

# Model-Based Machine Learning

Applied ML Days, Lausanne

30 January, 2018

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# Machine learning algorithms

generative adversarial network  
K-means clustering  
decision trees  
kernel PCA  
principal components  
convolutional networks  
Gaussian mixture  
Independent component analysis  
Boltzmann machines

Markov random field  
Radial basis functions  
logistic regression  
Kalman filter  
random forest  
deep networks  
Hidden Markov model  
support vector machines  
linear regression  
neural networks  
factor analysis



# The 'No Free Lunch' Theorem

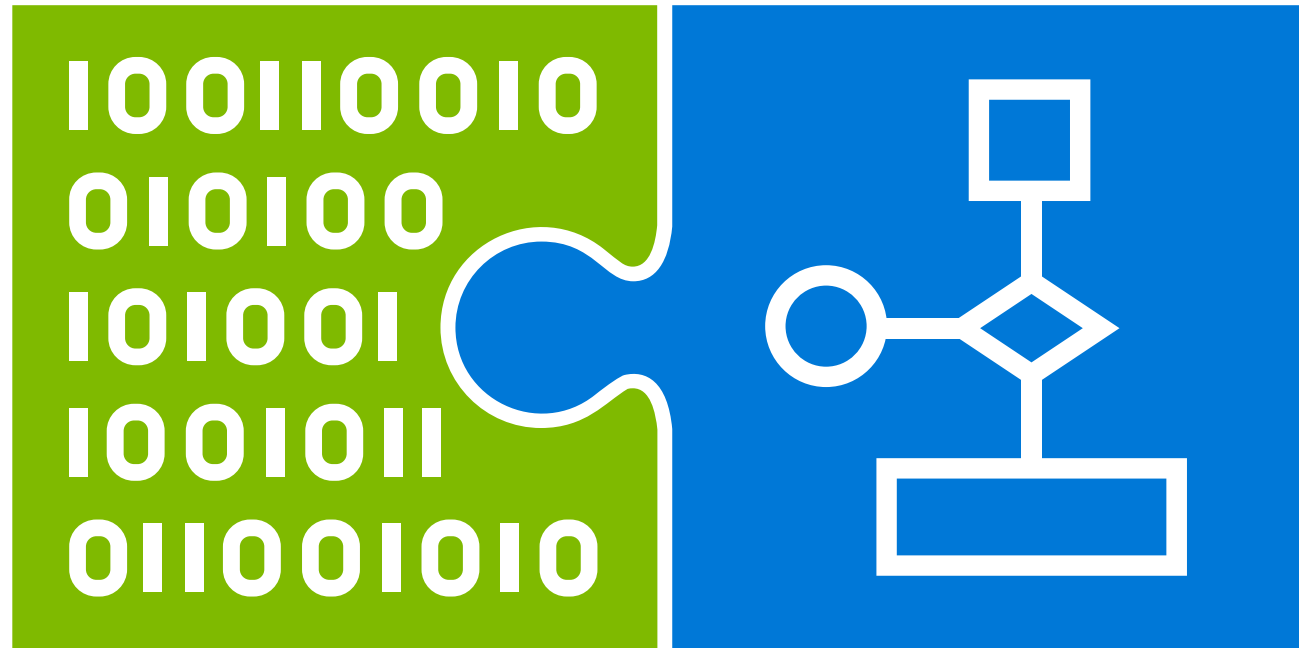
*Averaged over all possible data-generating distributions, every classification algorithm has the same error rate when classifying previously unobserved points.*

