

Deep Learning for Data Science

DS 542

<https://dl4ds.github.io/fa2025/>

Measuring Performance and Generalization



Plan for Today

- MNIST1D dataset model and performance
- Noise, bias, and variance
- Reducing variance
- Reducing bias & bias-variance trade-off
- Double descent
- Choosing hyperparameters

MNIST1D

Scaling down Deep Learning

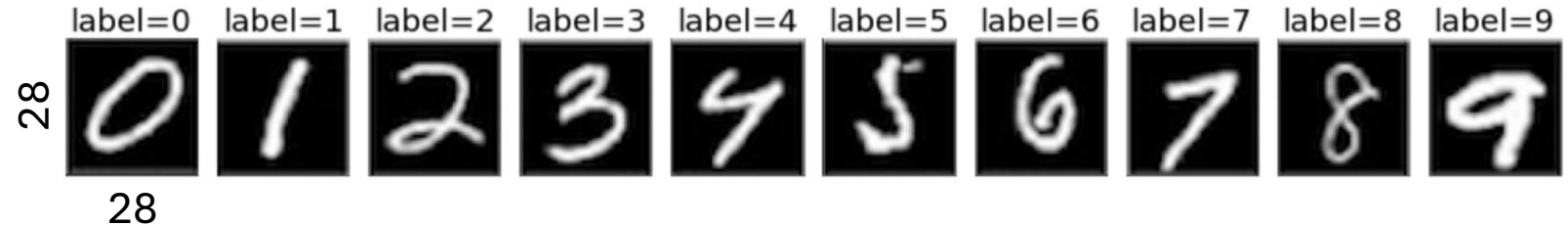
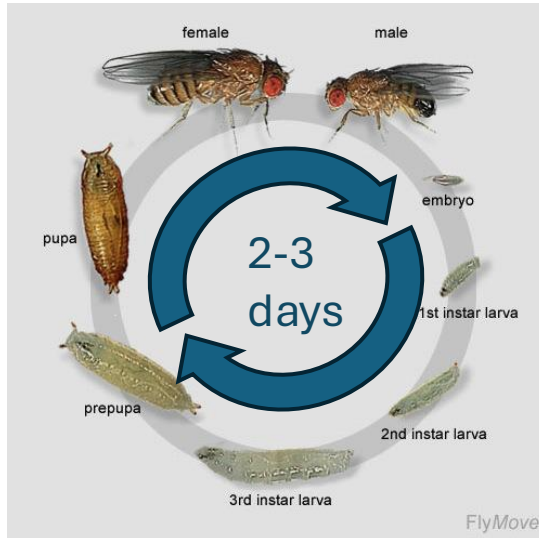
Sam Greydanus¹

“A large number of deep learning innovations including [dropout](#), [Adam](#), [convolutional networks](#), [generative adversarial networks](#), and [variational autoencoders](#) began life as MNIST experiments. Once these innovations proved themselves on small-scale experiments, scientists found ways to scale them to larger and more impactful applications.”

S. Greydanus, “Scaling down Deep Learning.” arXiv, Dec. 04, 2020. doi:
[10.48550/arXiv.2011.14439](https://arxiv.org/abs/10.48550/arXiv.2011.14439).
<https://github.com/greydanus/mnist1d>

MNIST Dataset

- 28x28x1 grayscale images
- 60K Training, 10K Test
- “Is to Deep Learning what fruit flies are to genetics research”

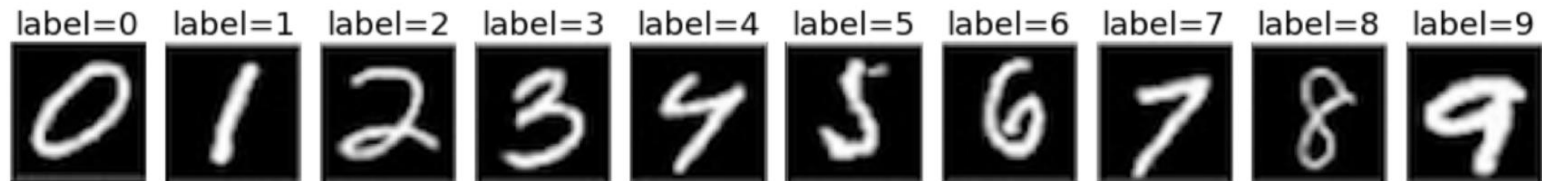


But poorly differentiates model performance:

Model Type	Accuracy
Logistic Regression	94%
MLP	99+%
CNN	99+%

MNIST 1D Dataset

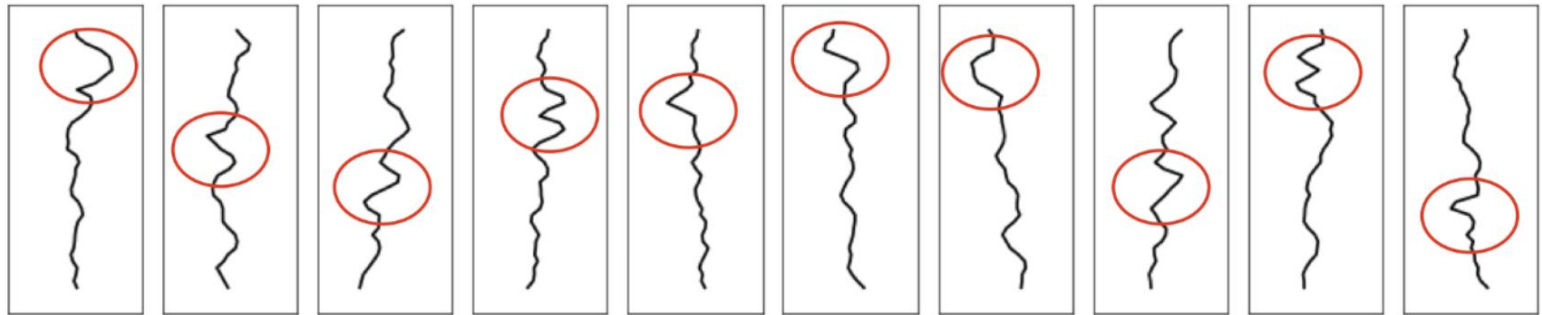
Original MNIST examples



Represent digits as 1D patterns



Pad, translate & transform



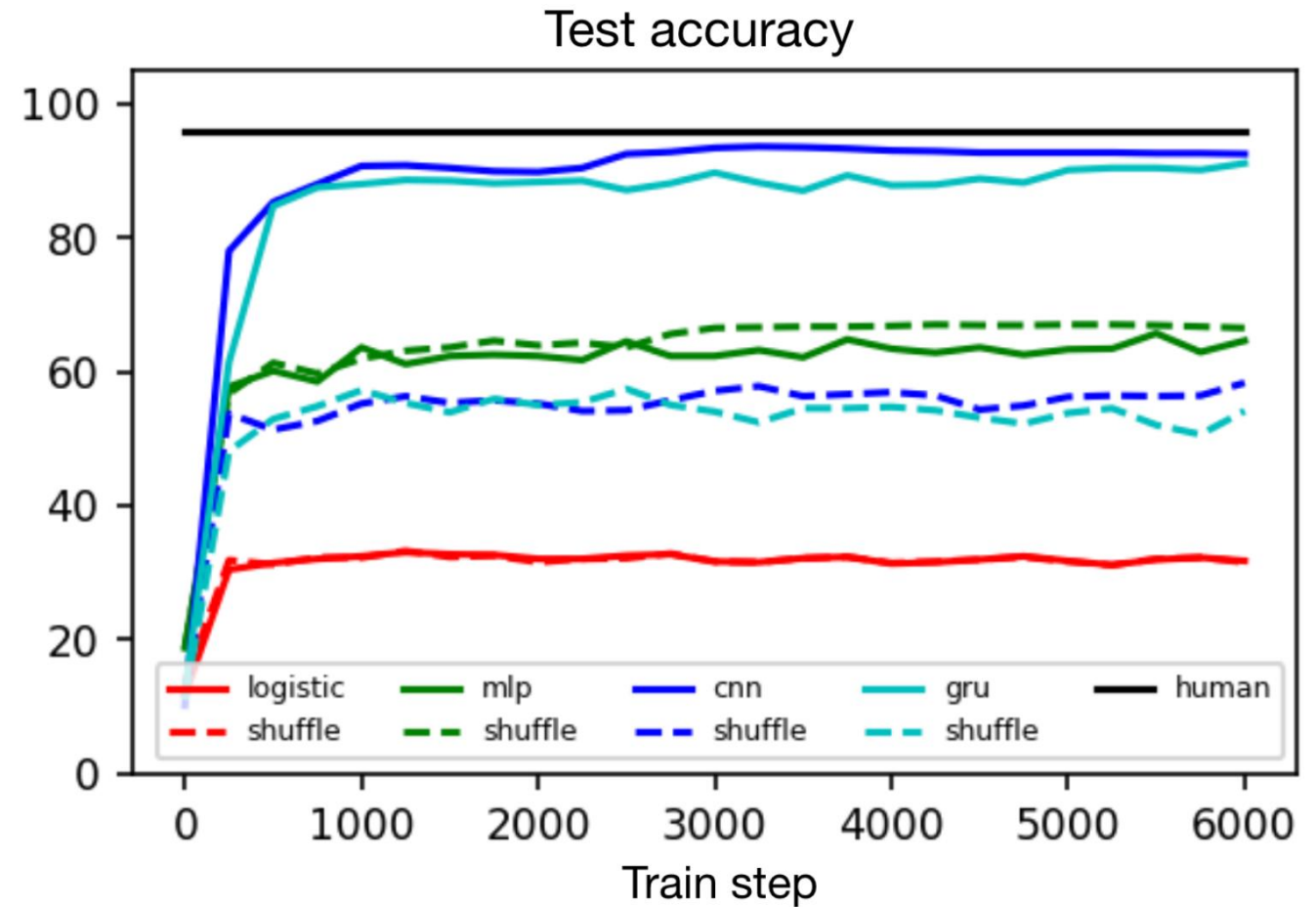
Fixed, 1-D, length-12
templates for each of 10
digit classes

Generate dataset by
programmatically applying
6 parametric
transformations.

E.g. pad, shear, translate, correlated noise, i.i.d. noise, interpolation.

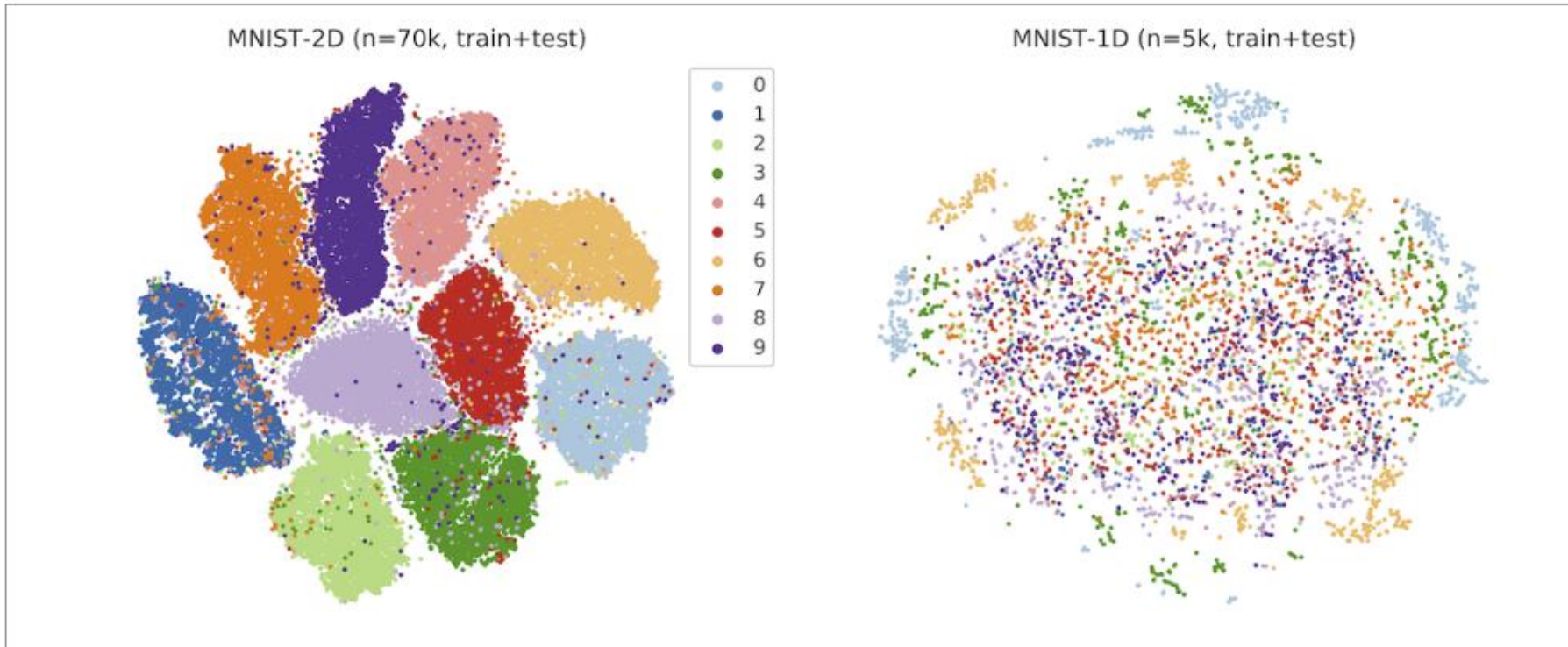
MNIST 1D

Differentiates performance of different model types much more than MNIST



Dataset	Logistic regression	Fully connected model	Convolutional model	GRU model	Human expert
MNIST	94 ± 0.5	> 99	> 99	> 99	> 99
MNIST-1D	32 ± 1	68 ± 2	94 ± 2	91 ± 2	96 ± 1
MNIST-1D (shuffled)	32 ± 1	68 ± 2	56 ± 2	57 ± 2	$\approx 30 \pm 10$

