

LyaNet: A Lyapunov Framework for Training Neural ODEs

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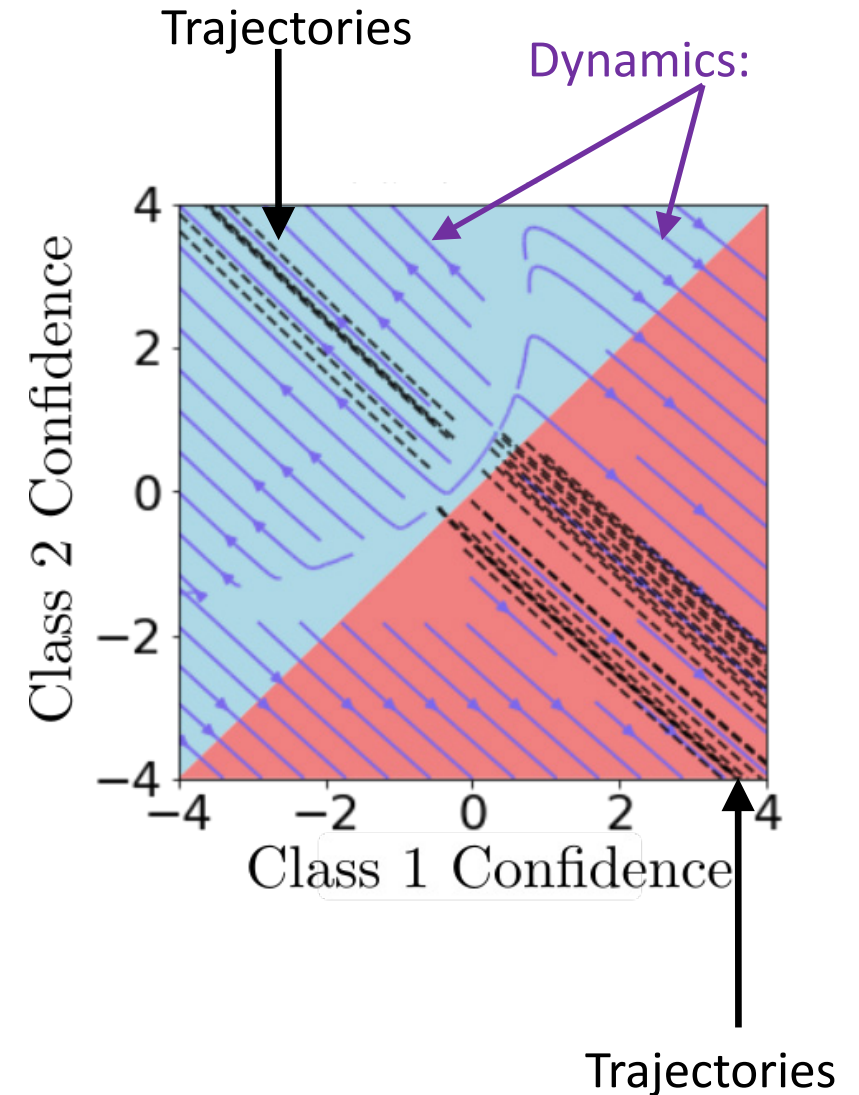


