

Introduction to Deep Operator Networks

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Universal Approximation Theorem for operators

Let,

- ▶ σ - a continuous non-polynomial function
- ▶ X - Banach space
- ▶ $K_1 \in X$, $K_2 \in \mathbb{R}^d$ - two compact sets
- ▶ $V \in C(K_1)$ - a compact set in functional space of K_1
- ▶ G - non-linear continuous operator that maps $V \rightarrow C(K_2)$

Then for any $\epsilon > 0$, there are positive integers n, p, m , constants

$c_i^k, \varepsilon_{ij}^k, \theta_i^k, \zeta_k \in \mathbb{R}$, $w_k \in \mathbb{R}^d$, $x_j \in K_1$, $i = 1, \dots, n$, $k = 1, \dots, p$, $j = 1, \dots, m$ such that

$$\left| G(u)(y) - \underbrace{\sum_{k=1}^p \sum_{i=1}^n c_i^k \sigma \left(\sum_{j=1}^m \varepsilon_{ij}^k u(x_j) + \theta_i^k \right)}_{\text{branch}} \underbrace{\sigma(w_k \cdot y + \zeta_k)}_{\text{trunk}} \right| < \epsilon \quad (1)$$

holds for $u \in V$ and $y \in K_2$.

Universal Approx. Thm for operators

In other form,¹

$$\left| G(u)(y) - \sum_{k=1}^p \underbrace{br_k(u(x_1), u(x_2), \dots, u(x_m))}_{\text{branch}} \underbrace{tr_k(y)}_{\text{trunk}} \right| < \epsilon \quad (2)$$

where, x_1, \dots, x_m are called sensor points.

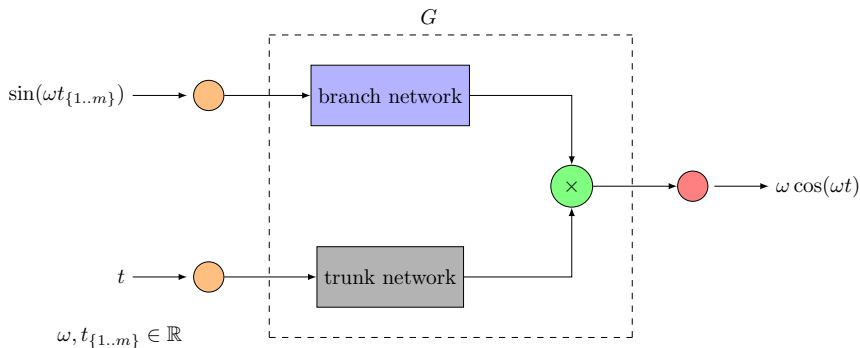
The functions $br()$, $tr()$, can be approximated by neural networks²

¹Goswami et al., “Physics-informed deep neural operator networks”.

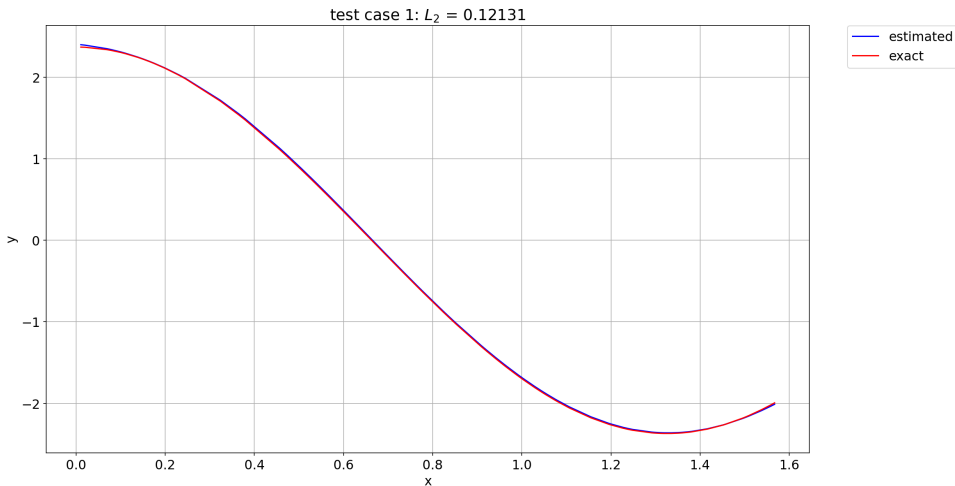
²Hornik, Stinchcombe, and White, “Multilayer feedforward networks are universal approximators”.