Visualizing MNIST and MNIST-1D with tSNE



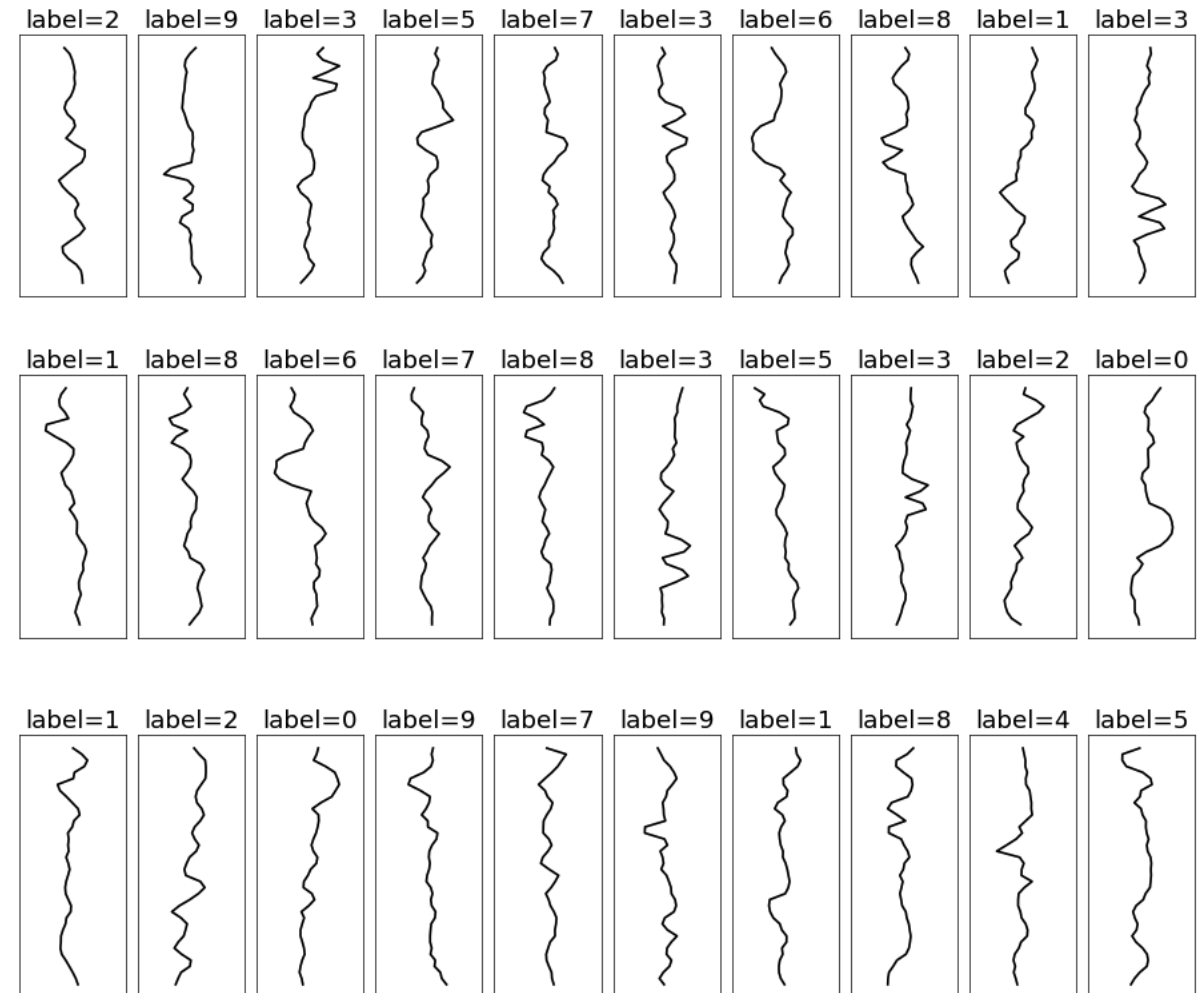
Visualizing the MNIST and MNIST-1D datasets with tSNE. The well-defined clusters in the MNIST plot indicate that the majority of the examples are separable via a kNN classifier in pixel space. The MNIST-1D plot, meanwhile, reveals a lack of well-defined clusters which suggests that learning a nonlinear representation of the data is much more important to achieve successful classification. Thanks to [Dmitry Kobak](#) for making this plot.

<https://twitter.com/hippopedoid>

MNIST1D Train and Test Set

- 1D, Length 40 samples
- 4,000 training samples
- 1,000 test samples (80/20 split)

Dataset Samples



Network

- 40 inputs
- 10 outputs
- Two hidden layers
 - 100 hidden units each
- SGD with batch size 100, learning rate 0.1
- 6000 steps (?? Epochs)

```
model = torch.nn.Sequential(  
    torch.nn.Linear(40, 100),  
    torch.nn.ReLU(),  
    torch.nn.Linear(100, 100),  
    torch.nn.ReLU(),  
    torch.nn.Linear(100, 10))
```

```
# choose cross entropy loss function  
loss_function = torch.nn.CrossEntropyLoss()  
  
# construct SGD optimizer and initialize learning rate and momentum  
optimizer = torch.optim.SGD(model.parameters(), lr = 0.1)  
  
# object that decreases learning rate by half every 10 epochs  
scheduler = StepLR(optimizer, step_size=10, gamma=0.5)  
  
# load the data into a class that creates the batches  
data_loader = DataLoader(TensorDataset(x_train,y_train), batch_size=100, shuffle=True)
```

• • •

```
# inference – just choose the max  
pred_train = model(x_train)  
pred_test = model(x_test)  
_, predicted_train_class = torch.max(pred_train.data, 1)  
_, predicted_test_class = torch.max(pred_test.data, 1)
```

```
=====
```

Layer (type:depth-idx)	Output Shape	Param #
Sequential	[1, 10]	–
└─Linear: 1-1	[1, 100]	4,100
└─ReLU: 1-2	[1, 100]	–
└─Linear: 1-3	[1, 100]	10,100
└─ReLU: 1-4	[1, 100]	–
└─Linear: 1-5	[1, 10]	1,010

```
=====
```

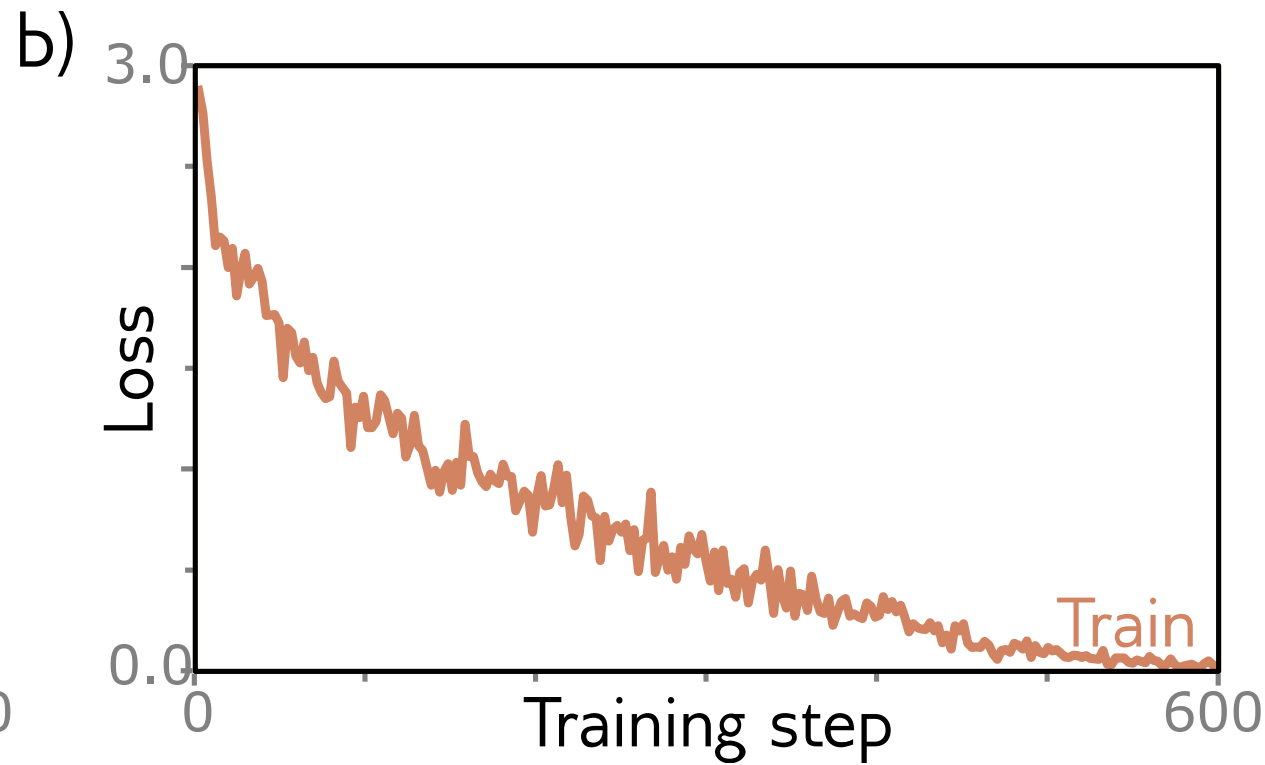
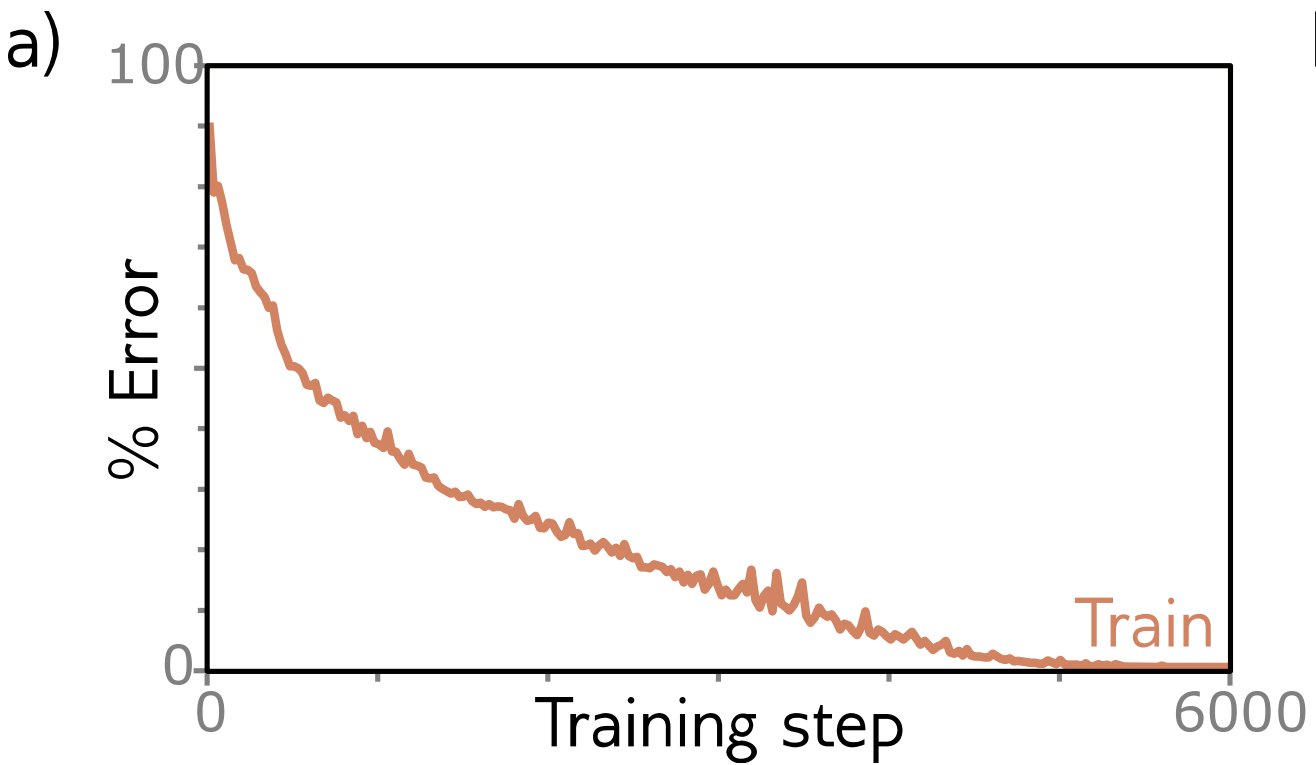
Total params: 15,210
Trainable params: 15,210
Non-trainable params: 0
Total mult-adds (Units.MEGABYTES): 0.02

```
=====
```