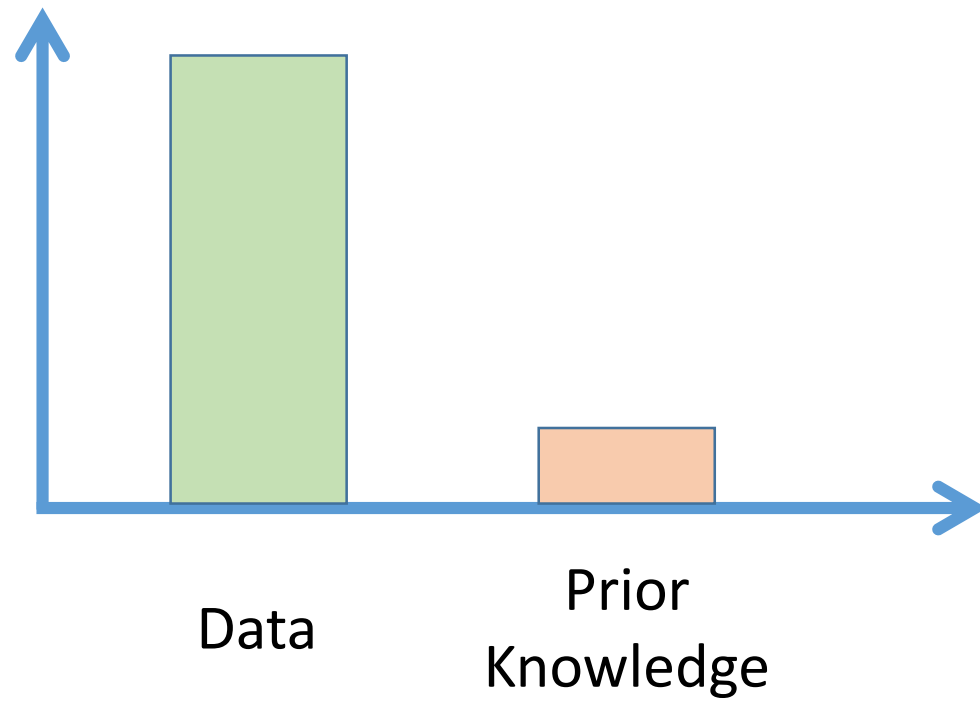
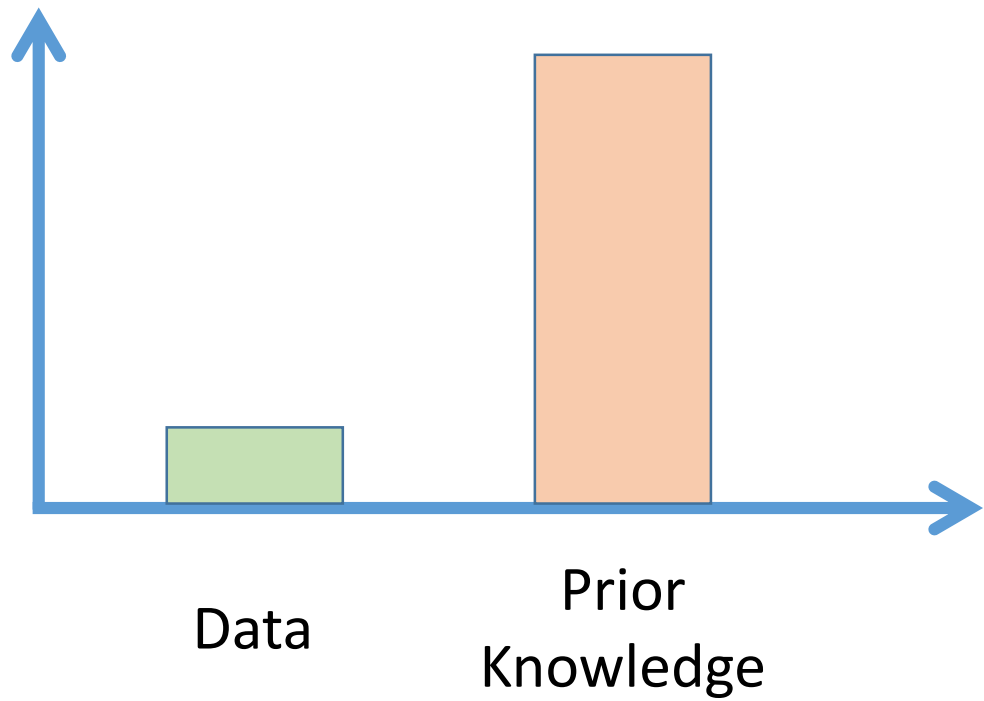
Wolpert (1996)