Data-driven deep-o-net example

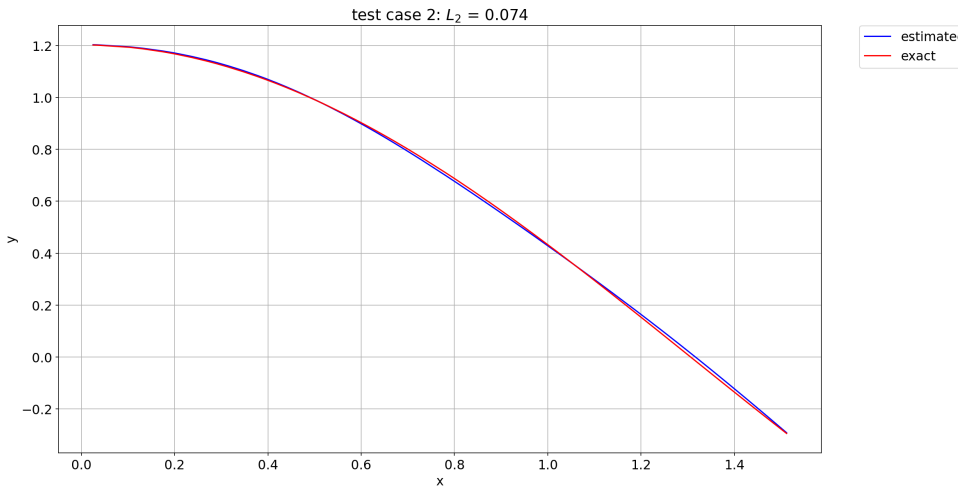
The operator $G : \sin(\omega t) \rightarrow \omega \cos(\omega t)$



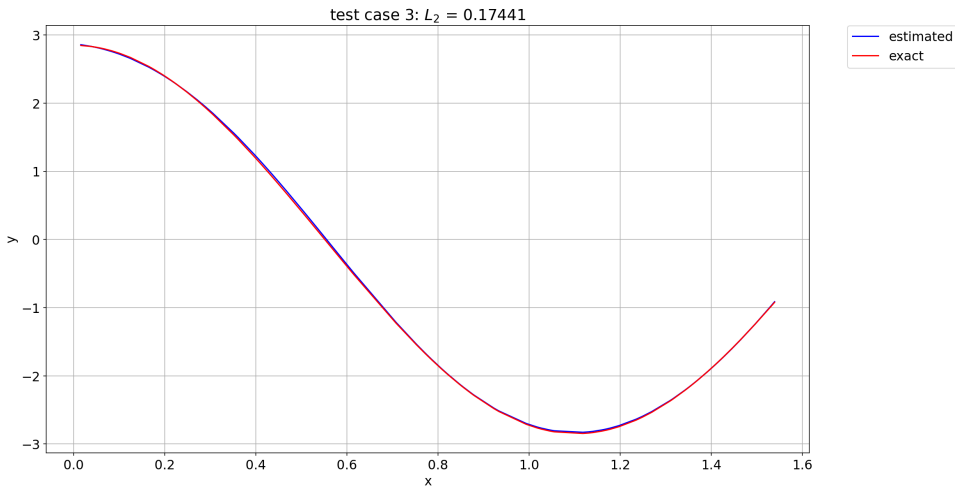
Data-driven deep-o-net example - results