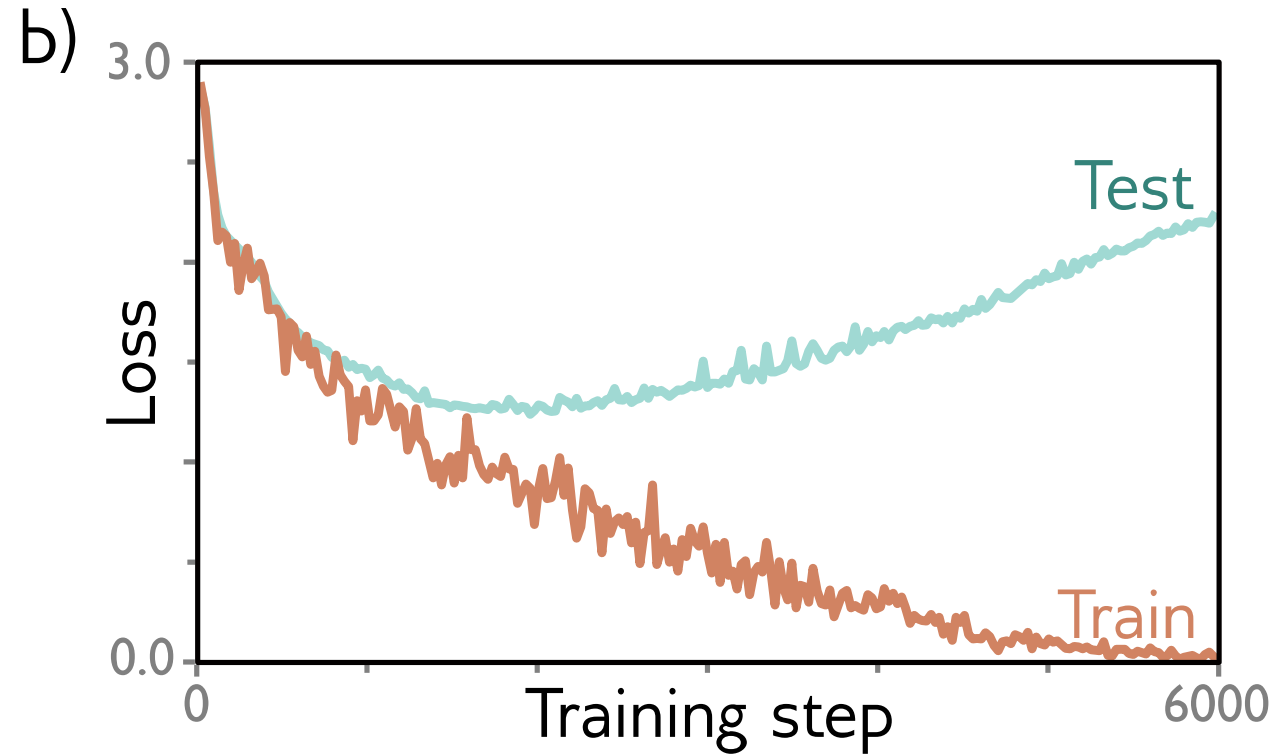
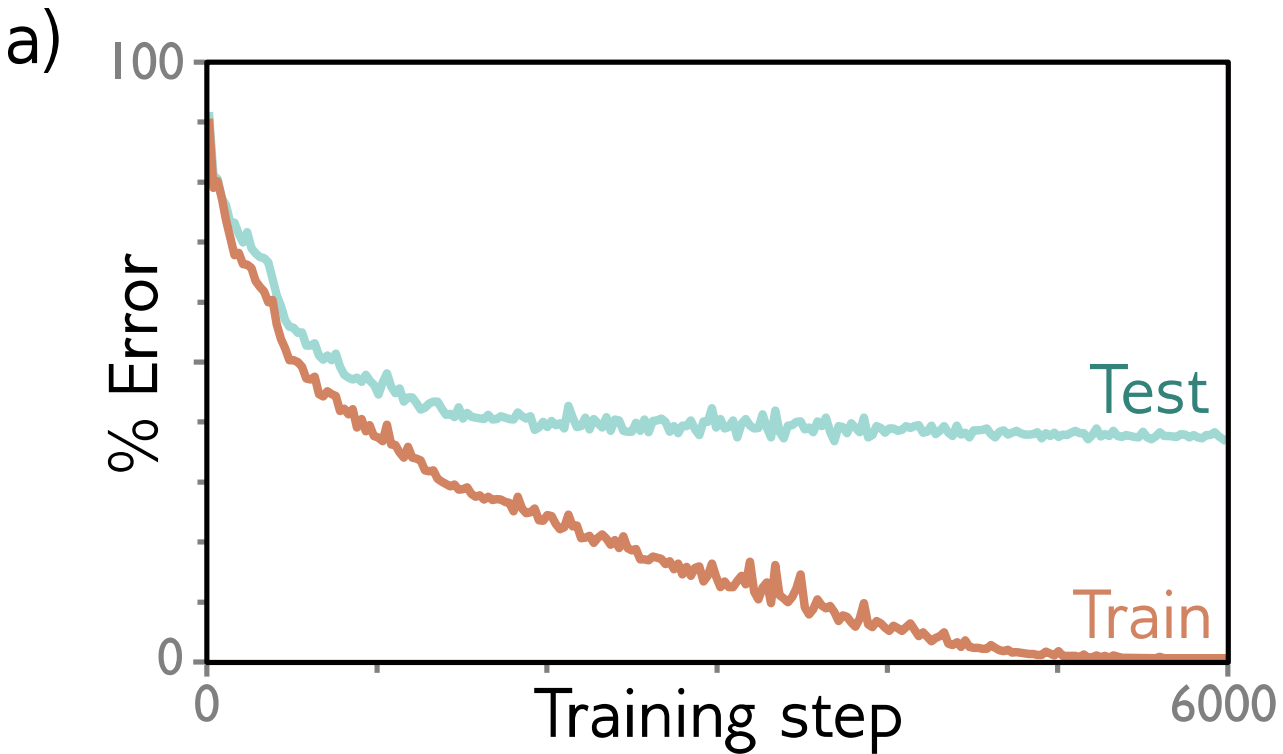
Input size (MB): 0.00
Forward/backward pass size (MB): 0.00
Params size (MB): 0.06
Estimated Total Size (MB): 0.06

```
=====
```

Results



Need to use separate test data



The model has not **generalized** well to the new data

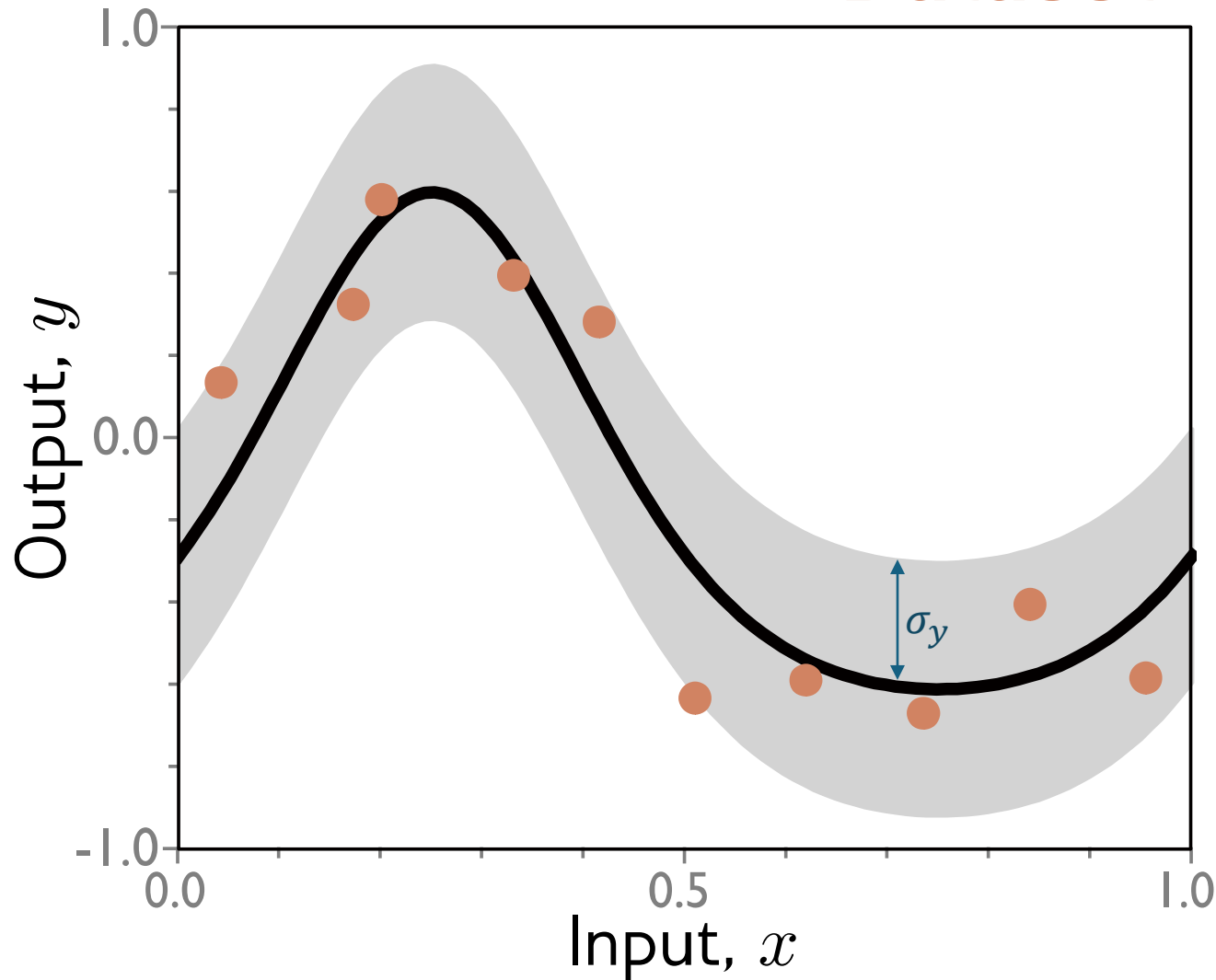
Any Questions?



- MNIST1D dataset model and performance
- Noise, bias, and variance
- Reducing variance
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Regression example with Toy Model

Dataset



"True" function:

$$y = e^{\sin(2\pi x)}$$

Add small uniform noise to x:

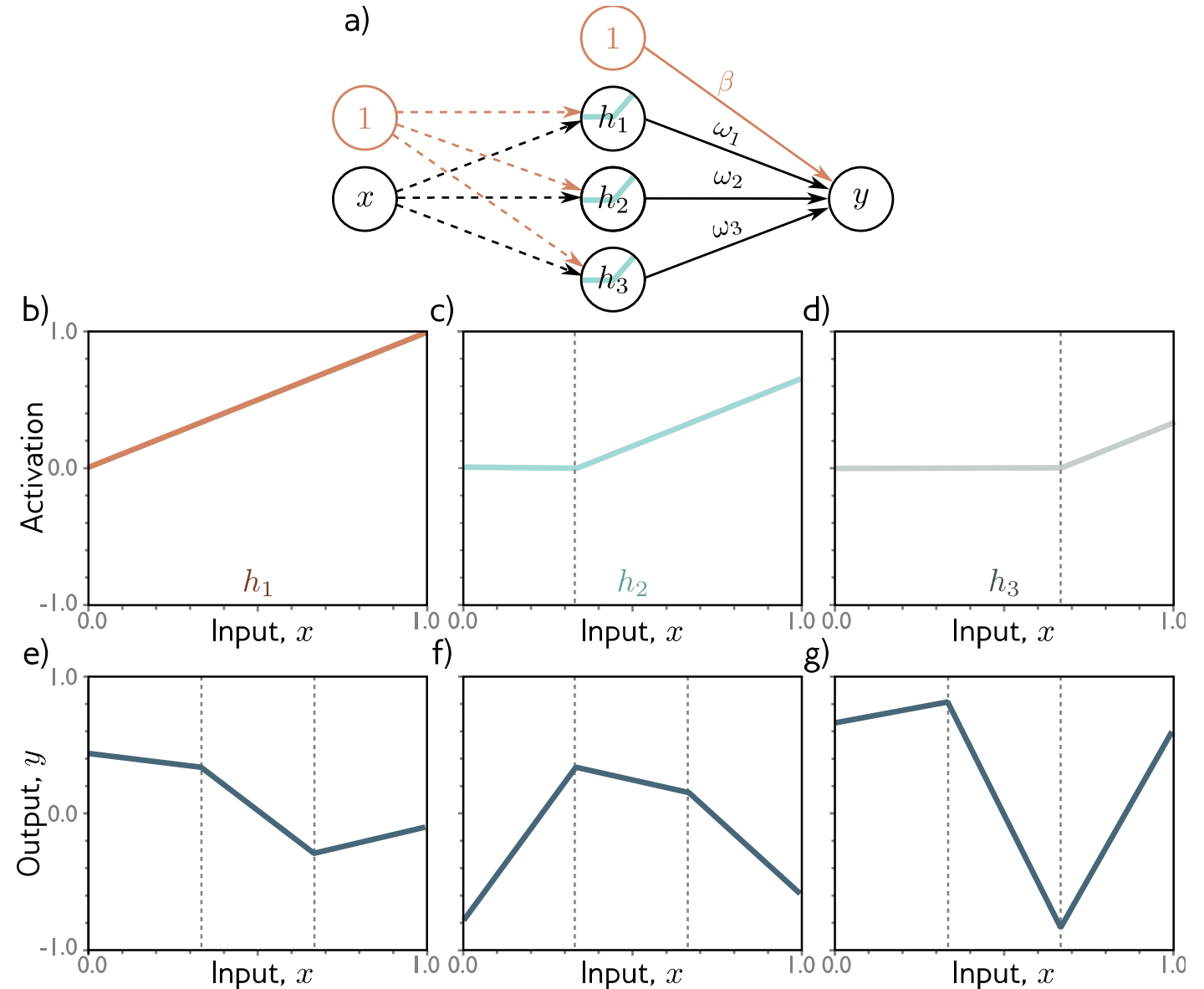
$$x = x + \mathcal{U}(\pm 1/\text{num_data})$$