Challenges

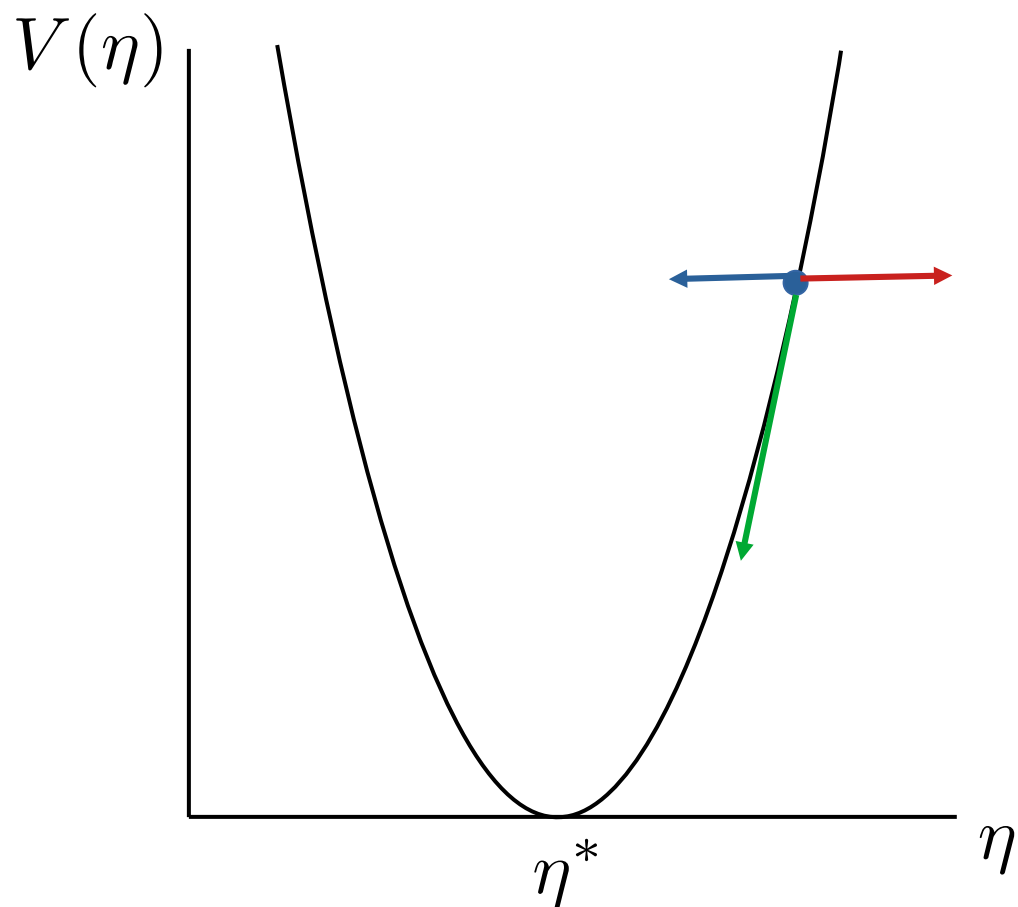
- Dynamics for single data point (x, y) .
- 2-Dimensional Hidden Logits State
- Binary Classification: **Red Class** is correct

Undesirable Chaotic Dynamics:

- Dynamics often flow in the wrong direction
- Solutions are "fragile"
- Poor Generalization