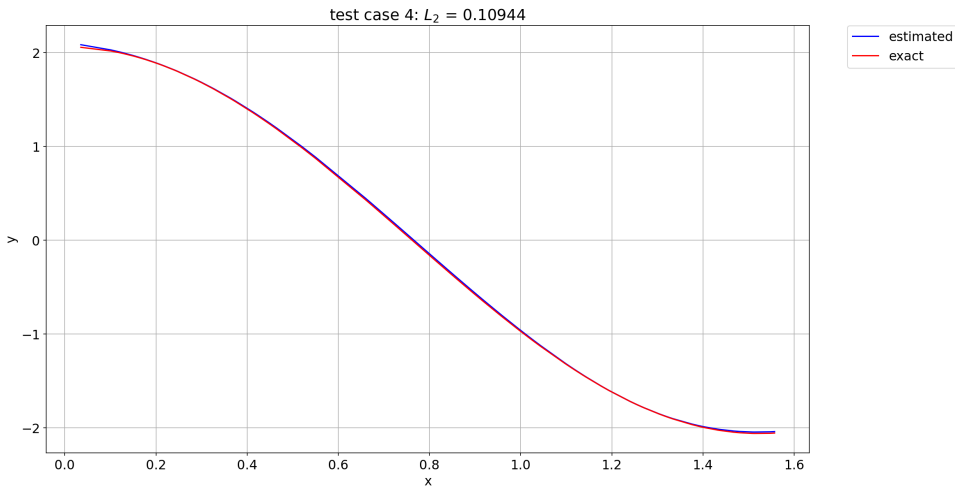
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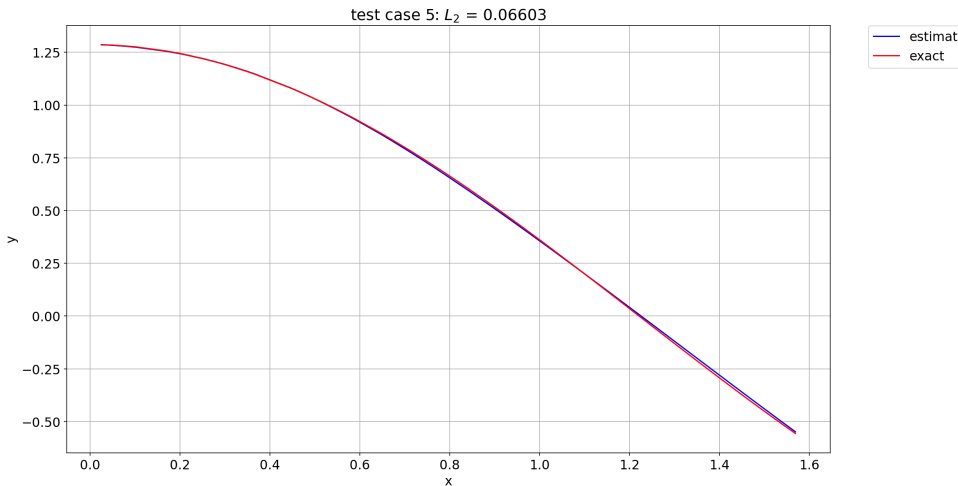
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Data-driven deep-o-net Experiment

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Mathematically not!

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With a trained model, tried evaluating

$$\hat{G} : \cos(\omega t) \rightarrow -\omega \sin(\omega t)$$

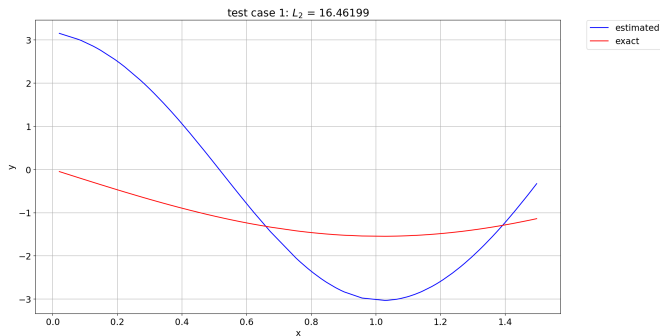
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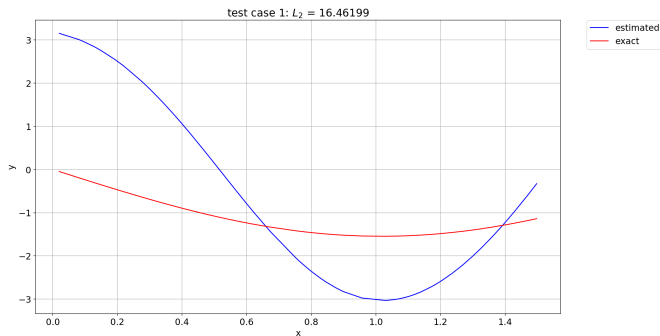
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Network approximates Operator with encoded input function