Add small Gaussian noise to y:

$$y = y + \mathcal{N}(0, \sigma_y)$$

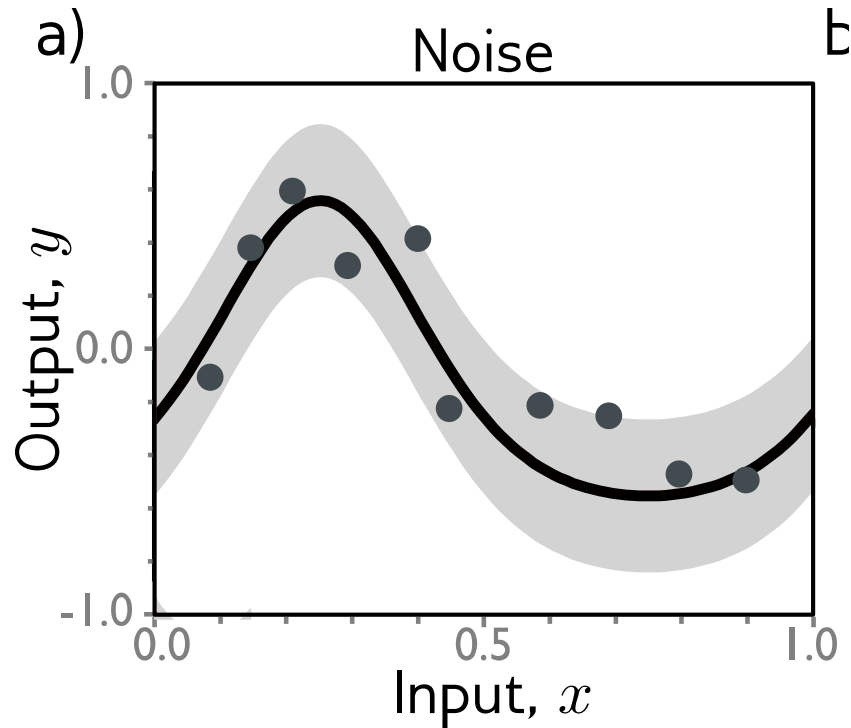
Toy model

- D hidden units
- First layer fixed so “joints” divide interval evenly, e.g. $0, 1/D, 2/D, \dots, (D-1)/D$
- Second layer trained
- But... now linear in \mathbf{h}
 - so convex cost function
 - can find best soln in closed-form
- A piecewise linear model with D regions.

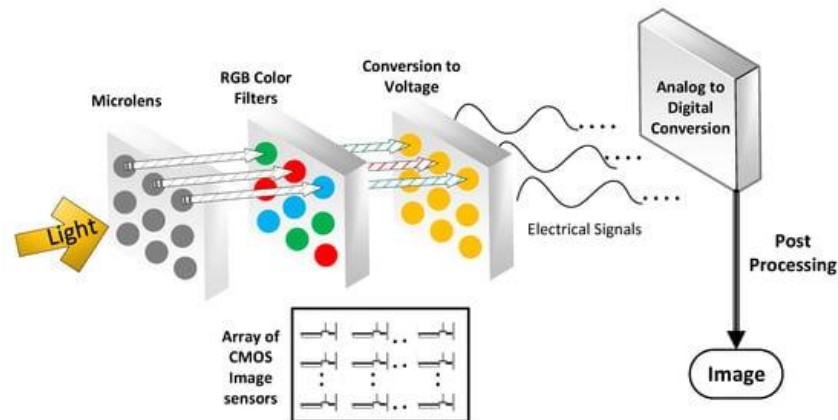


Three possible sources of error:
noise, *bias* and *variance*

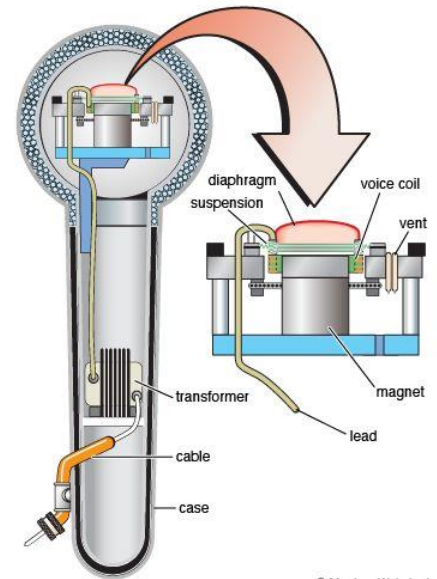
Noise, bias, and variance



- Genuine stochastic nature of the underlying model
- Noise in measurements, e.g. from sensors
- Some variables not observed
- Data mislabeled



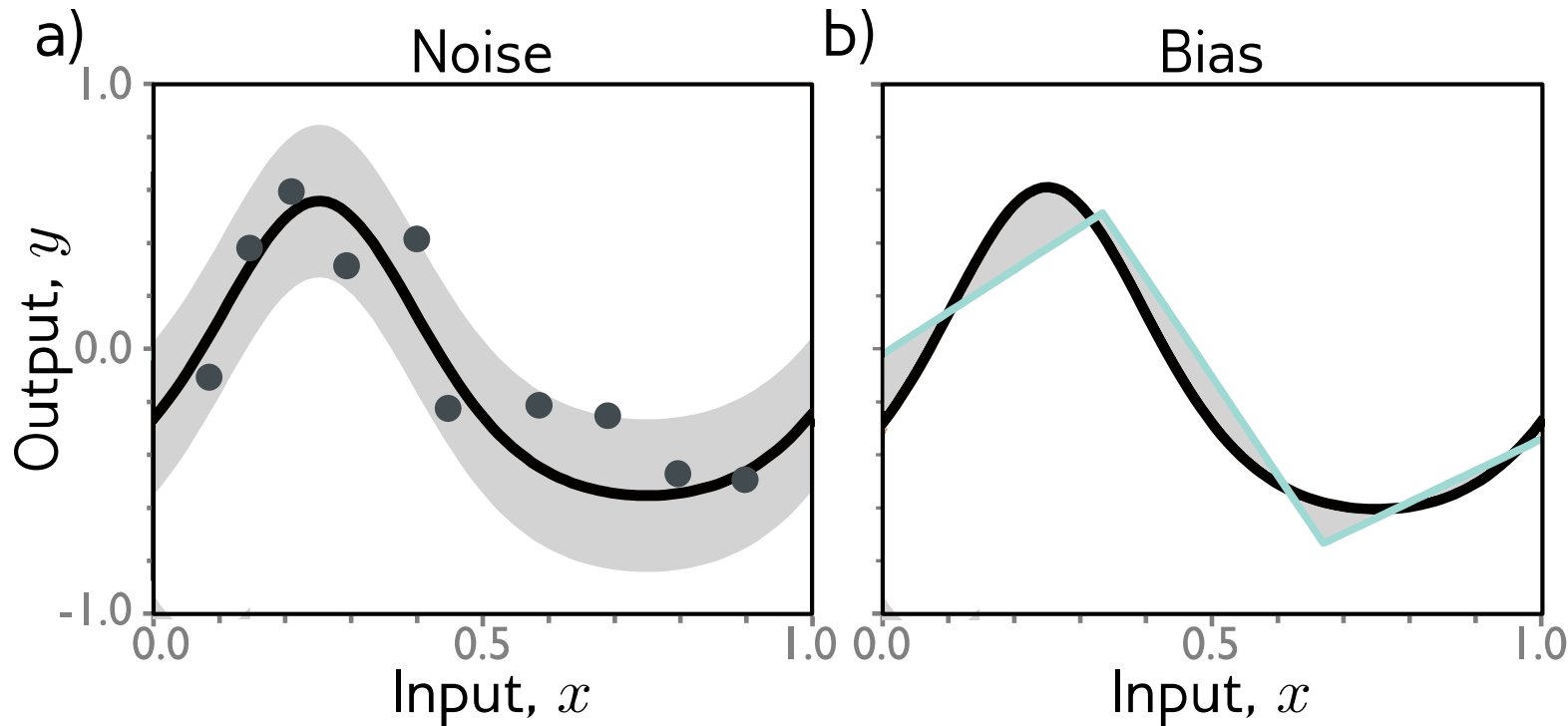
<https://images.app.goo.gl/2PuBhaFpfdL9Pyjb8>



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<https://images.app.goo.gl/CMDaXSCdX4pqN8Yx7>

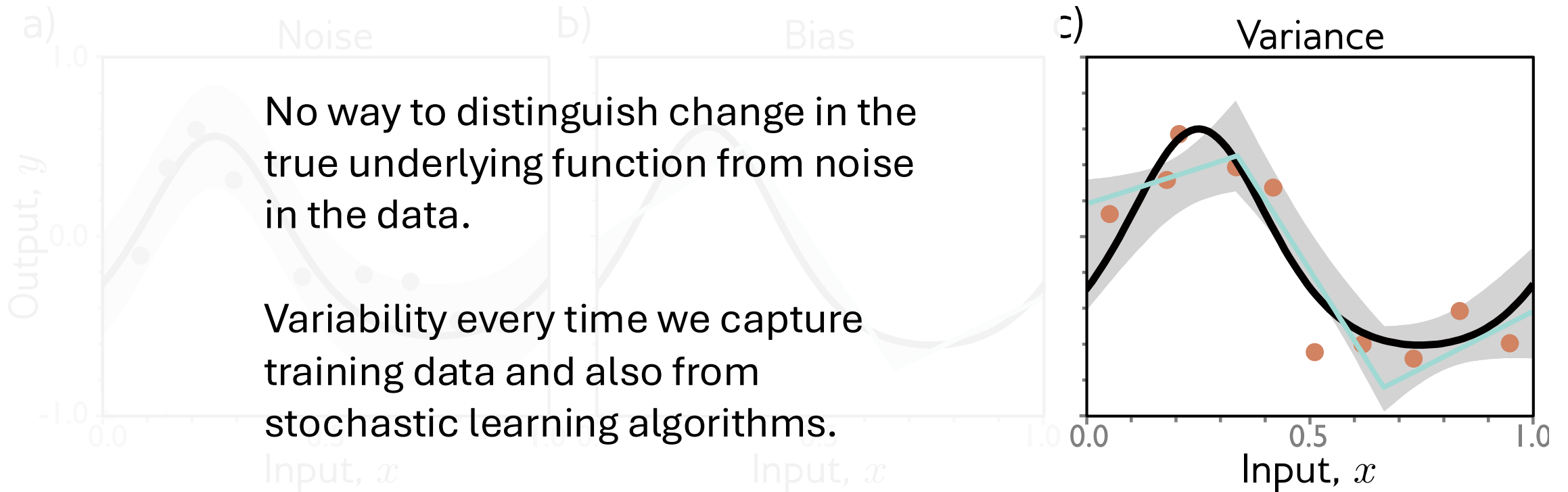
Noise, **bias**, and variance



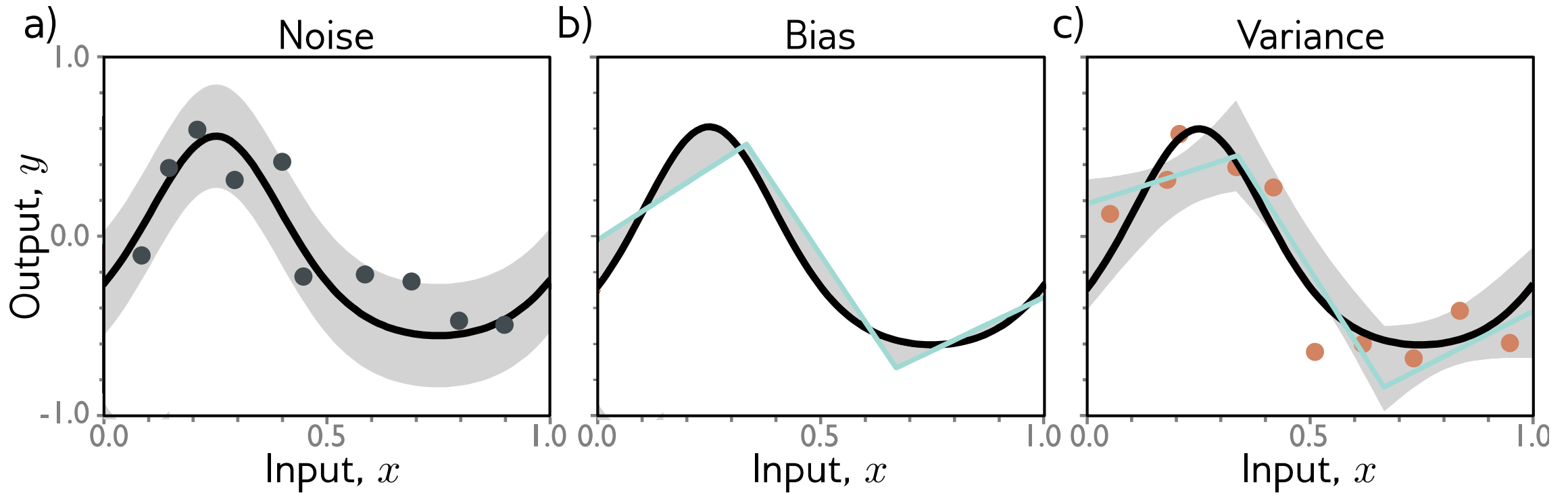
Bias occurs because the model lacks precision or capacity to accurately match the underlying function.