There is no universal machine learning algorithm

The goal of machine learning is to find an algorithm that is well matched to the problem being solved

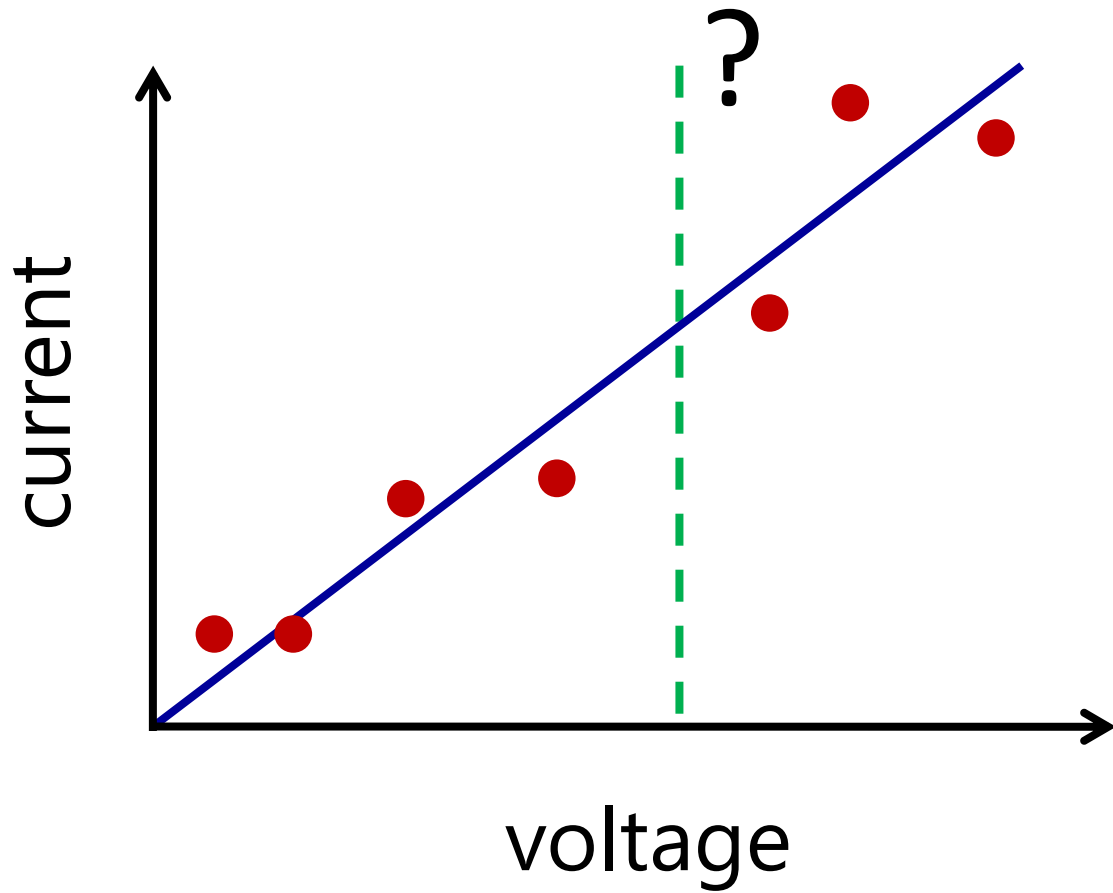
# Machine Learning







# 'Big data'



# Model-based machine learning

Derive the appropriate ML algorithm by making modelling assumptions explicit

*Traditional:*

“how do I map my problem onto standard algorithms”?