Physics Informed Deep-o-net

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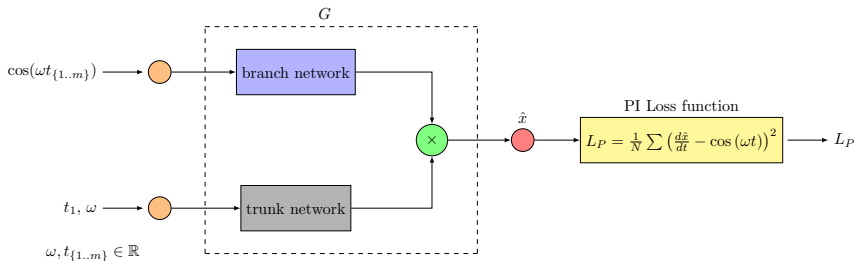
Tried with operator, $G : \cos(\omega t) \rightarrow x$

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with constraint, $\dot{x} = \cos(\omega t)$

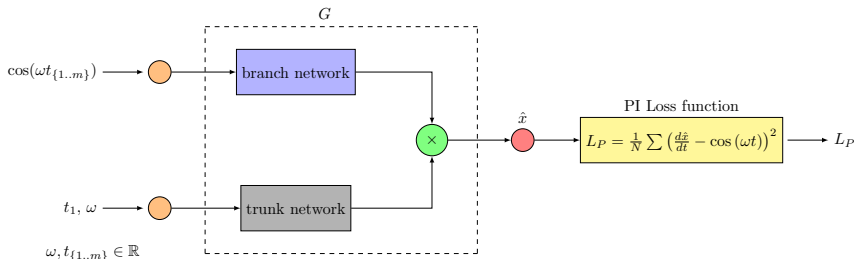
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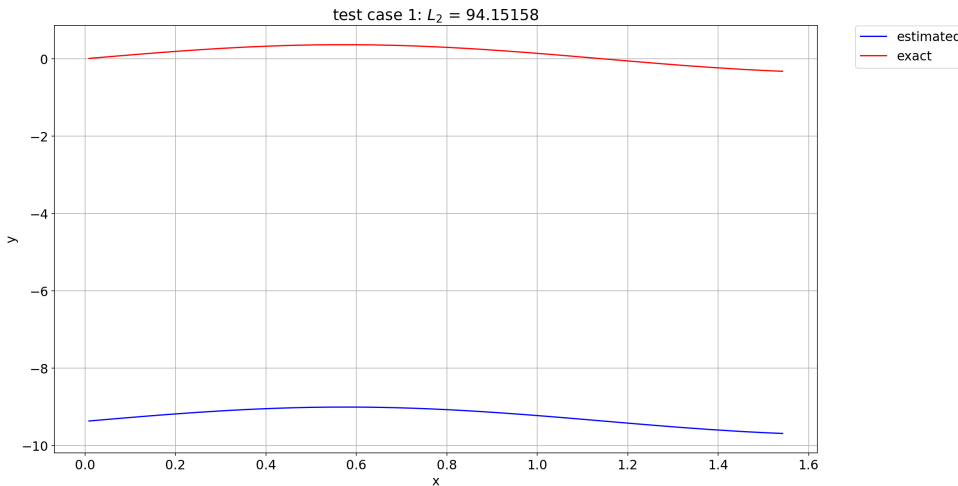
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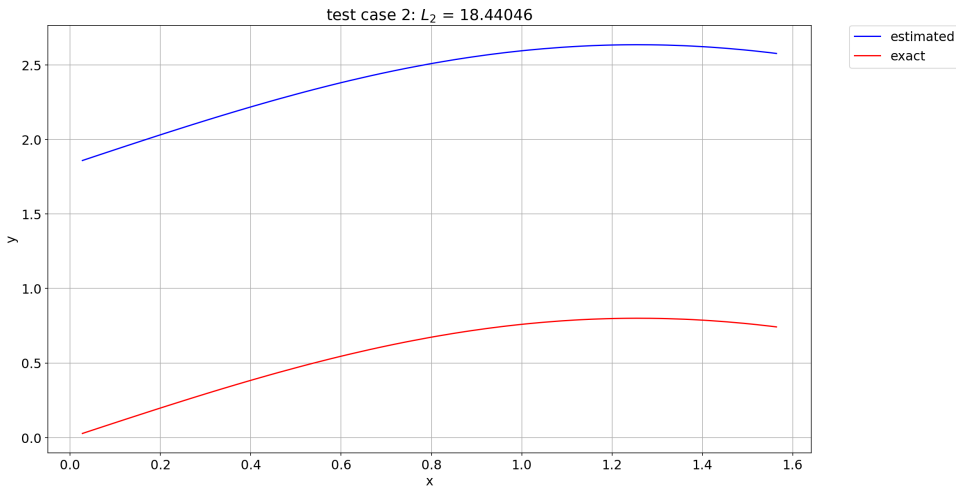