E.g. optimal fit with 3 hidden units and 3 line segments

Noise, bias, and variance



Noise, bias, and variance



Least squares regression only

$$L[x] = (f[x, \phi] - y[x])^2$$

- We can show that:

$$\mathbb{E}_y[L[x]] = (f[x, \phi] - \mu[x])^2 + \sigma^2$$

- And then:

$$\mathbb{E}_{\mathcal{D}}[\mathbb{E}_y[L[x]]] = \underbrace{\mathbb{E}_{\mathcal{D}}[(f[x, \phi[\mathcal{D}]] - f_{\mu}[x])^2]}_{\text{variance}} + \underbrace{(f_{\mu}[x] - \mu[x])^2}_{\text{bias}} + \underbrace{\sigma^2}_{\text{noise}}$$

Expectation over noise
in training data

Expectation over
noise in test data

Actual model

Best possible model if
we had infinite data

True function

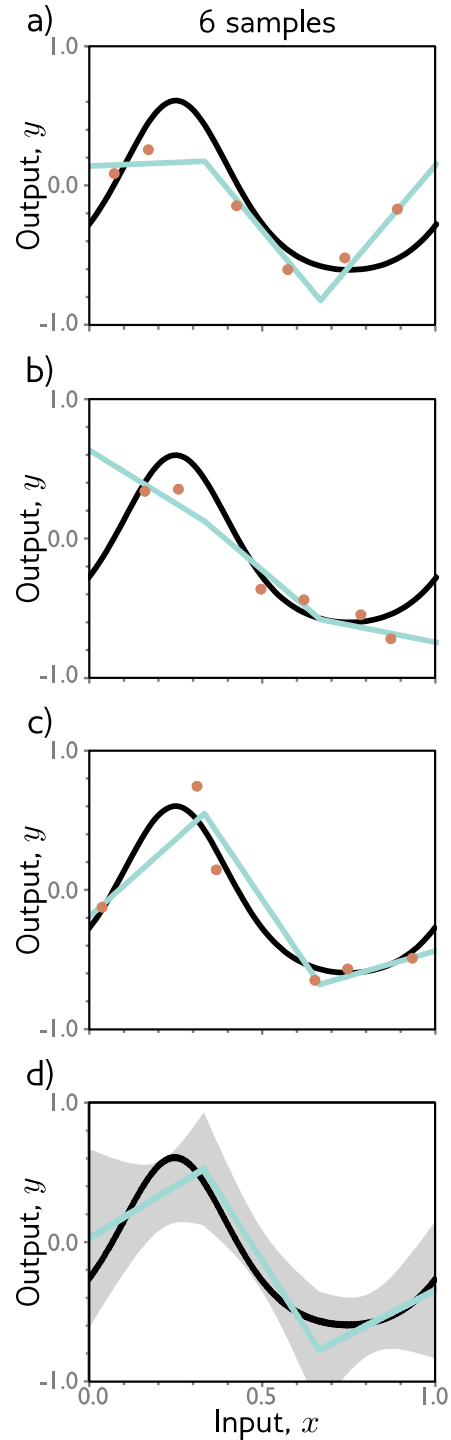
More complex interactions between noise, bias and variance in more complex models.

Any Questions?



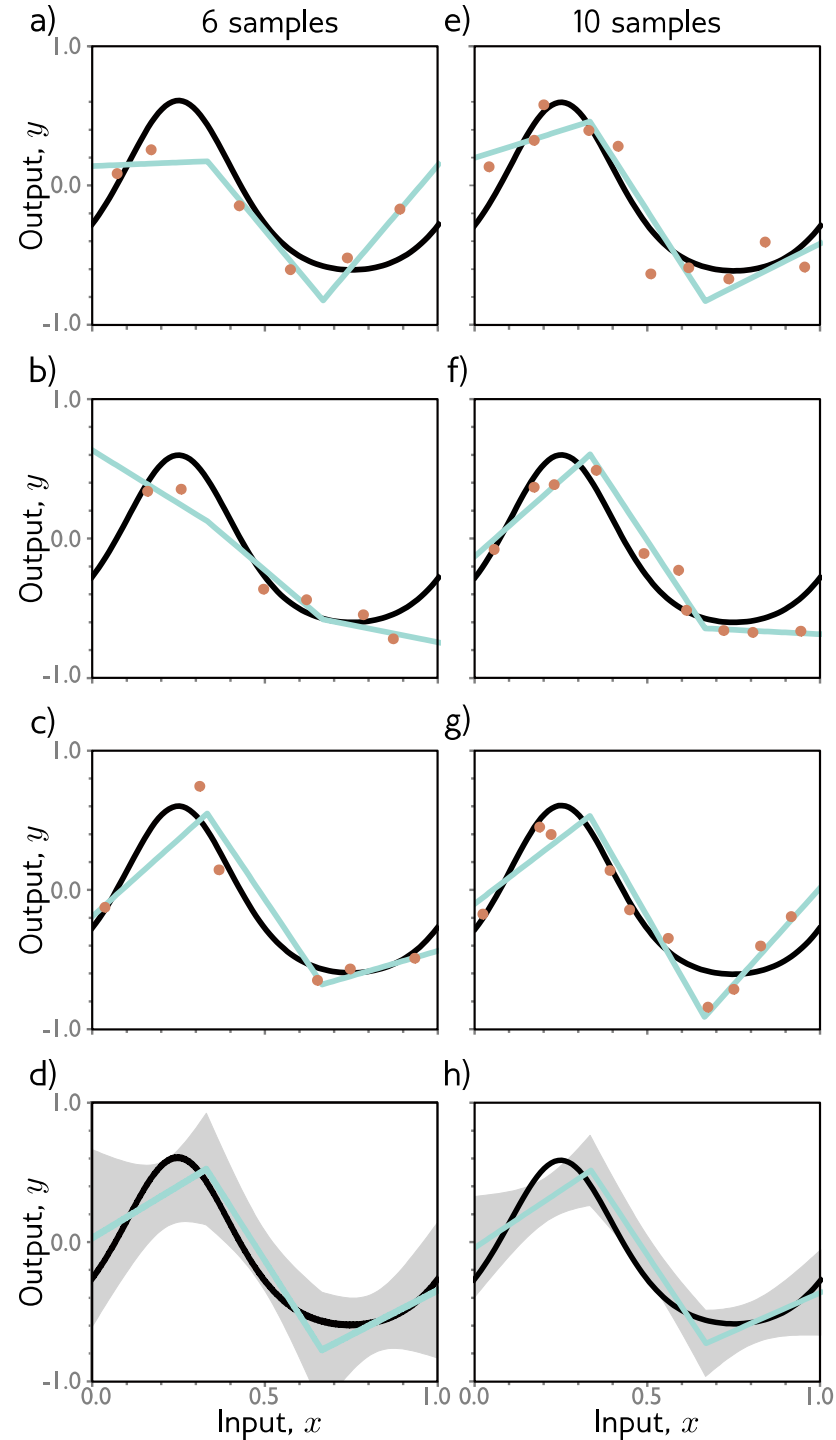
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Variance



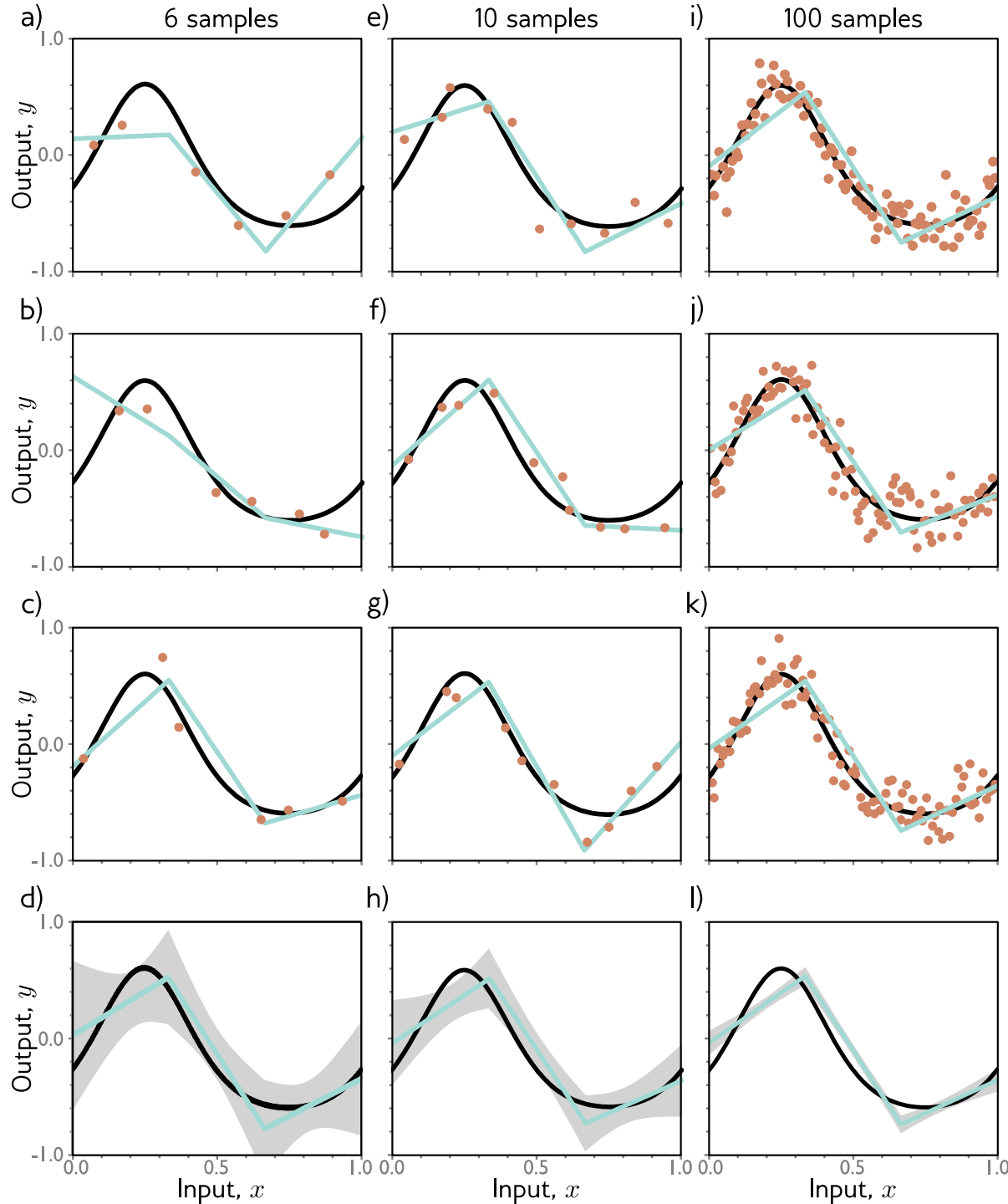
When measuring (capturing) 6 different data samples with a fixed model (e.g. 3 hidden units), we get different optimal fits every time.

Variance



Can reduce
variance by
adding more
samples

Variance



Can reduce
variance by
adding more
samples

Any Questions?

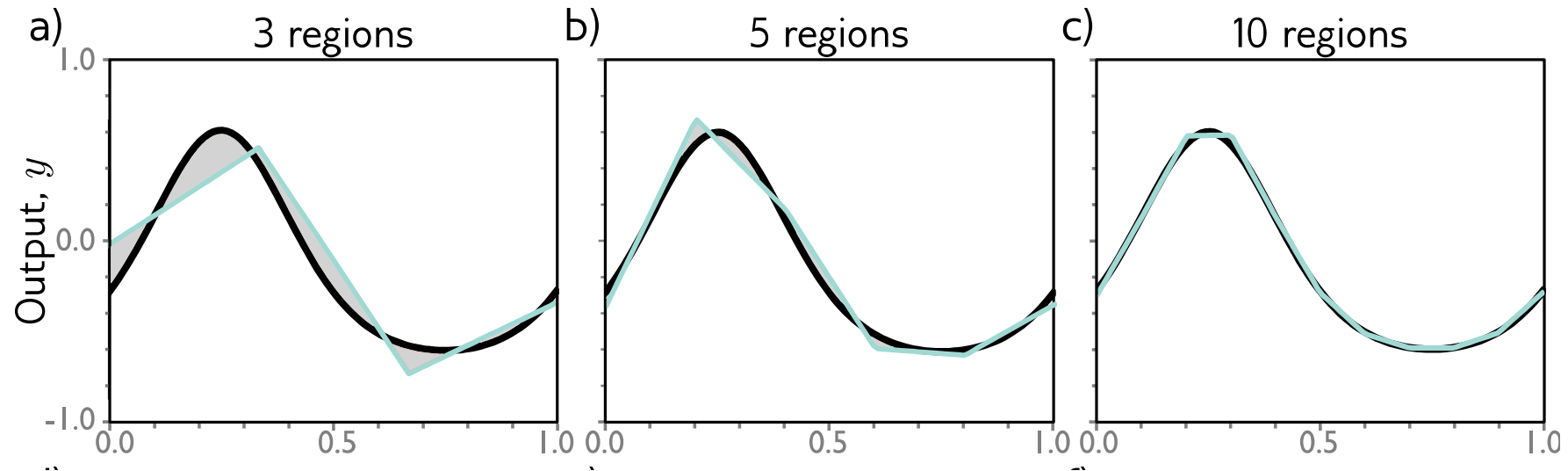


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Reducing bias

(example with the true function)

