

Geospatial Data Science  
Content Block II: *Techniques*  
Lab 8

Machine learning for geospatial data

Austin J. Brockmeier, Ph.D.

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# Outline

Lab 8: scikit-learn, regression, neural networks, convolutional neural networks



By The scikit-learn developers - [github.com/scikit-learn/scikit-learn/blob/master/doc/logos/scikit-learn-logo.svg](https://github.com/scikit-learn/scikit-learn/blob/master/doc/logos/scikit-learn-logo.svg), BSD,  
<https://commons.wikimedia.org/w/index.php?curid=71445288>

# Designing a Machine Learning System

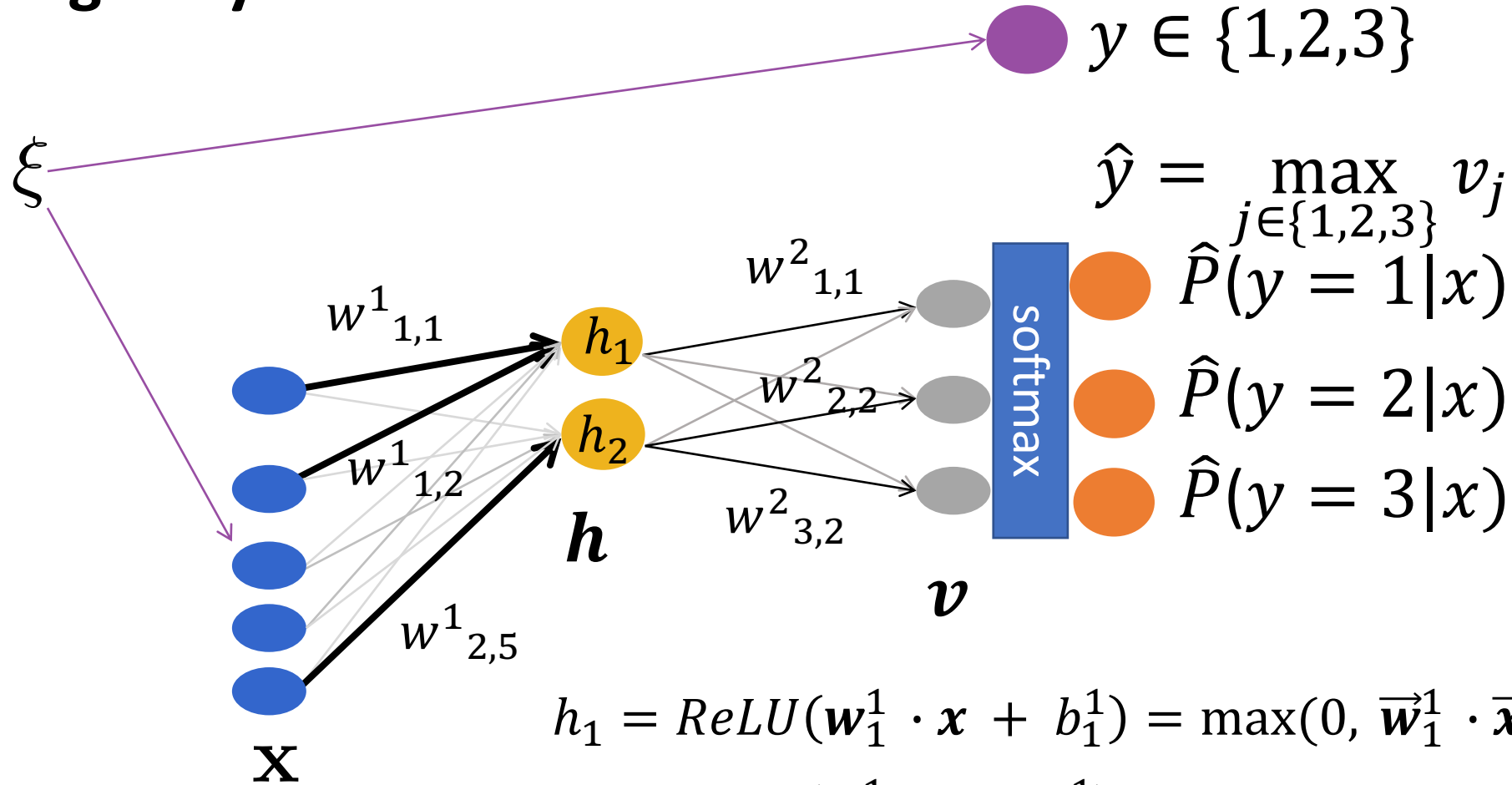
1. **Goal:** What is the task?
2. **Data:**
3. **Model:**
4. **Fitness:**
5. **Training:**
6. **Selection:**