This includes automatic differentiation, \dot{x}

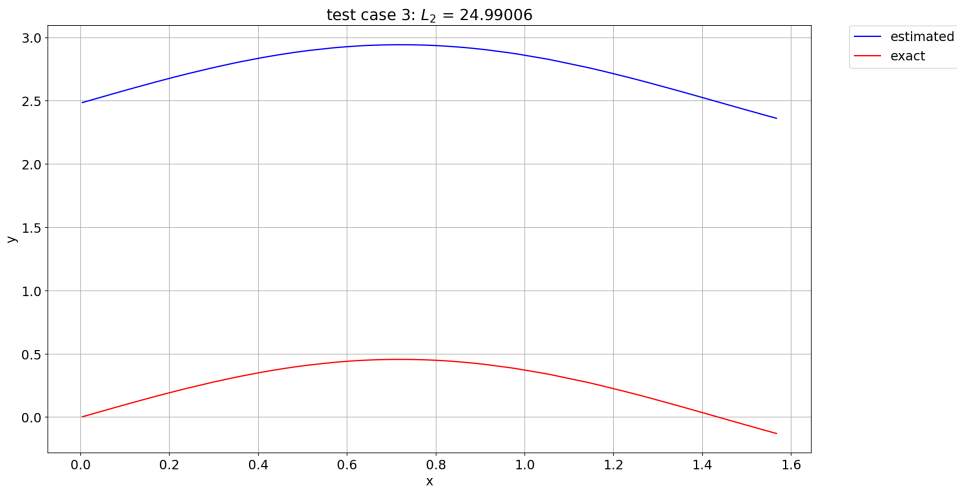
Physics Informed Deep-o-net - results



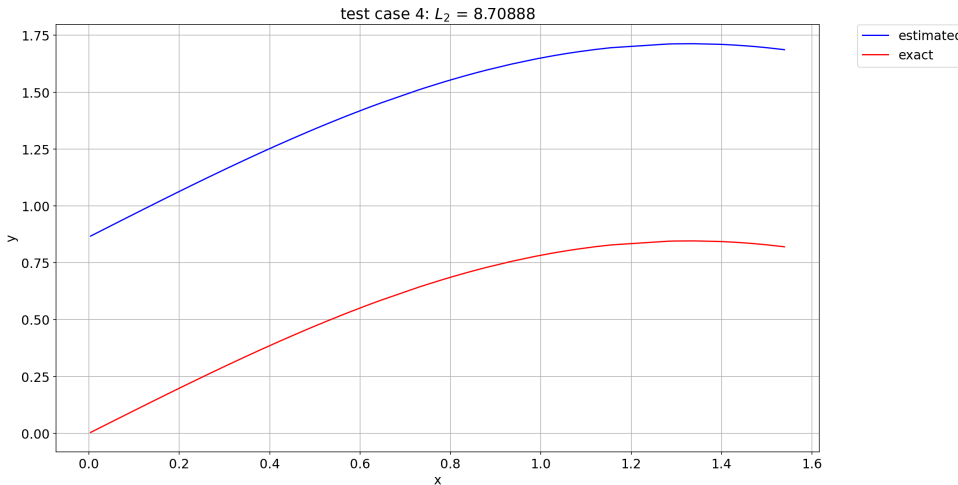
Physics Informed Deep-o-net - results