Bias

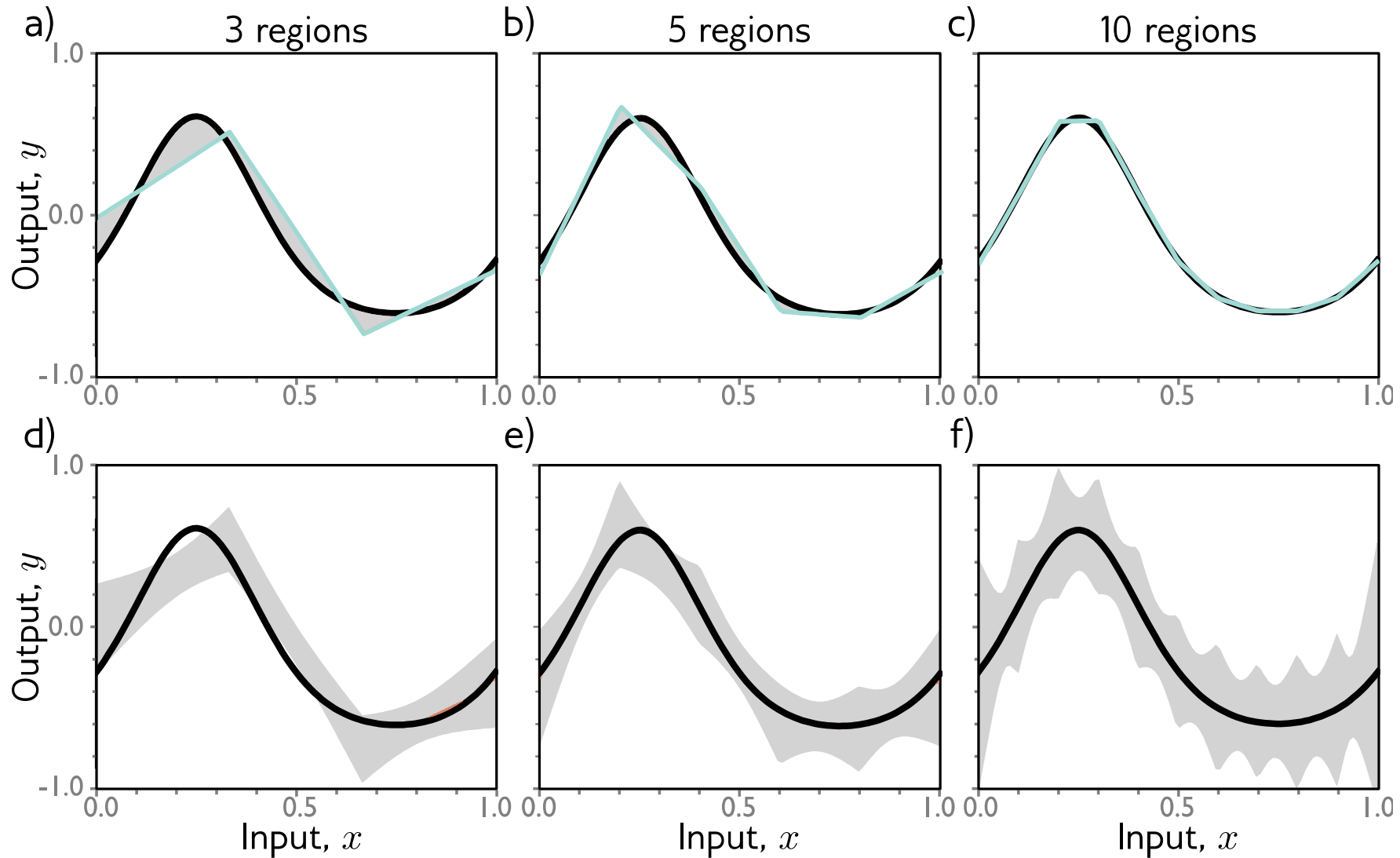


We can reduce bias by adding more model capacity.

In this case, adding more hidden units.

Reducing bias \rightarrow Increases variance!!

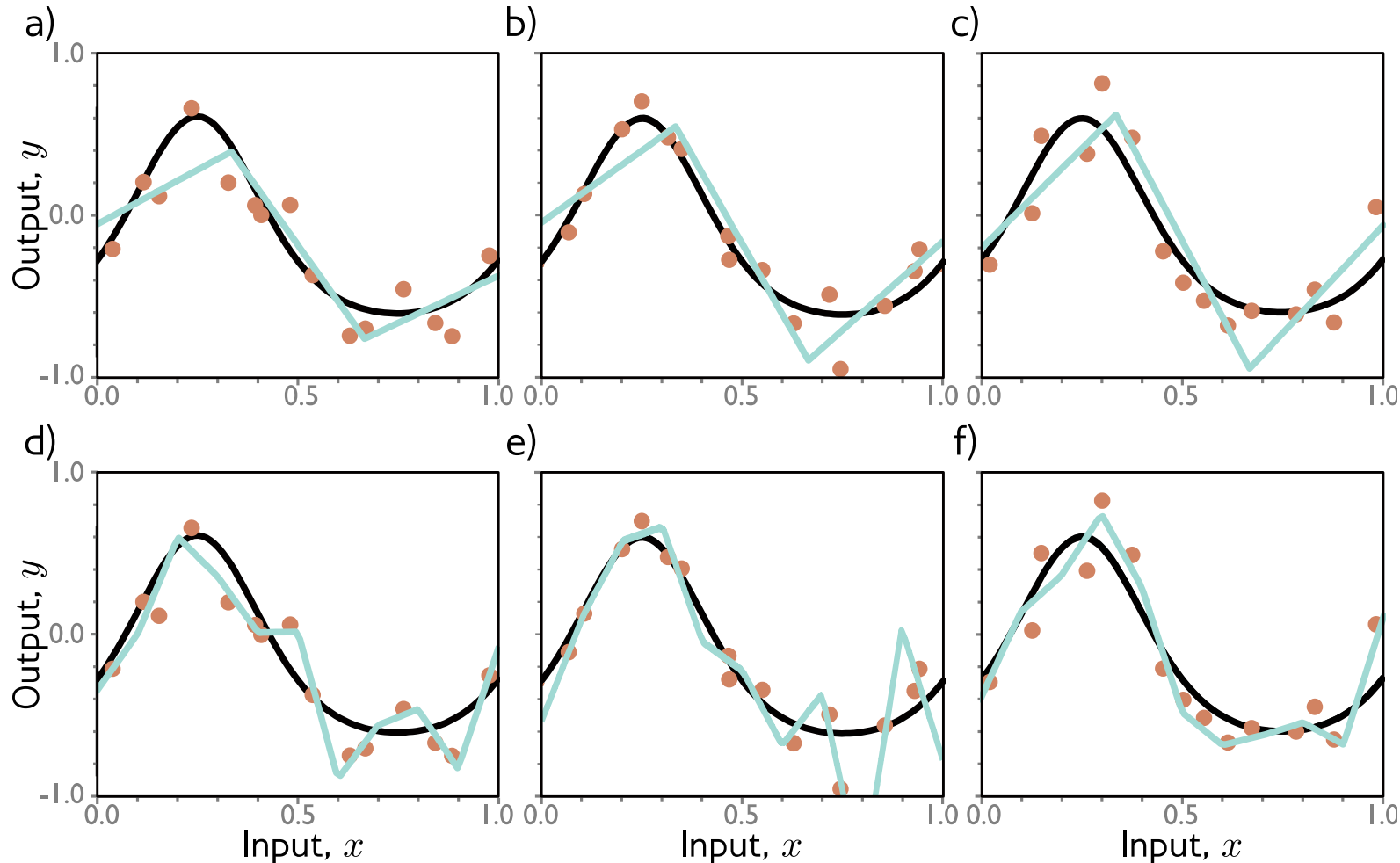
Bias



Variance

Why does variance increase? Overfitting

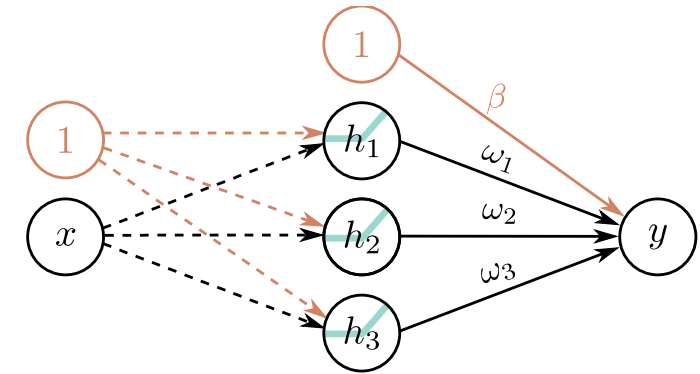
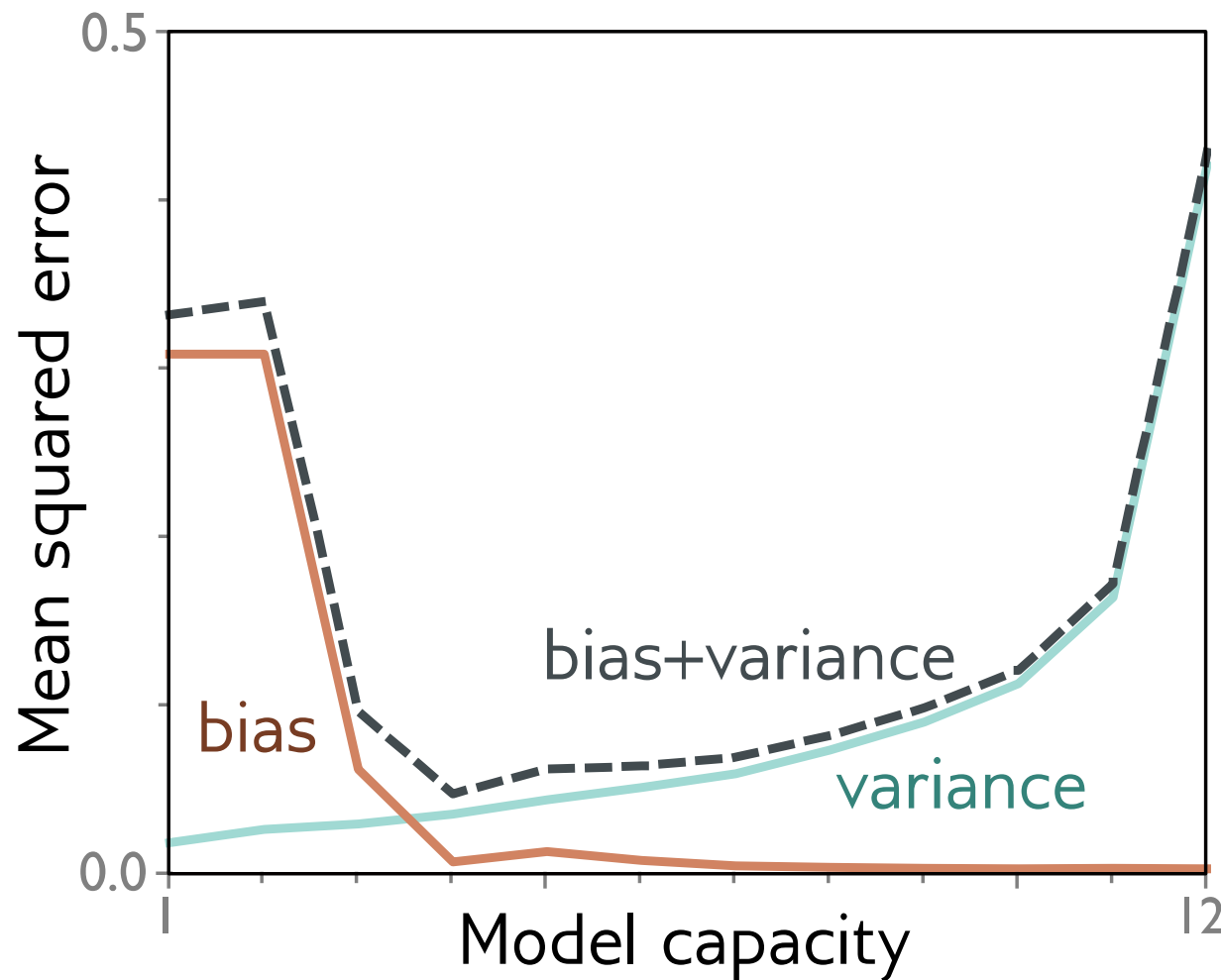
3 Regions



10 Regions

Describes the training data better, but not the true underlying function (black curve)
Many ways to fit a sample of 15 data points

Bias and variance trade-off for the simple linear model



$$\mathbb{E}_{\mathcal{D}} \left[\mathbb{E}_y [L[x]] \right] = \underbrace{\mathbb{E}_{\mathcal{D}} \left[(f[x, \phi[\mathcal{D}]] - f_{\mu}[x])^2 \right]}_{\text{variance}} + \underbrace{(f_{\mu}[x] - \mu[x])^2}_{\text{bias}} + \underbrace{\sigma^2}_{\text{noise}}$$

Number of hidden units

But does picking model capacity to minimize bias & variance hold for more complex data and models?

