# Discovery of differential equations using probabilistic grammars: Appendix 

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#### Abstract

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## A Learning curve estimation

ProGED performs Monte-Carlo sampling of candidate equations from a distribution defined by a PCFG. We obtain performance estimates, such as learning curves and the computation time required to find the solution, through bootstrapped resampling.

As the result of running ProGED, we have a list of $N$ candidate equations, each with an associated error-of-fit and a probability of the right-hand side expression derived from the grammar. We randomly sample a sequence of $N$ models from this list without repetition and following the expression probabilities. We then calculate the cumulative minimum of error across this sequence and thus obtain a single learning curve. We repeat this procedure many (at least 1000) times and average the learning curves. The resulting curve estimates the expected best error-of-fit ProGED would achieve with a given number of models sampled. In this way, we simulate repeating the random sampling experiment many times.

## B Reconstructed equations

Table 1. Each methods' best reconstructed model for the VDP data set.

| method | obs. Reconstructed ODEs |
| :---: | :---: |
| SINDy | XY $\dot{x}=0.99998 y$ |
|  | XY $\dot{y}=-0.99998 x+0.49991 y-0.49991 x^{2} y$ |
| L-ODEfind | XY $\dot{x}=0.99990 y$ |
|  | XY $\dot{y}=-0.99995 x+0.49980 y-0.49986 x^{2} y$ |
|  | $\mathrm{X} \quad \ddot{x}=-0.99993 x+0.49918 \dot{x}-0.49961 x^{2} \dot{x}$ |
|  | $\overline{\mathrm{Y}} \quad \ddot{y}=0.01273+0.76615 \dot{y}-0.43317 y^{3}+0.80264 y^{2} \dot{y}-0.06486 \dot{y}^{3}$ |
| GPoM | XY $\dot{x}=0.99975 y$ |
|  | XY $\dot{y}=-1.00014 x+0.50107 y-0.49953 x^{2} y$ |
|  | X $\quad \dot{x}=1 y$ |
|  | Х $\dot{y}=-0.99992 x+0.50070 y-0.49965 x^{2} y$ |
|  | $\dot{x}=1 y$ |
|  | Y $\dot{y}=0.01552+0.44866 x-1.22201 y-0.56297 y^{3}+$ |
|  | $+0.88215 x^{2} y+0.10693 x y^{2}+0.03426 y^{3}$ |
| ProGED | XY $\dot{x}=0.99999 y$ |
|  | ХY $\dot{y}=-0.99998 x+0.49991 y-0.49991 x^{2} y$ |
|  | XY $\dot{x}=1.00001 y$ |
|  | XY $\dot{y}=-0.99998 x+0.49997 y-0.49998 x^{2} y$ |
|  | X $\quad \dot{x}=0.24209 y$ |
|  | X $\dot{y}=-4.17027 x+0.54101 y-0.54491 x^{2} y$ |
|  | $\mathrm{Y} \quad \dot{x}=0.56267 y$ |
|  | Y $\dot{y}=-1.78159 x+0.50896 y-1.59903 x^{2} y$ |

Table 2. The first-order variants of the second-order ODEs reconstructed by the LODEfind method and reported in Table 1.

| Obs. Inverted ODEs |
| :---: | :--- |
| $\mathrm{X} \quad$$\dot{x}=y$ <br> $\dot{y}=-0.99993 x+0.49918 y-0.49961 x^{2} y$ |
| $\mathrm{Y} \quad$$\dot{x}=0.01273+0.76615 x-0.43317 y^{3}+0.80264 y^{2} x-0.06486 x^{3}$ <br> $\dot{y}=x$ |