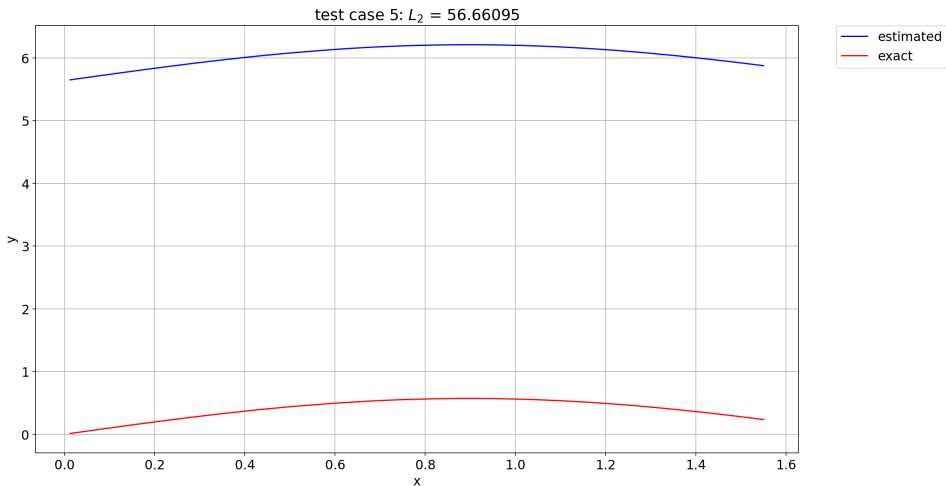
Physics Informed Deep-o-net - results



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Physics Informed Deep-o-net - Results not mached