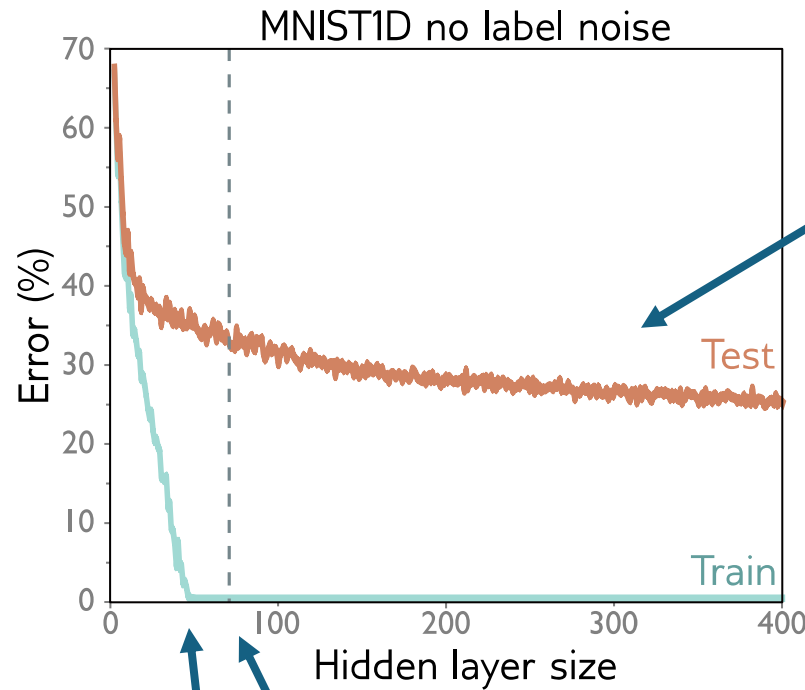
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Train and Test Error versus # of Hidden Layers

- 10,000 training examples
- 5,000 test examples
- Two hidden layers
- Adam optimizer
- Step size of 0.005
- Full batch
- 4000 training steps

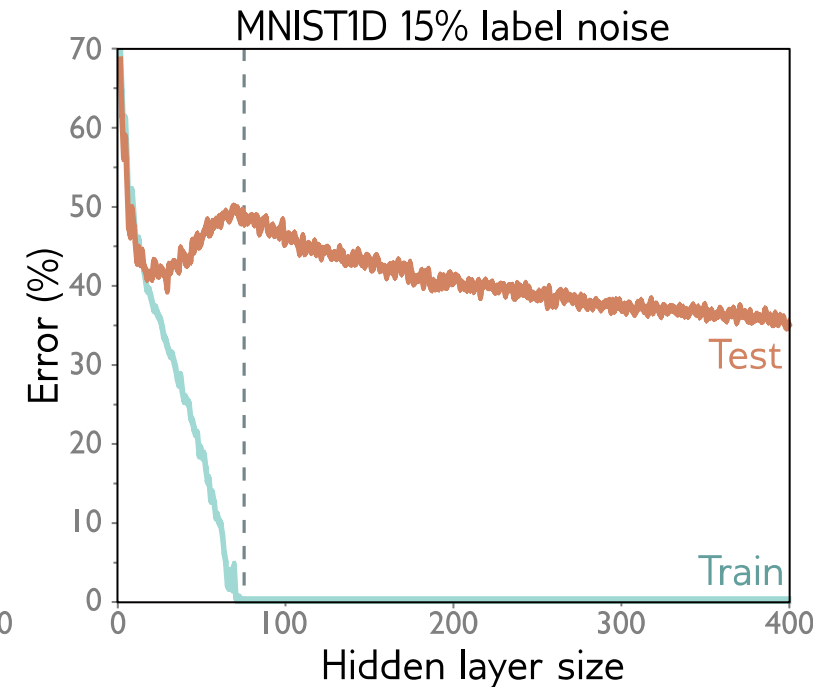
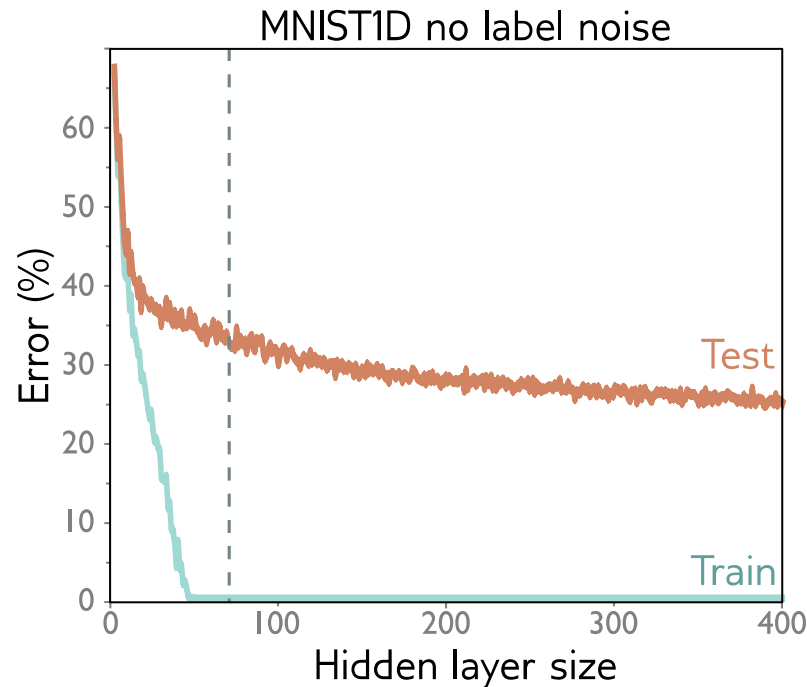


Test error keep decreasing even as we keep increasing model capacity!

Training parameters = Training examples

Model has *memorized* the training set
Why do we say that?

Now randomize
15% of the
training labels

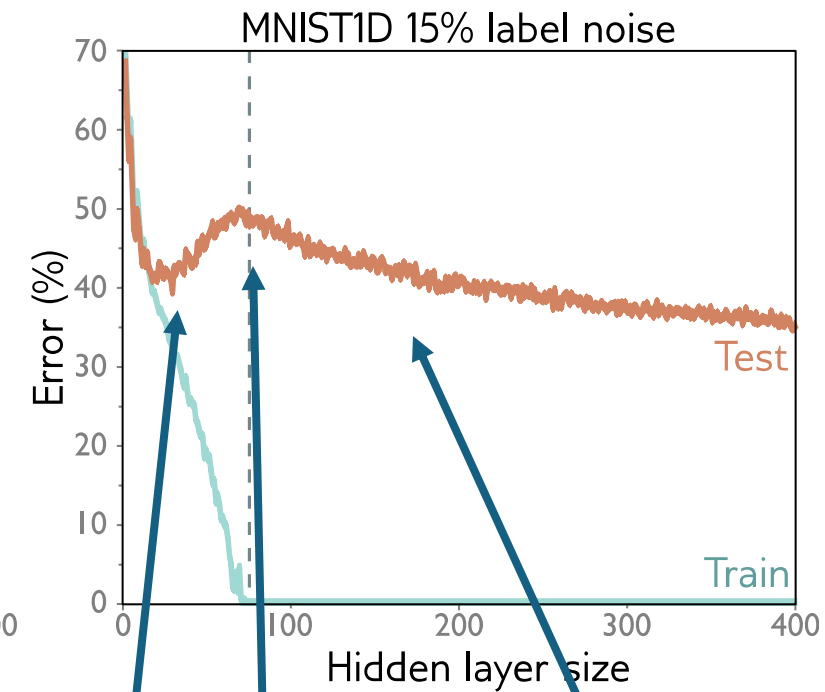
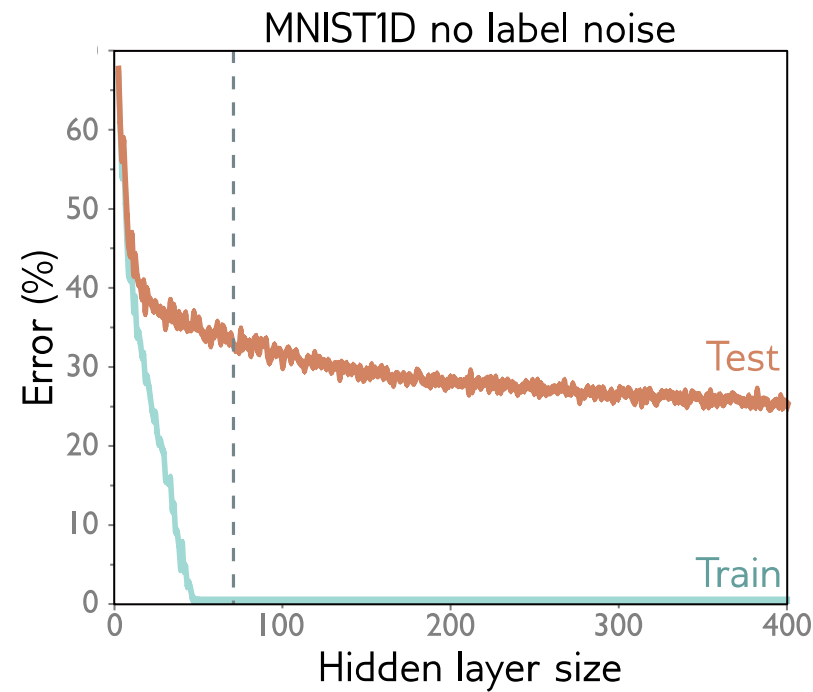


Now we see what looks like bias-
variance trade-off as we increase
capacity to the point where the model
fits training data.

Reminder: vertical dashed line is where:
training parameters = # training samples

But then???

Double Descent



Classical or under-parameterized regime

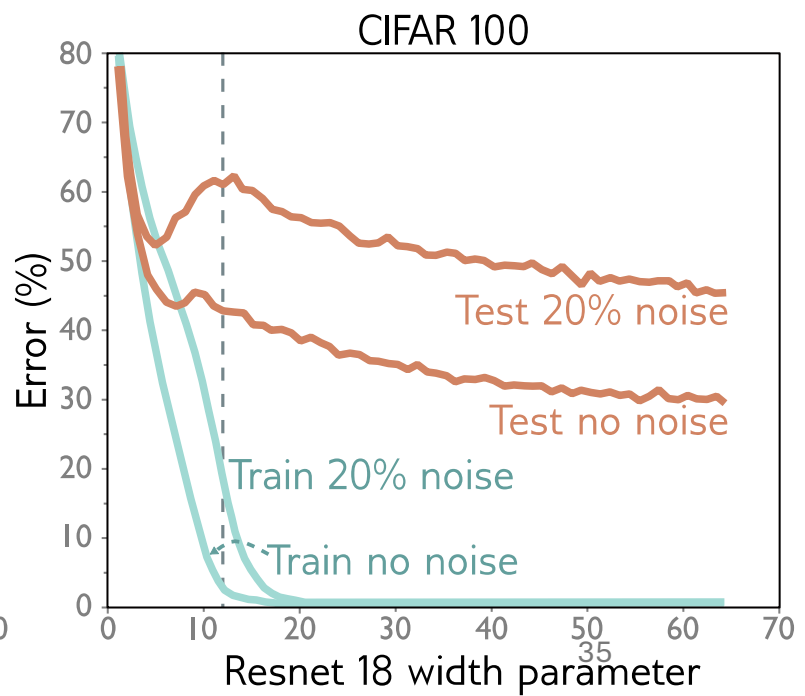
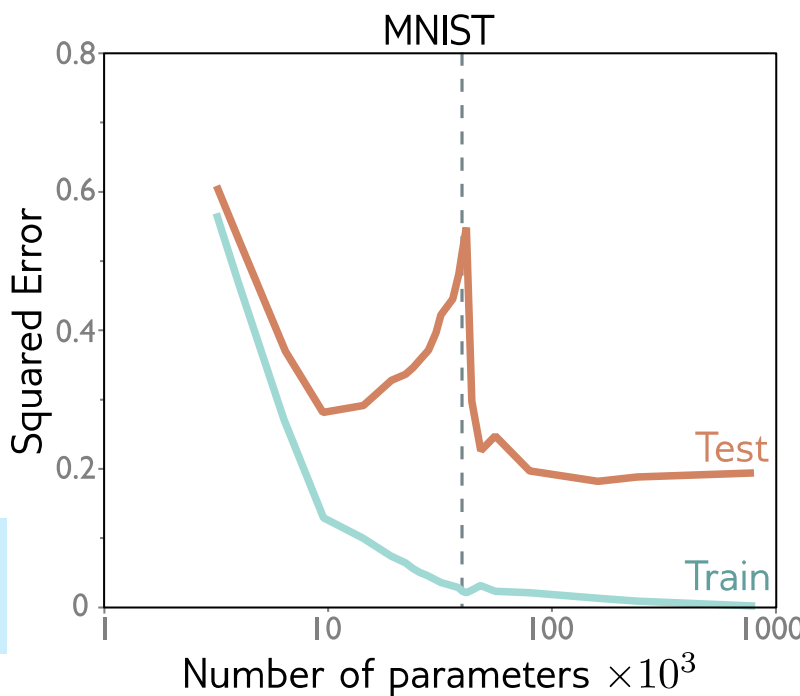
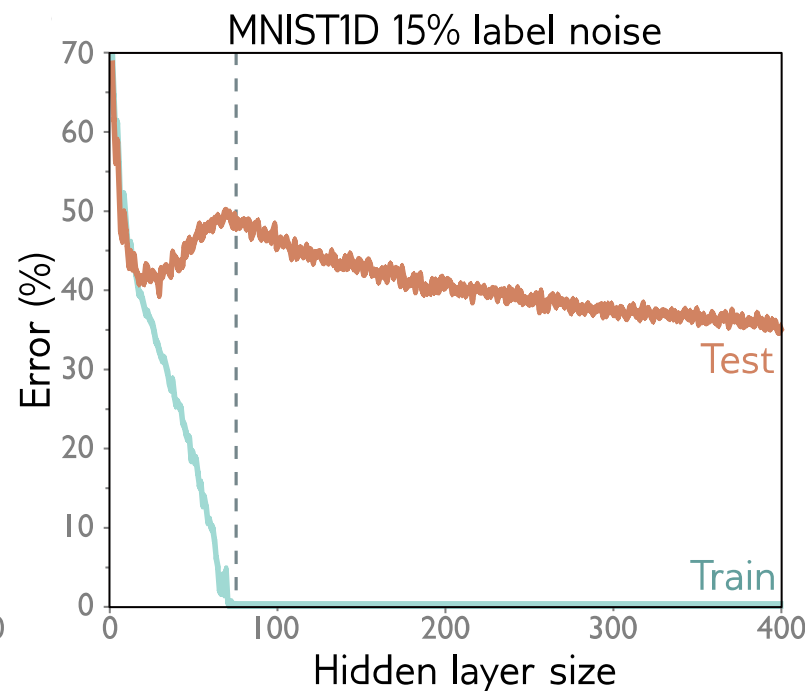
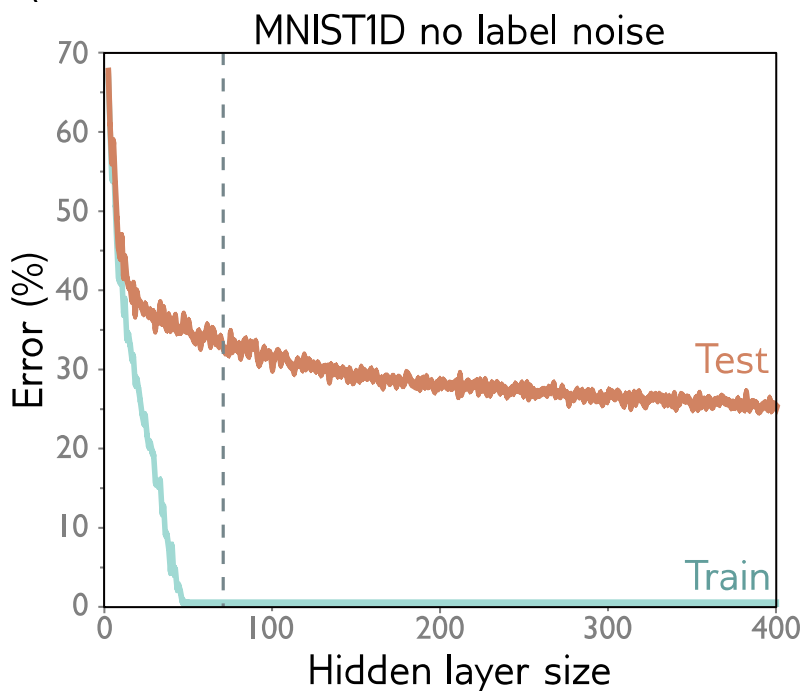
Modern or over-parameterized regime