# Model: Popular choices

- **Linear model**
- k-nearest neighbor
- Decision trees
- **Random forest**, gradient boosting
- Neural networks
- Convolutional neural network
- Kernel ridge regression
- Support vector machine
- **Gaussian processes**

Artificial neural networks consist of layers of processing connected together

# Single layer neural network



$$\hat{y} = \max_{j \in \{1, 2, 3\}} v_j$$

$$\hat{P}(y = 1|x)$$

$$\hat{P}(y = 2|x)$$

$$\hat{P}(y = 3|x)$$

$$h_1 = \text{ReLU}(\mathbf{w}_1^1 \cdot \mathbf{x} + b_1^1) = \max(0, \bar{\mathbf{w}}_1^1 \cdot \bar{\mathbf{x}} + b_1^1)$$

$$h_2 = \text{ReLU}(\mathbf{w}_2^1 \cdot \mathbf{x} + b_2^1)$$

$$v_j = \mathbf{w}_j^2 \cdot \mathbf{h} + b_j^2, \text{ for } j = 1, \dots, 3$$

$$\hat{P}(y = j|x) = \frac{e^{v_j}}{\sum_k e^{v_k}}, \text{ for } j = 1, \dots, 3$$

# CNN: Where's Waldo?

(Prediction yes or no for each image patch)



?



?



?