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Integral Operator

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Integral Operator

$$x = \int \cos(\omega t)$$

Physics Informed Deep-o-net - Results not mached

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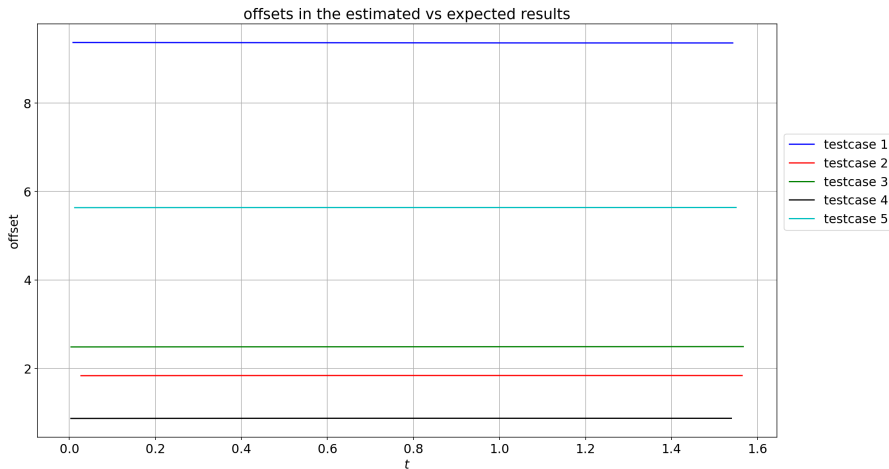
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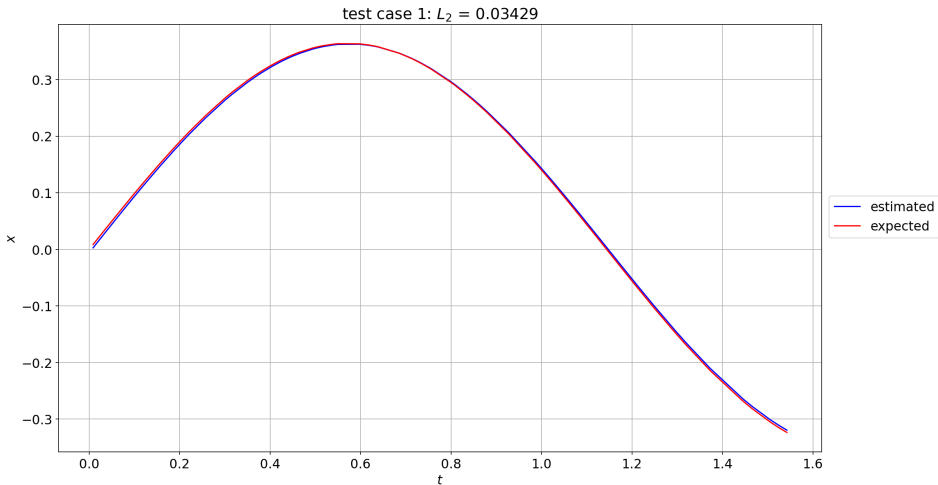
Integral Operator

$$x = \int \cos(\omega t) + C$$

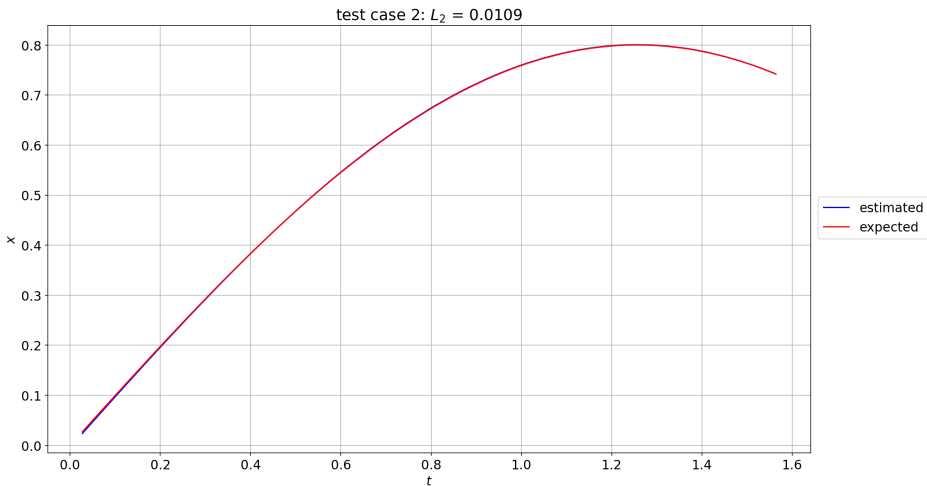
Physics Informed Deep-o-net - integral constant