*Model-based:*

“what is the *model* that represents my problem”?

Machine learning algorithm



# PCA as an algorithm

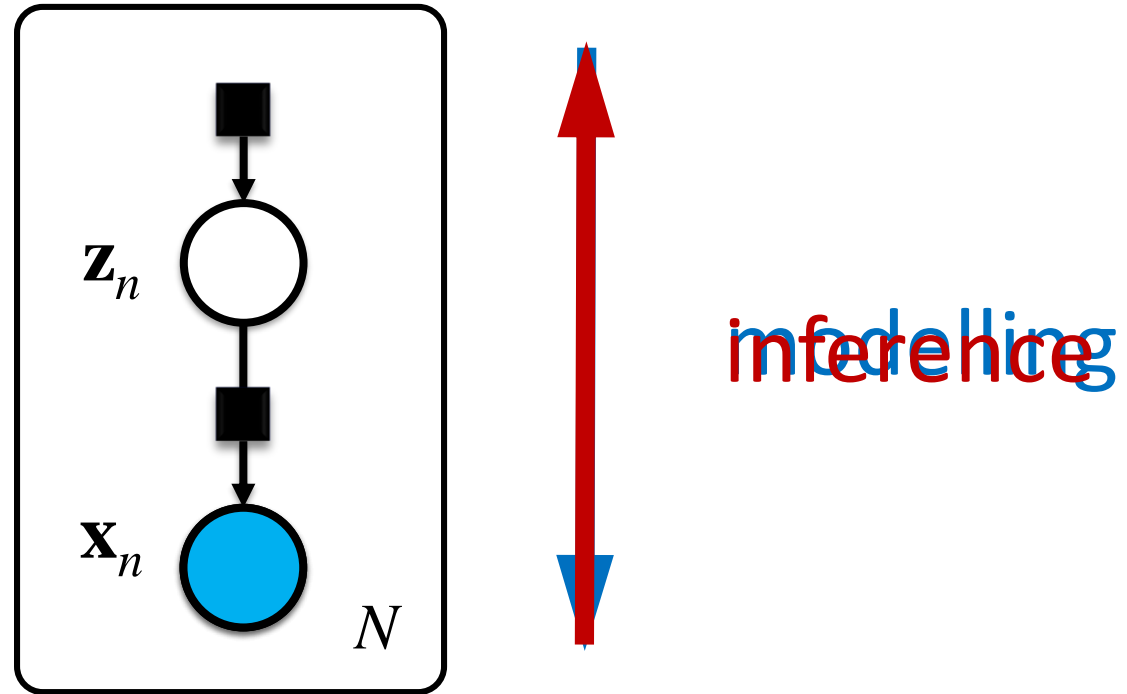
$$\bar{\mathbf{x}} = \frac{1}{N} \sum_{n=1}^N \mathbf{x}_n$$

$$\mathbf{S} = \frac{1}{N} \sum_{n=1}^N (\mathbf{x}_n - \bar{\mathbf{x}})(\mathbf{x}_n - \bar{\mathbf{x}})^T$$