Idea: Enforce Always Making Progress



Progress

$$\frac{\partial V}{\partial \eta} f(\eta) \leq -\kappa V(\eta)$$

No Progress

$$\frac{\partial V}{\partial \eta} f(\eta) > -\kappa V(\eta)$$

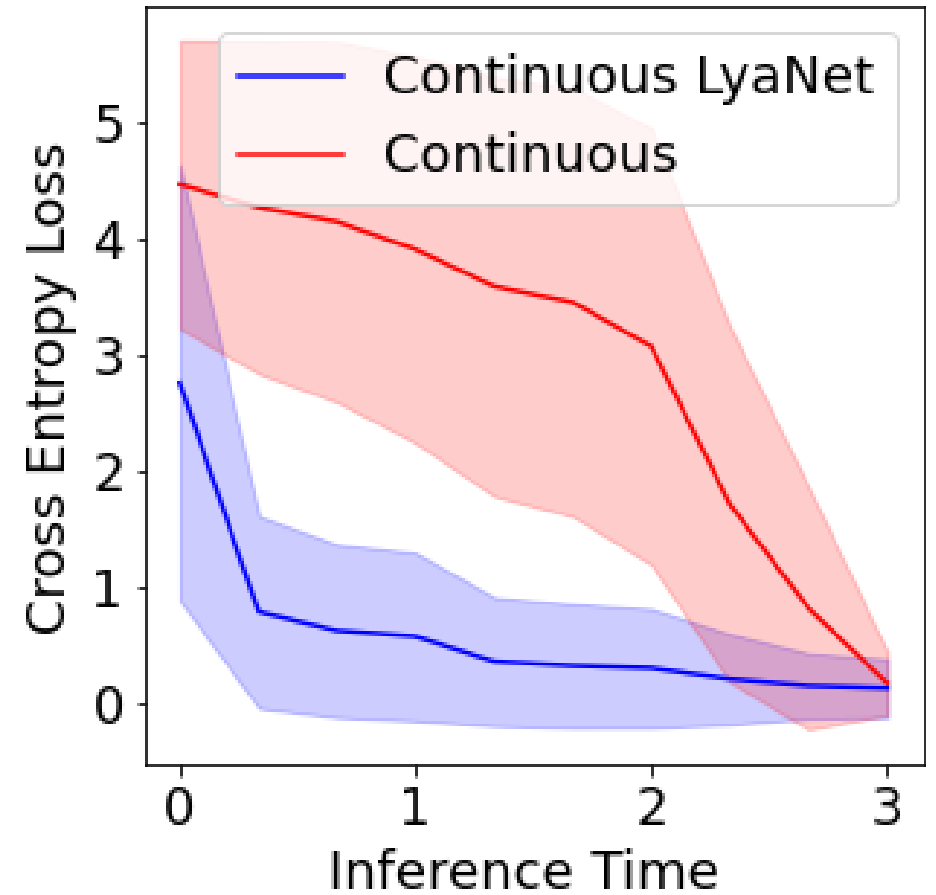
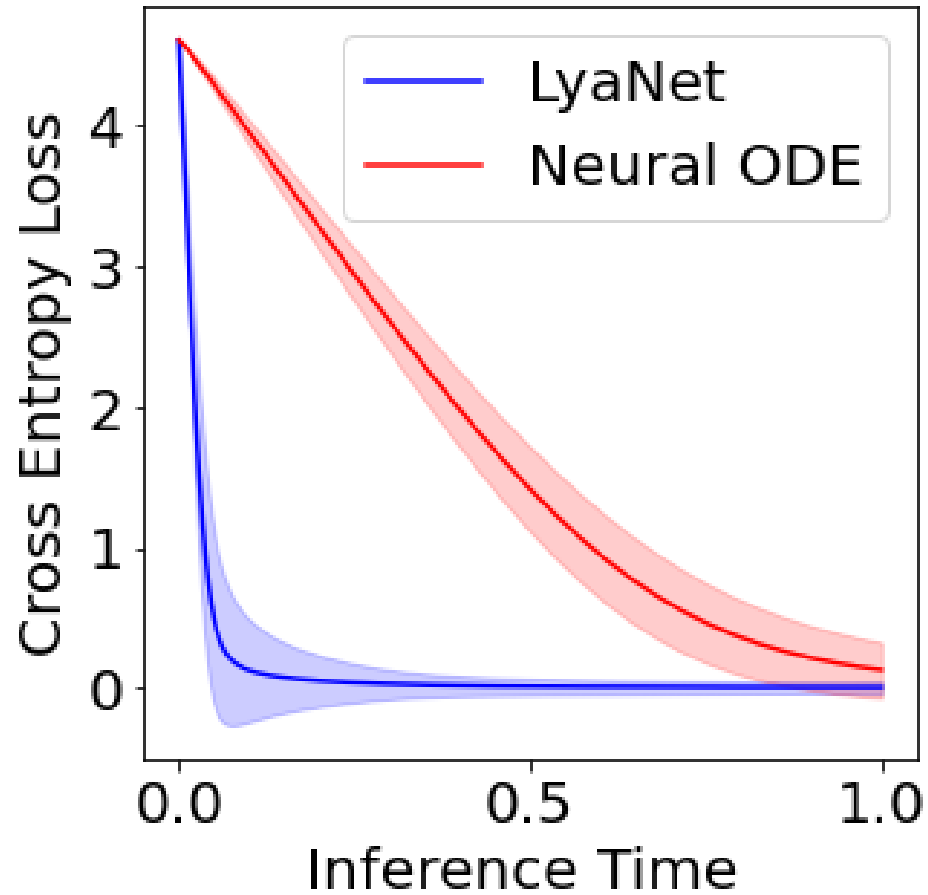
Measuring Progress on Supervised Loss

For a sample, define a potential function:

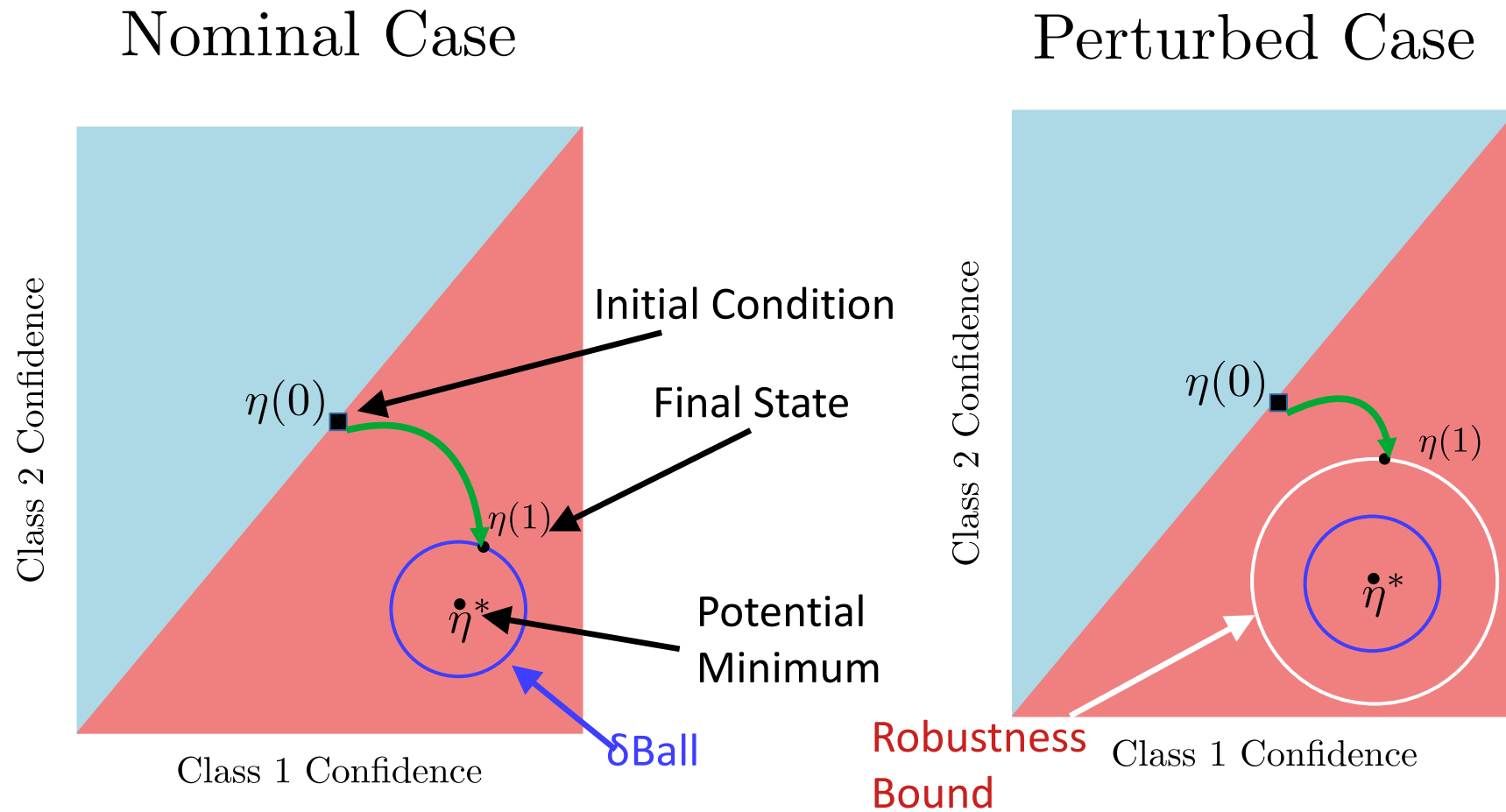
$$\begin{aligned} V_{\eta^*}(\eta) &= \mathcal{L}(\eta, \eta^*) \\ &= \|\eta - \eta^*\|_2^2 \end{aligned}$$

Similar potential function can be selected for Cross-Entropy.

Exponential Convergence (CIFAR-10)



Our Method: Robustness Guarantees (Illustrated)



Conclusions

- **LyaNet:** A novel approach for training Neural ODEs using Lyapunov Stability
- **Main benefits:**
 - Fast convergence of inference dynamics
 - Empirical and provable robustness
 - Allows adaptation of control theoretic tools to the learning context
- **Other Applications:**
 - Enforce constraints on the learned dynamics

