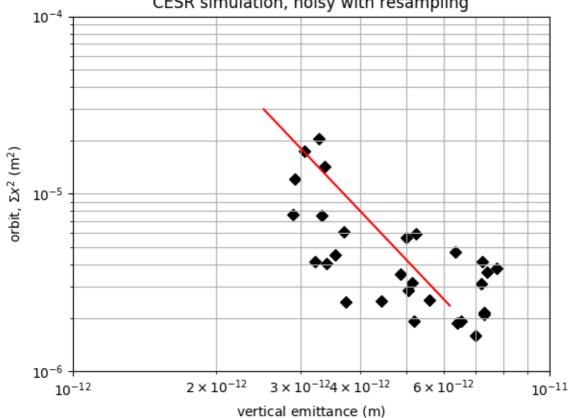


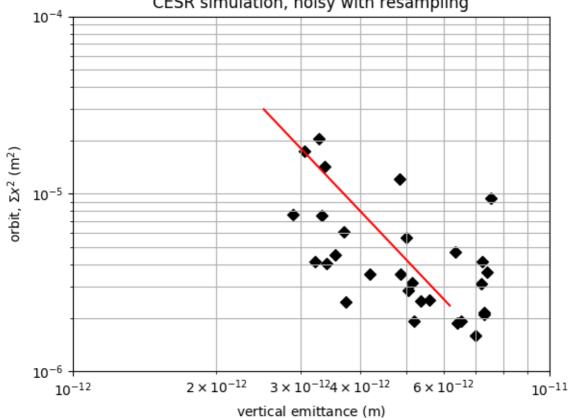


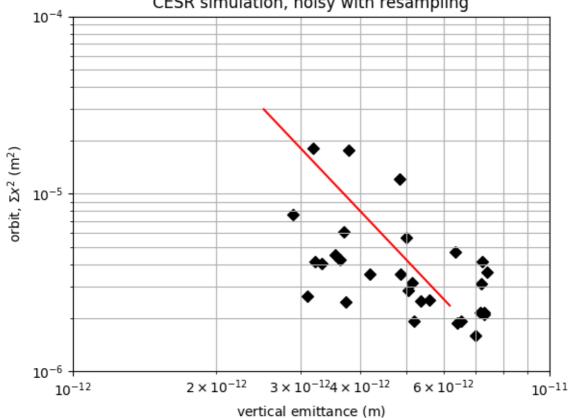


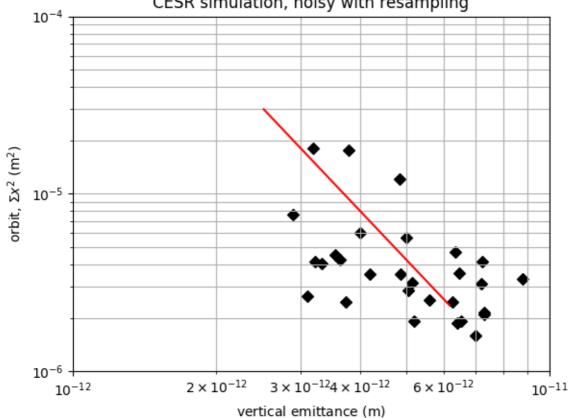


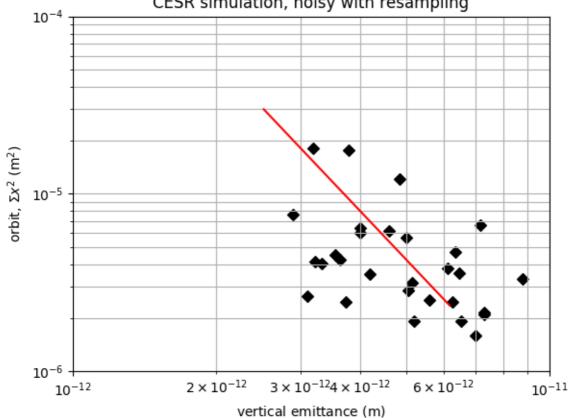


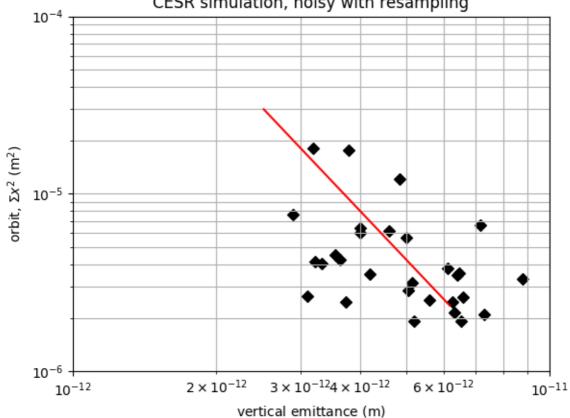


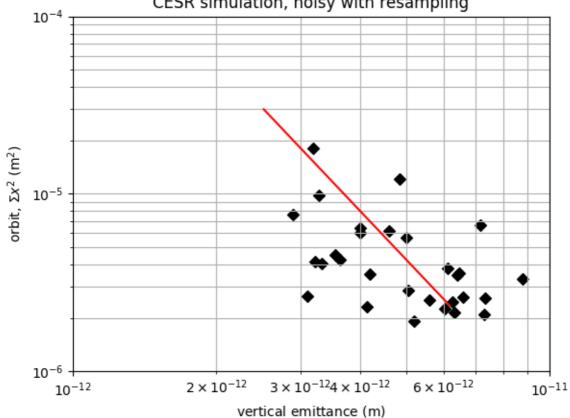










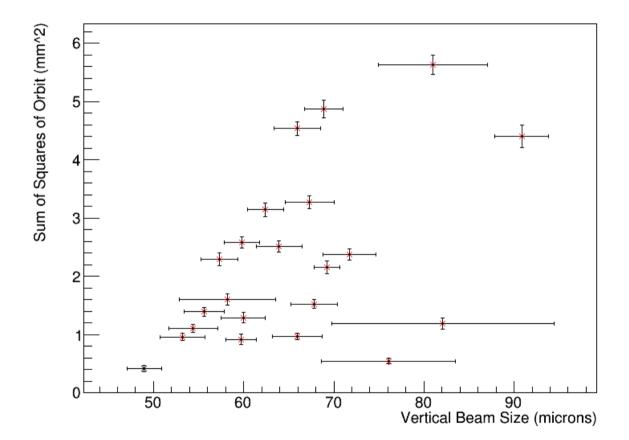


- Needed: a model-agnostic search for optimal performance *trade-offs*
- We tested an elitist genetic algorithm with resampling on bdad simulations of CESR
- Solution set shows randomness but converges in statistics
- Numerical evidence that power-law fit to solution set is an unbiased estimate of tradeoff front



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Preliminary Results





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Final Thoughts

- Any real-life online optimization *metaheuristic* is likely to be a combination of model-cognizant and modelagnostic parts;
- Machine safety needs to "filter" trial solutions preventing them from adopting forbidden states;
- Noise handling and maximizing throughput are always key issues;
- CESR is an ideal platform to deploy new kinds of online optimization strategies, including AI and stochastic algorithms.