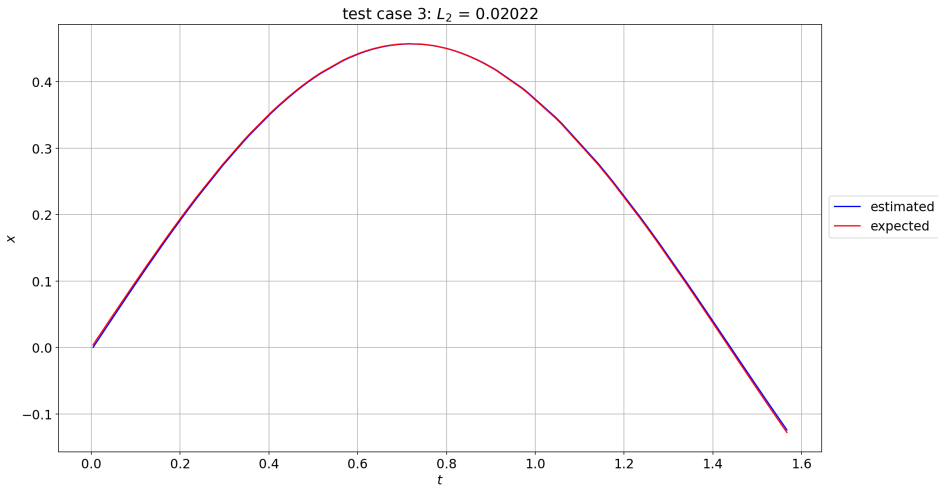
Physics Informed Deep-o-net - corrected results



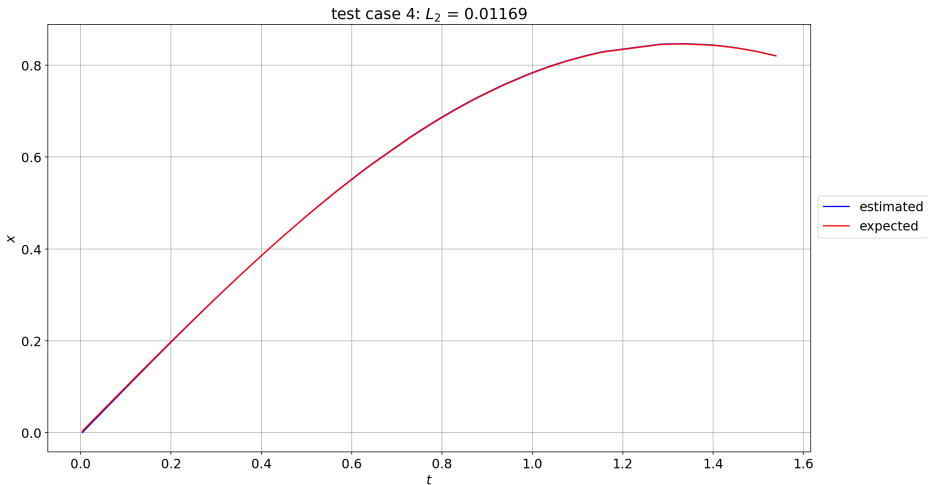
Physics Informed Deep-o-net - corrected results



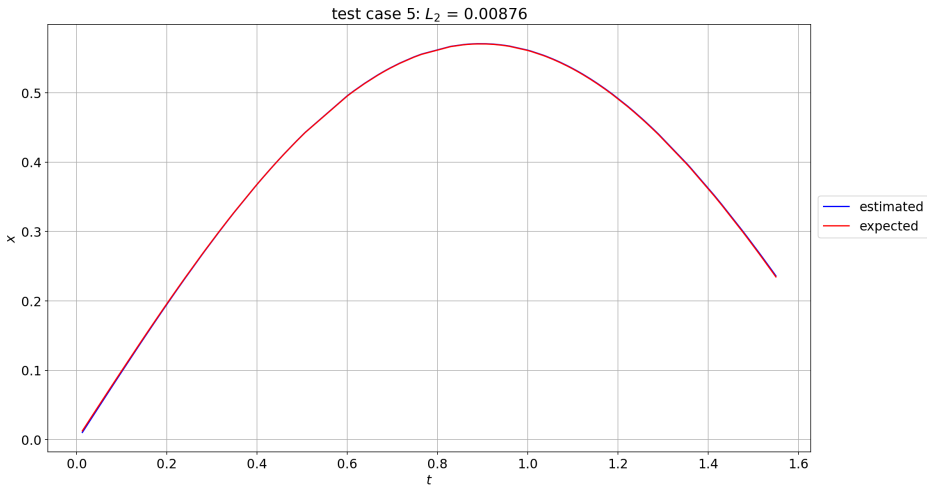
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Source code

Available on github

<https://github.com>