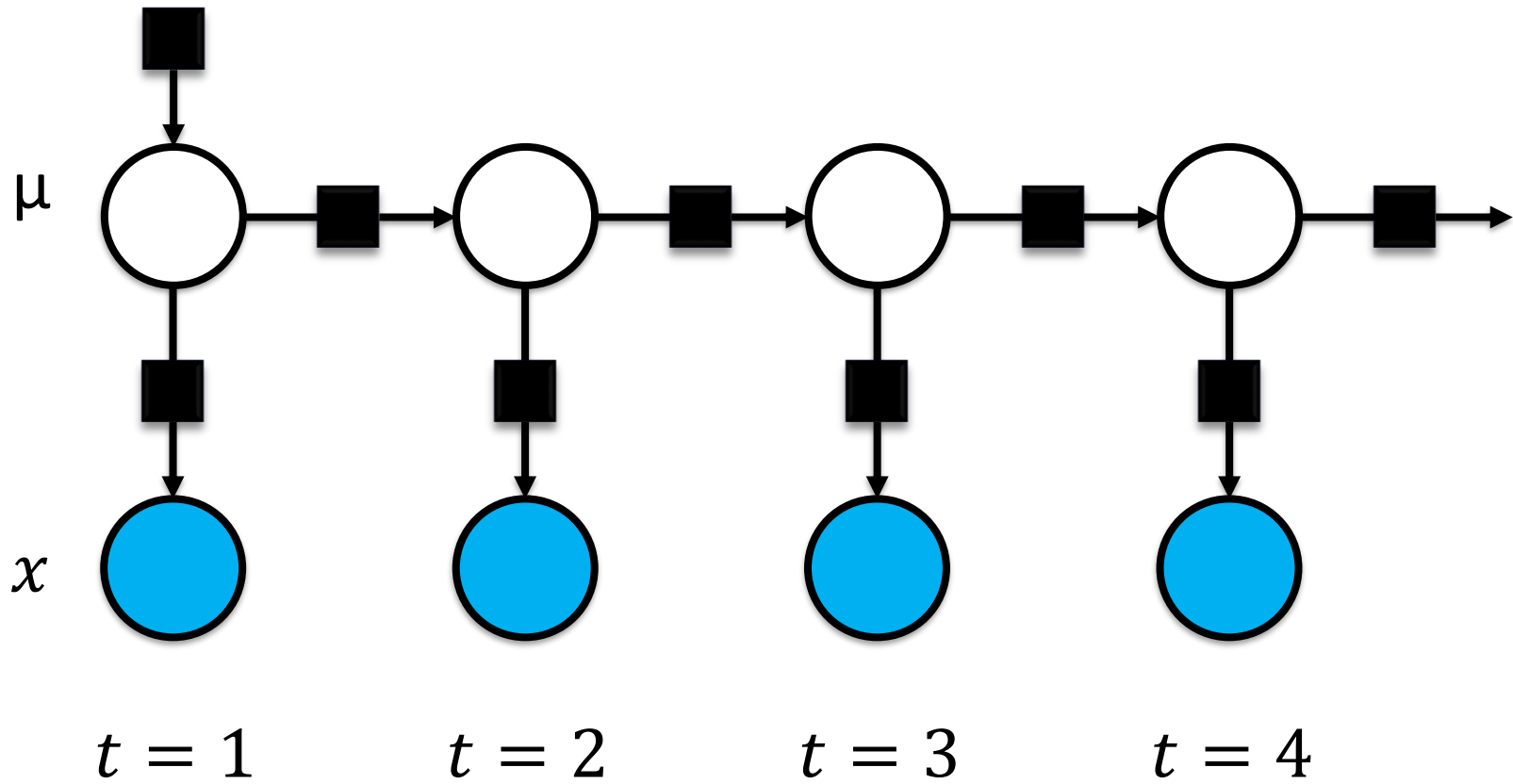
$$\mathbf{S}\mathbf{u}_i = \lambda_i \mathbf{u}_i$$