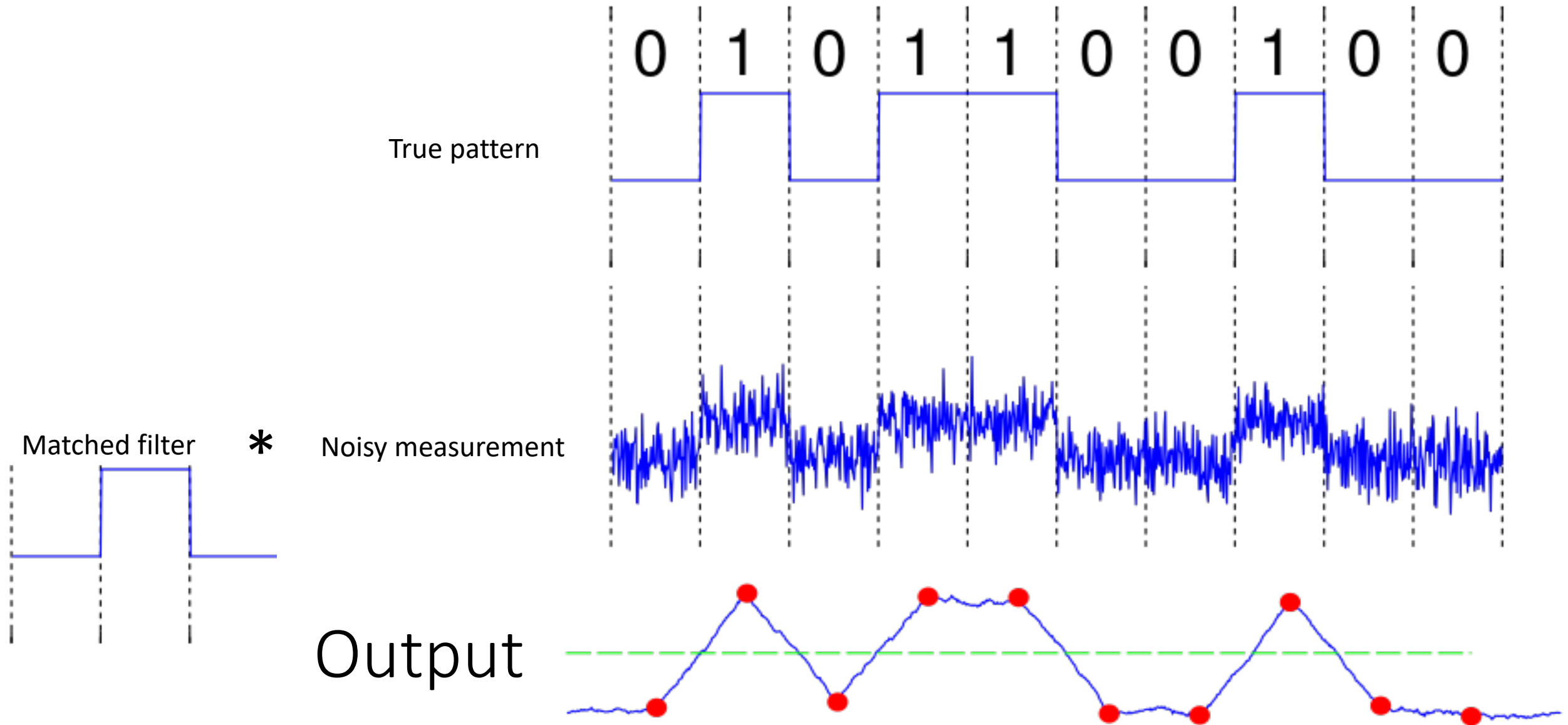


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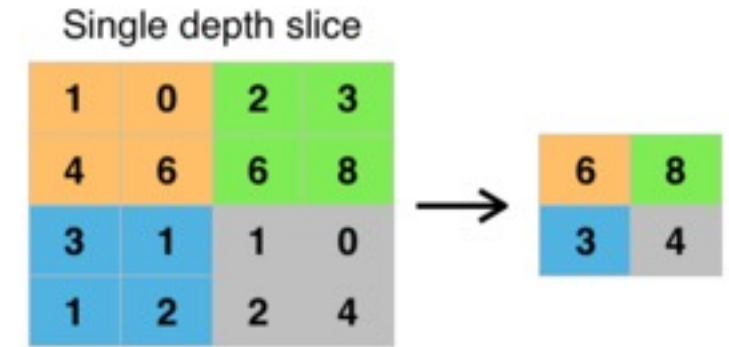
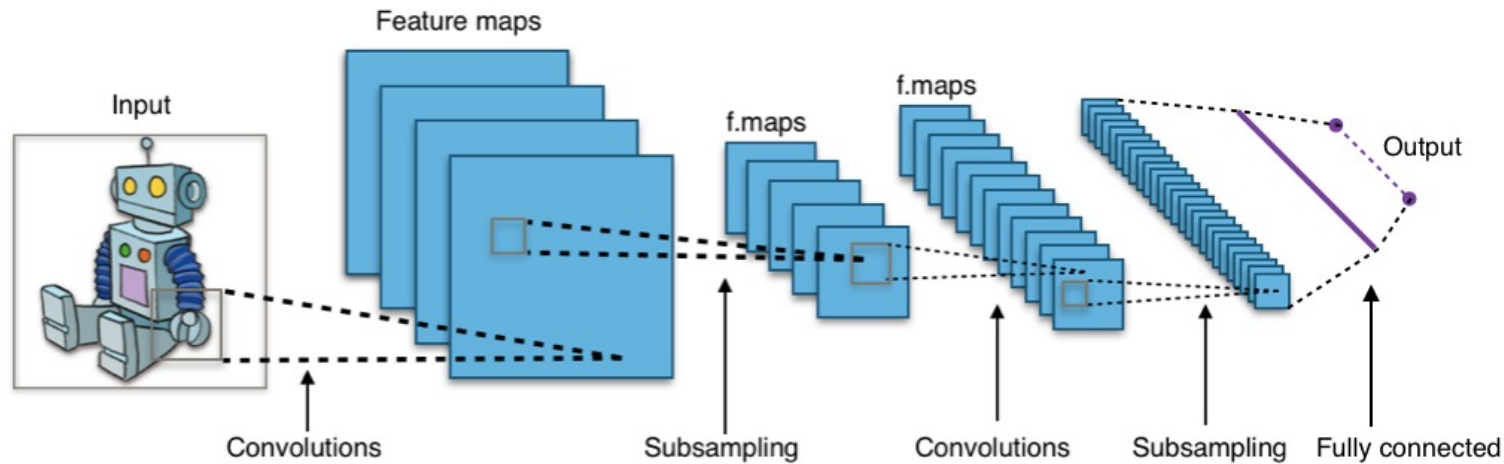
[https://eng.libretexts.org/Bookshelves/Electrical\\_Engineering/Signal\\_Processing\\_and\\_Modeling/Signals\\_and\\_Systems\\_%28Baraniuk\\_et\\_al.%29/13%3A\\_Capstone\\_Signal\\_Processing\\_Topics/13.04%3A\\_Matched\\_Filter\\_Detector](https://eng.libretexts.org/Bookshelves/Electrical_Engineering/Signal_Processing_and_Modeling/Signals_and_Systems_%28Baraniuk_et_al.%29/13%3A_Capstone_Signal_Processing_Topics/13.04%3A_Matched_Filter_Detector)