Critical regime

Reminder: vertical dashed line is where:
training parameters = # training samples

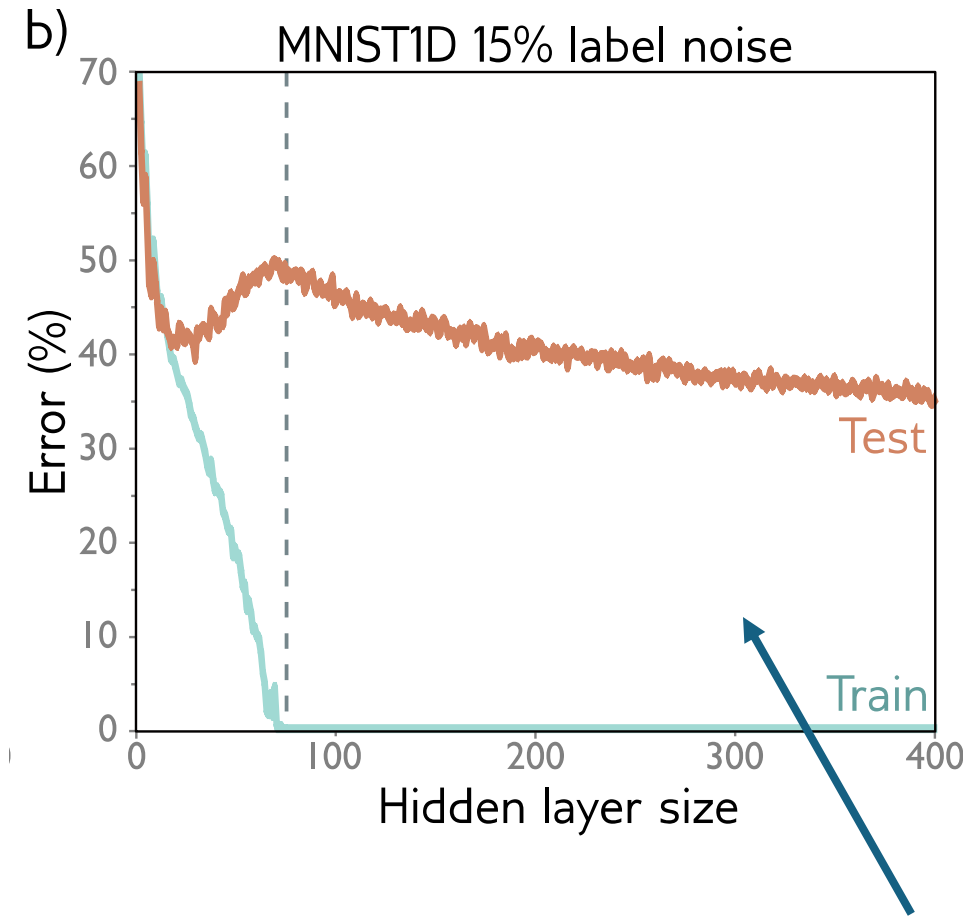
Same
phenomenon
shows up on
MNIST and
CIFAR100

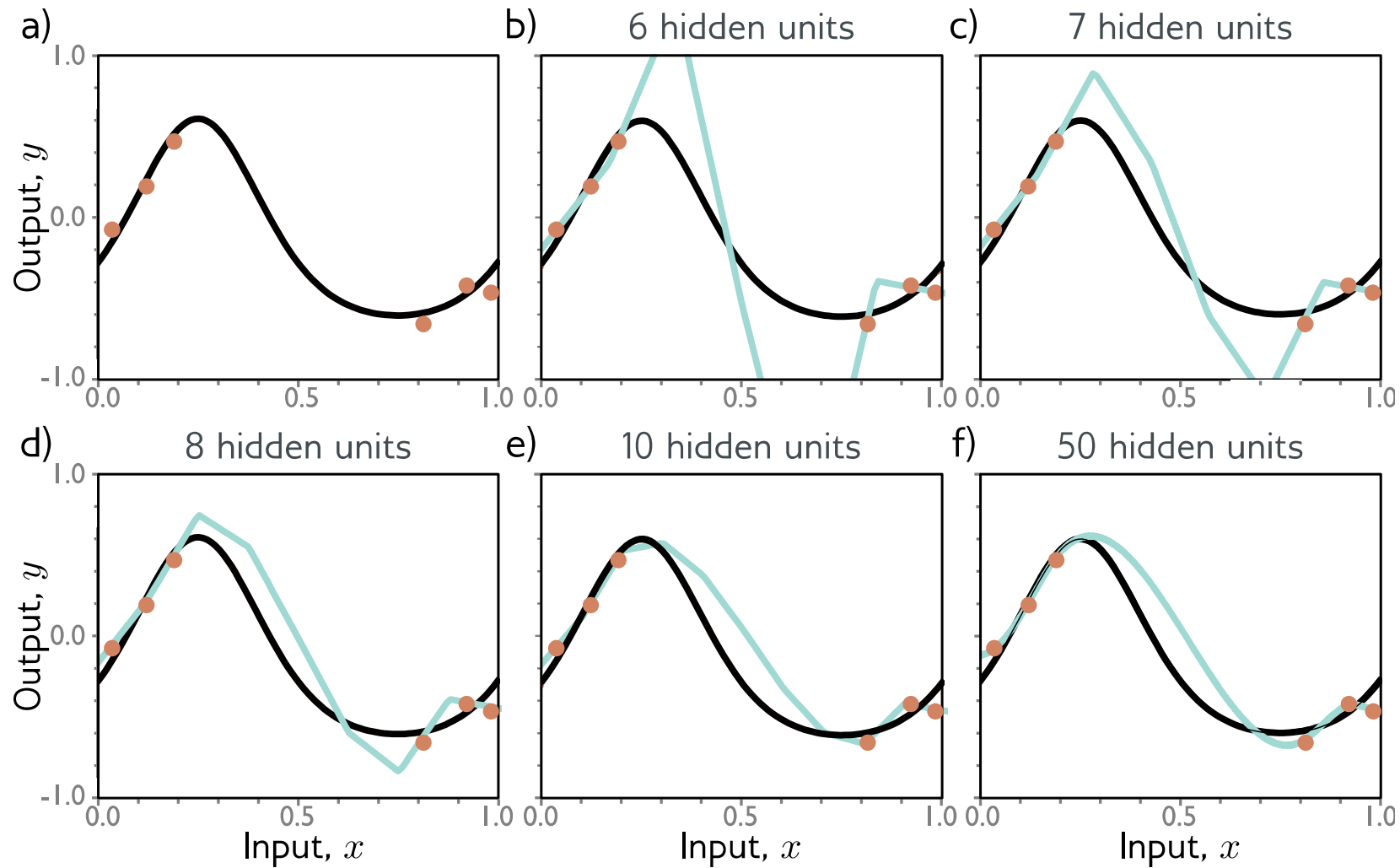
Reminder: vertical dashed line is where:
training parameters = # training samples



Double Descent

- Note that training loss is very close to zero.
- Whatever is happening isn't happening at training data points
- Model never sees test set during training
- Must be happening between the data points??



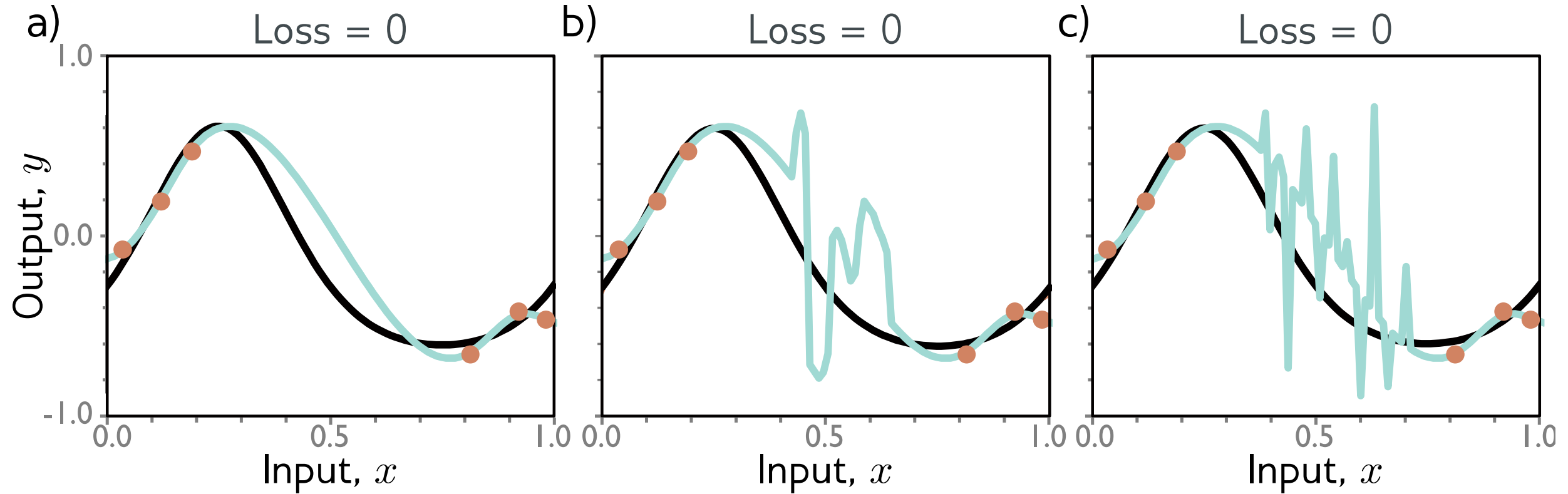


Potential explanation:

- can make smoother functions with more hidden units
- being smooth between the datapoints is a reasonable thing to do

But why?

Next Week: How to bias for smoothness?



- All of these solutions are equivalent in terms of loss.
- Why should the model choose the smooth solution?
- Tendency of model to choose one solution over another is **inductive bias**

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Choosing hyperparameters

- Don't know bias or variance
- Don't know how much capacity to add
- How do we choose capacity in practice?
 - Or model structure
 - Or training algorithm
 - Or learning rate
- Third data set – **validation set**
 - Train models with different hyperparameters on training set
 - Choose best hyperparameters with validation set
 - Test once with test set

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