retain  $M < D$  eigenvectors

# PCA as a model



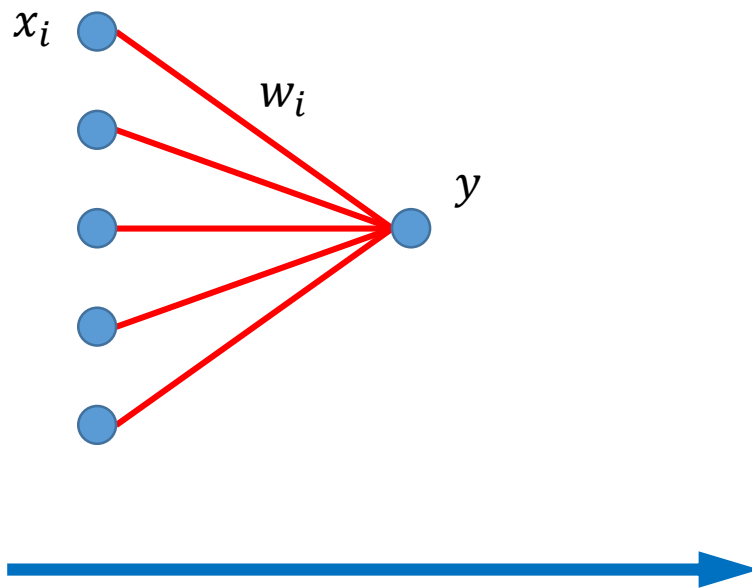
M. E. Tipping and C. M. Bishop (1997)



