## 2. Instantiate and train DySymNet via backpropagation and weight pruning

DYSYMNET:  

$$x_1$$

$$x_2$$

$$id$$

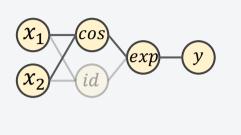
$$y_1 = sin(1.3x_1 + 0.5x_2) + 1.5x_2$$

$$x_{1}$$

$$x_{2}$$

$$exp$$

$$y_{i} = cos(x_{1}) + exp(0.6x_{2})$$



 $y_N = exp(cos(0.9x_1 + 1.2x_2))$ 

71. Sample N DYSYMNET descriptions autoregressively via RNN 
$$\frac{Num.\ of}{ops\ of\ L_{1:n}}$$
  $\frac{Operators\ of}{L_{1:n}}$   $\frac{Operators\ of}{L_{1:n}}$   $\frac{L_n}{L_n}$ 

$$\tilde{f}_{i}^{*}(x) = BFGS(y_{i})$$

$$\nabla_{\theta} J(\theta; \varepsilon) \approx \frac{1}{\varepsilon N} \sum_{i=1}^{N} \left[ R(\tilde{f}_{i}^{*} | \mathcal{D}) - R_{\varepsilon} \right] \cdot \mathbf{1}_{R \geq R_{\varepsilon}} \nabla_{\theta} \log p(\tilde{f}_{i}^{*} | \theta)$$

$$\nabla_{\theta} \mathcal{H}(\theta) = \lambda_{H} \sum_{i=1}^{N} \nabla_{\theta} \mathcal{H}(\tilde{f}_{i}^{*} | \theta)$$