

### Rationale

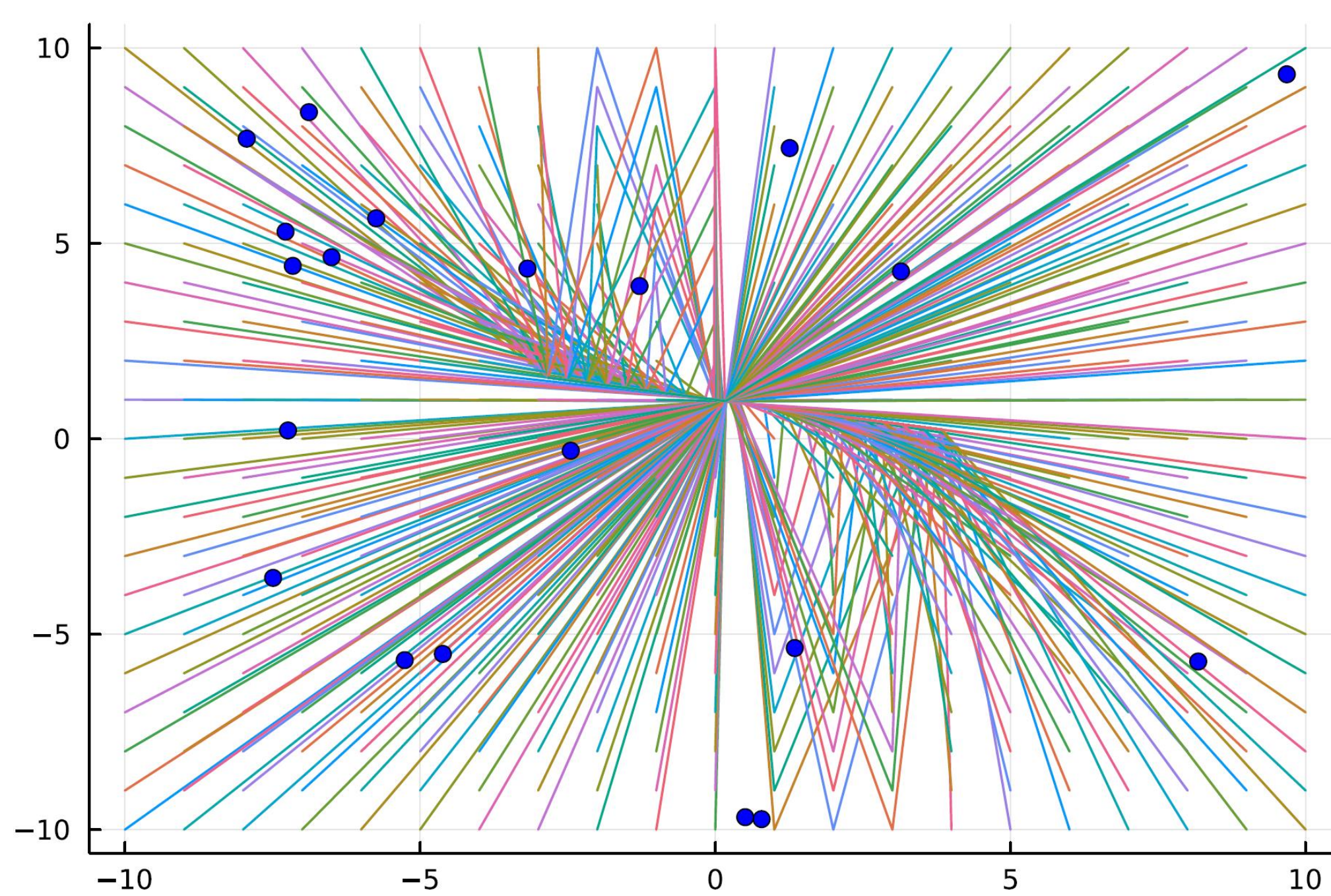
Each new generation of GPUs contain smaller number formats with a smaller *representable range*. Analysing how algorithms will behave with these smaller formats is essential, since **algorithms that previously worked in higher precision may fail to converge to the correct result** or become inaccurate when using these smaller number formats. However, the majority of the prior numerical analysis of algorithms in floating-point has ignored the representable range, since the formats found in modern computers have a representable range which is large enough to not affect most computations.

### Proposed Method

We propose analyzing the representable range necessary to implement specific algorithms by treating the algorithm as a non-linear discrete-time dynamical system and then performing reachability analysis on a Koopman operator linearization of the algorithm. Using the data-driven Koopman operator estimation for this analysis allows for **building the Koopman operator of complex, or even black-box, numerical codes** by only capturing the necessary algorithm state information at each iteration. The learned Koopman operator is then used to **compute the reachable sets of the iterative method**, which then provide the relevant information to determine what data types will have a suitable representable range.

### Illustrative Example

#### Sampled trajectories and RBF centers



#### Setup

- Gauss-Seidel stationary iterative method to solve  $Ax = b$  with  $A = \begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix}$ ,  $b = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$
- Use 20 Gaussian Radial Basis Functions with randomly placed centers and original system states as observables
- Construct Koopman operator using trajectory samples in the space  $[-10, 10] \times [-10, 10]$
- Approximate reachable set using Monte Carlo simulation of learned Koopman operator's trajectories

#### Results

- Identify  $x_2 \geq 0$  as an invariant set – stays positive if the number starts positive
- Approximated reachable invariant sets can inform required number format -  $x_2$  could be unsigned number while  $x_1$  must be a signed number

### Further work

This work is very preliminary and there are many open questions and possible research directions to improve the method and analysis:

- How to choose the observables for complicated non-linear/blackbox algorithms – **learned observables?**
- Can the algorithm's Koopman operator generalize to classes of problems?
- Formal guarantees in the reachability analysis
  - Use zonotopes in observable space
  - Reachability directly during Koopman operator estimation
- Integrating this method into a design or co-design workflow to optimize the number format or numerical method

#### Simulated reachability from initial conditions using the Koopman operator