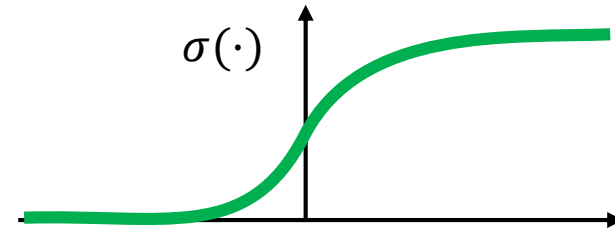
*The Kalman filter*

*The hidden Markov model*

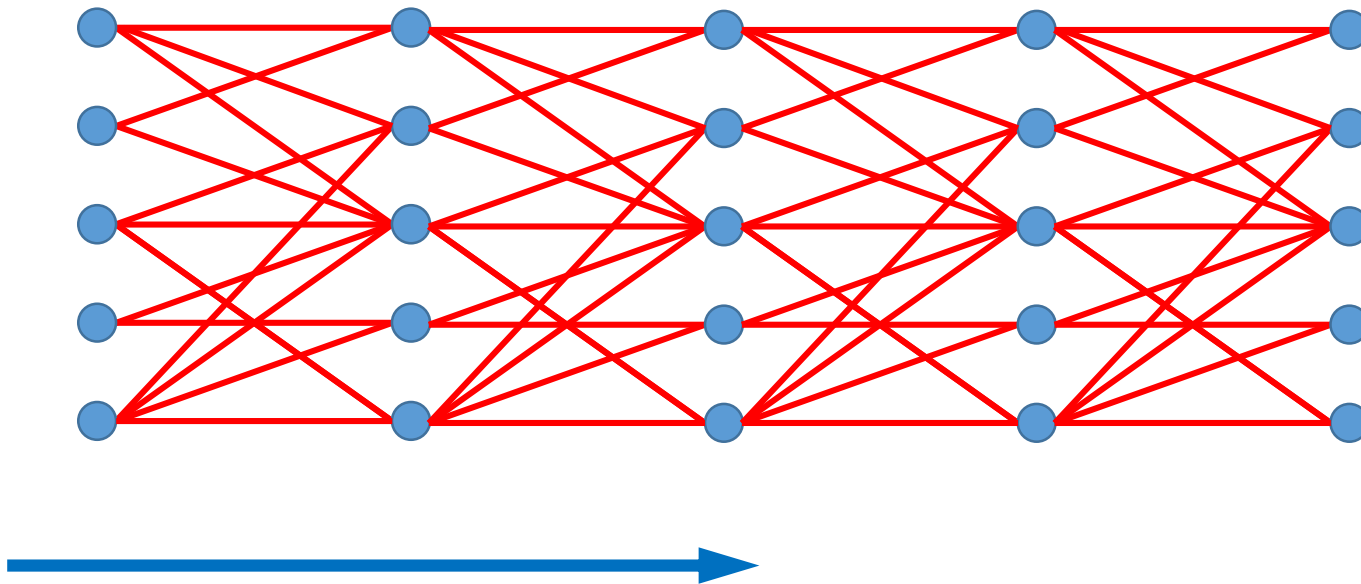
# Logistic Regression



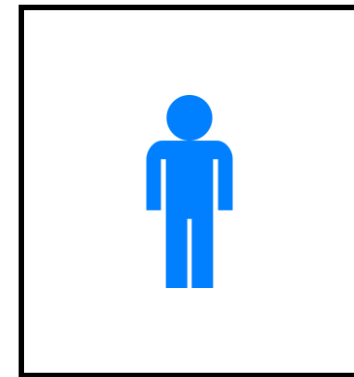
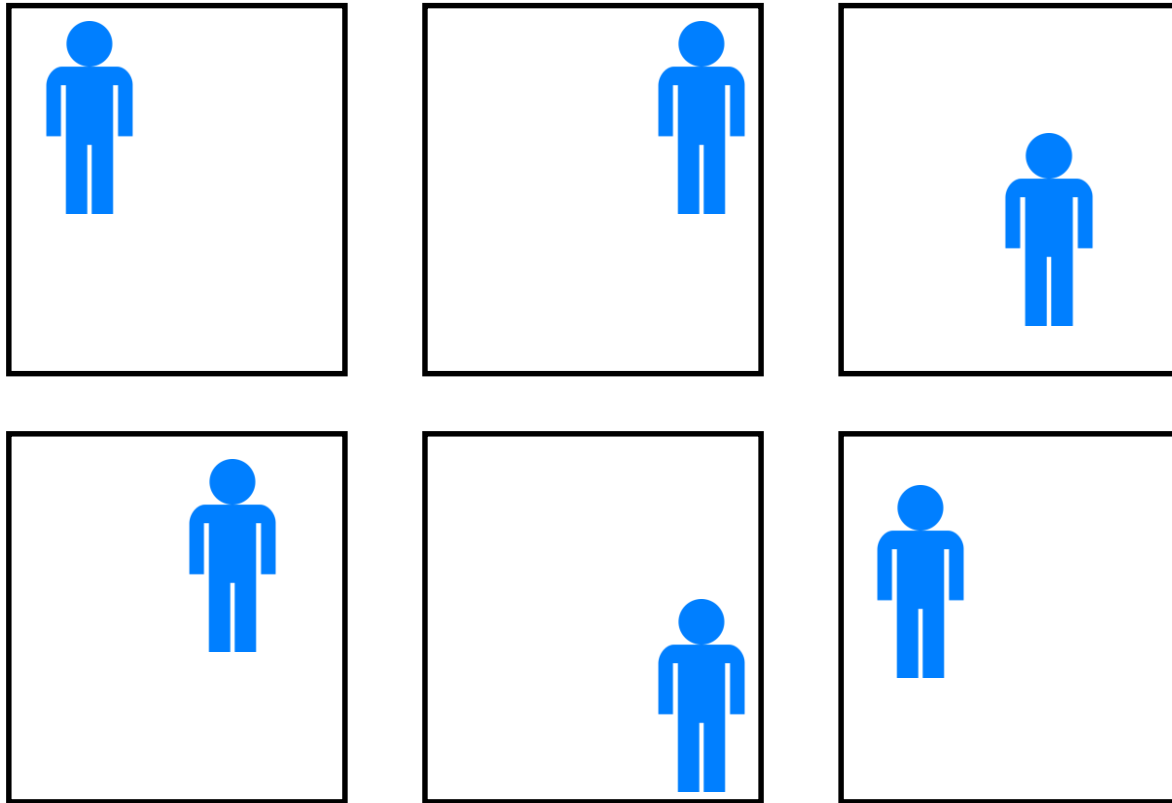
$$y = \sigma \left( \sum_i w_i x_i \right)$$