# Convolution or matched filtering

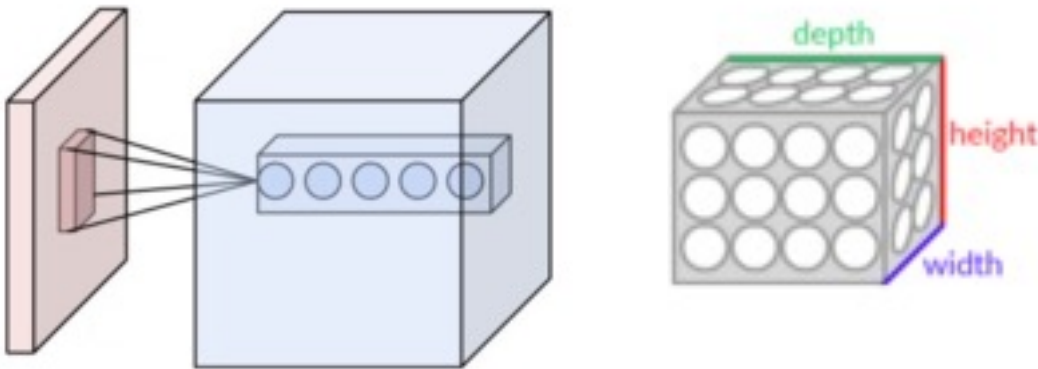




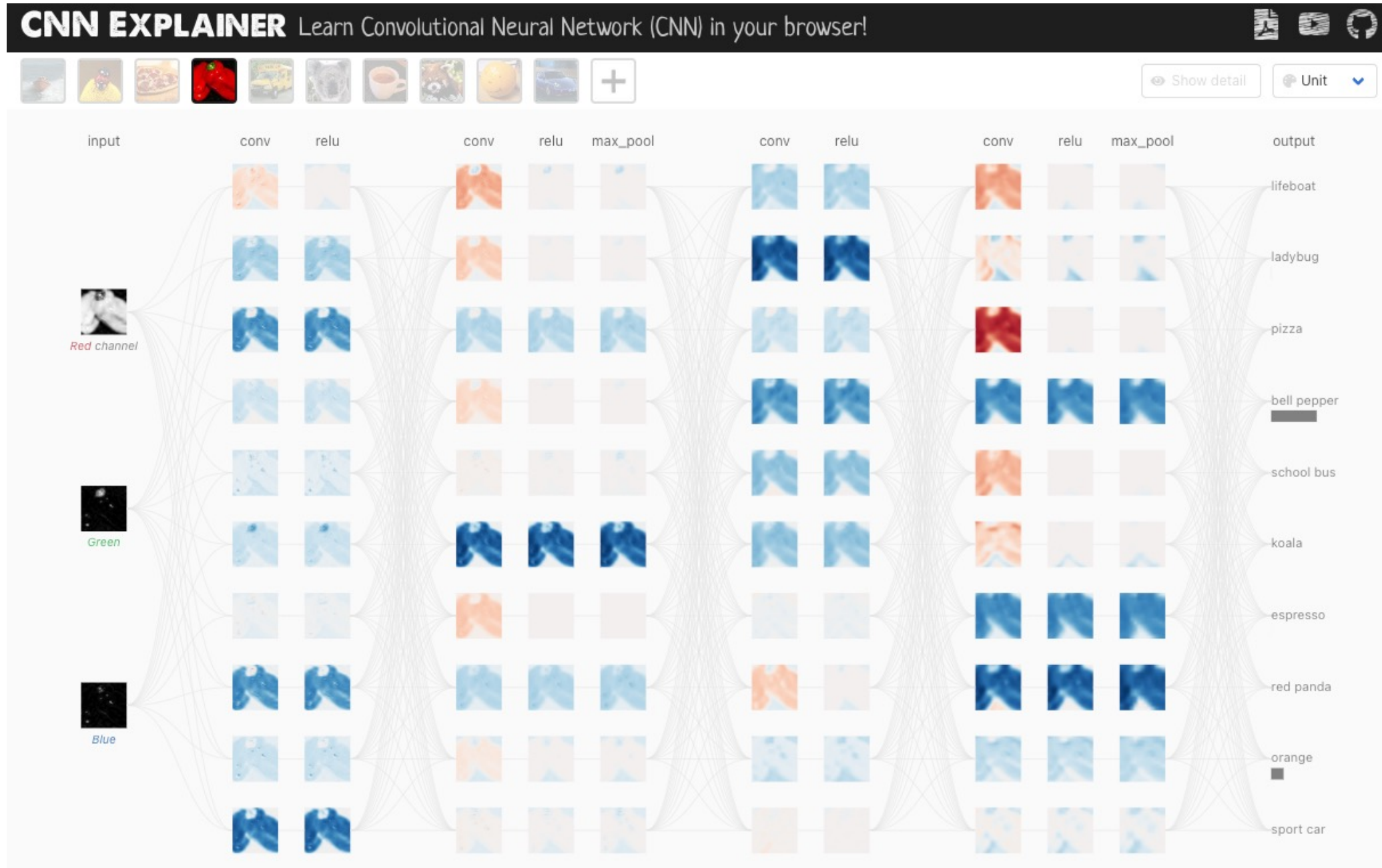
# Depth is the number of channels/attributes/layers



Subsampling via **max pooling** with a 2x2 filter and stride = 2



<https://poloclub.github.io/cnn-explainer>



# Example of CNN classifier on Fashion MNIST

- Error rate is 273/2646

$y = \text{shirt},$   
 $\hat{y} = \text{coat}$



$y = \text{top},$   
 $\hat{y} = \text{top}$



$y = \text{shirt},$   
 $\hat{y} = \text{shirt}$



$y = \text{coat},$   
 $\hat{y} = \text{shirt}$

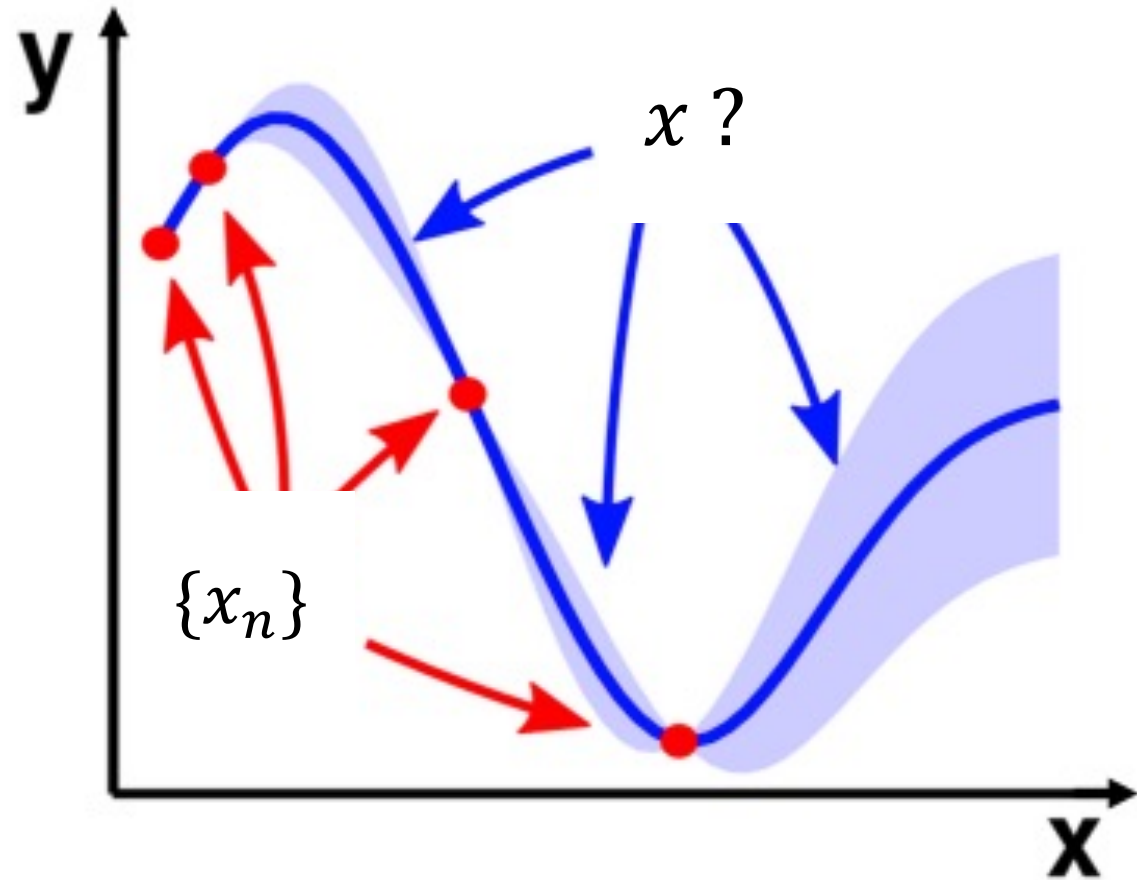


$y = \text{top},$   
 $\hat{y} = \text{top}$



# Non-linear models

- k-nearest neighbor
- Decision trees, random forests, gradient boosting
- Neural networks
- Kernel ridge regression
- Gaussian process



**Phase 1.** Fit relationship

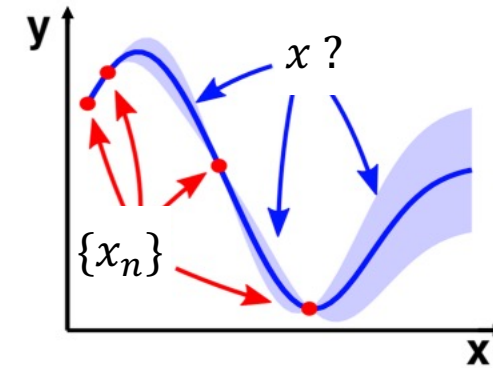
**Phase 2:** Find  $x$  that gives a specific  $y$  with high confidence (near seen data) and fits constraints!