# Deep Neural Networks



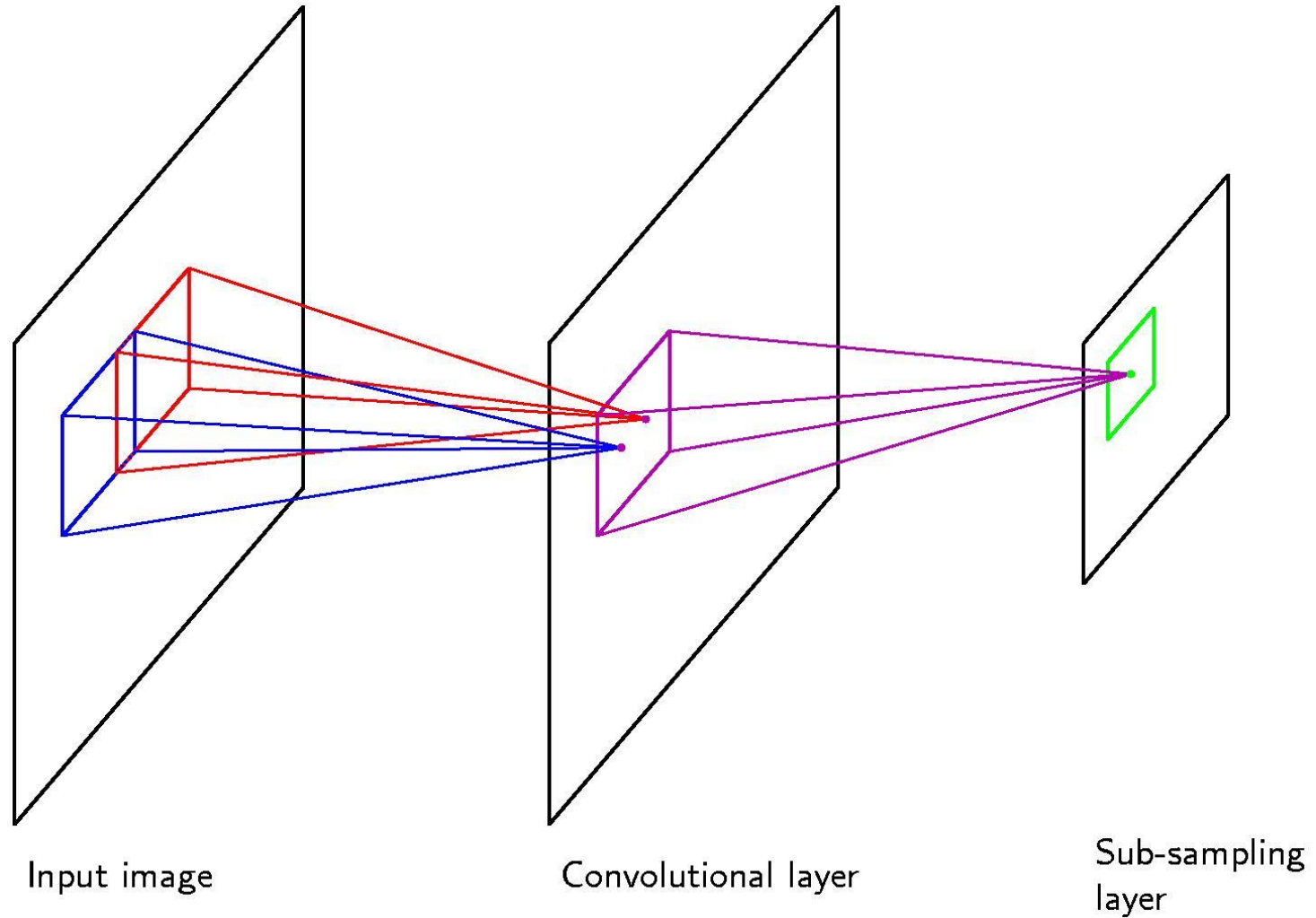
# Data and prior knowledge



Translation invariance



# Convolutional Neural Networks



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# Model-Based Machine Learning

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John Winn and Christopher Bishop  
with  
Thomas Diethe

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Thank you!