- Kernel regression
  - Advanced by Prof. Grace Wahba at UW-Madison

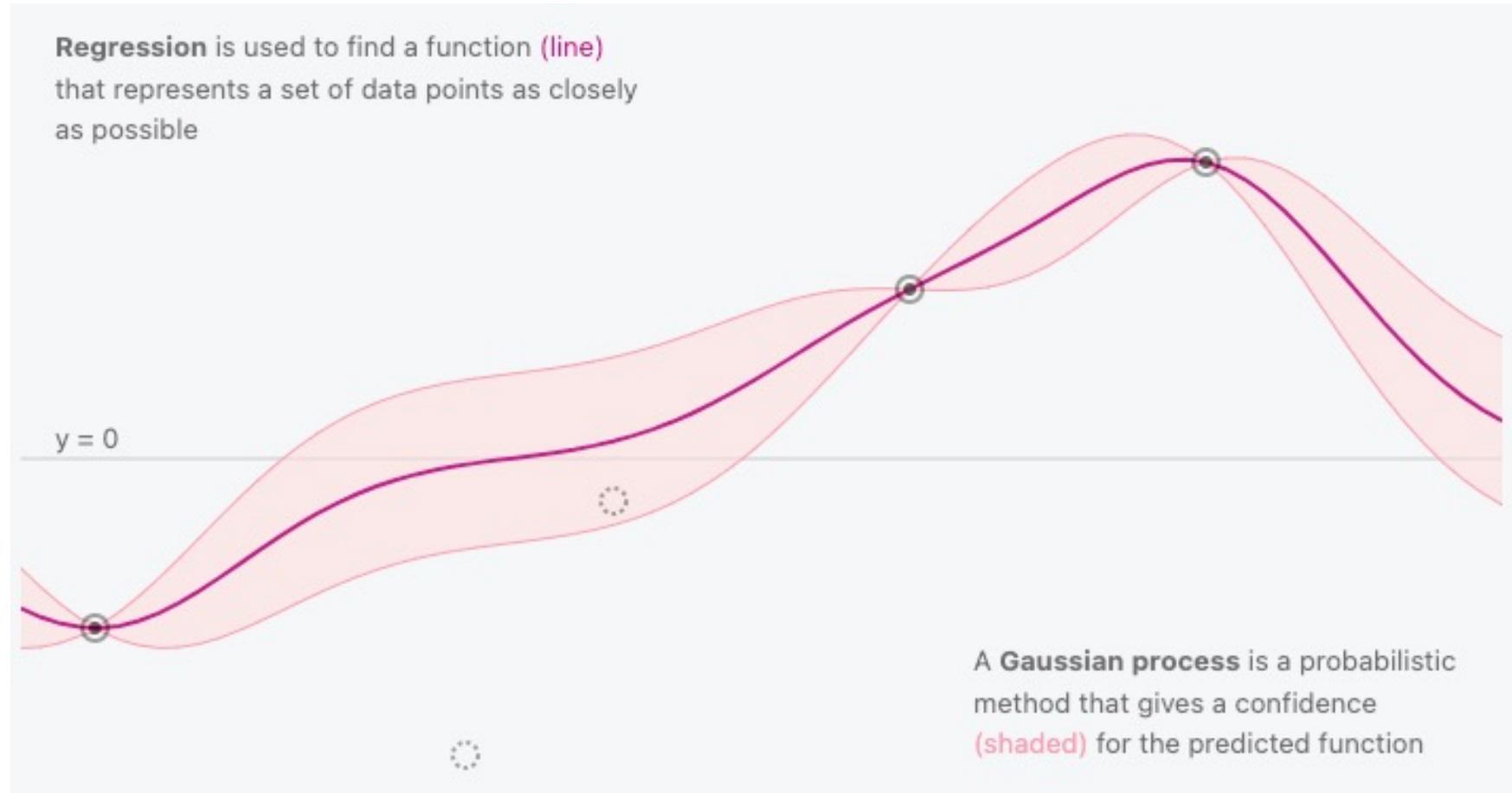


$$E[Y|x, \{(x_i, y_i)\}_{i=1}^n] = \bar{f}(x) = [\kappa(x, x_1), \dots, \kappa(x, x_n)]\mathbf{K}^{-1}\vec{y} = \mathbf{K}\vec{\alpha}$$

```
krr.fit(X, y).predict(x)
```



# Gaussian process (Kriging)



The predicted value at  $x$  is normally distributed with mean  $f(x)$ , and variance  $\sigma_x^2$   
 $\mathcal{N}(f(x), \sigma_x^2)$

$$\sigma_x^2 = \text{cov}(f(x), f(x)) = \kappa(x, x) - [\kappa(x, x_1), \dots, \kappa(x, x_N)] \mathbf{K}^{-1} [\kappa(x, x_1), \dots, \kappa(x, x_N)]$$