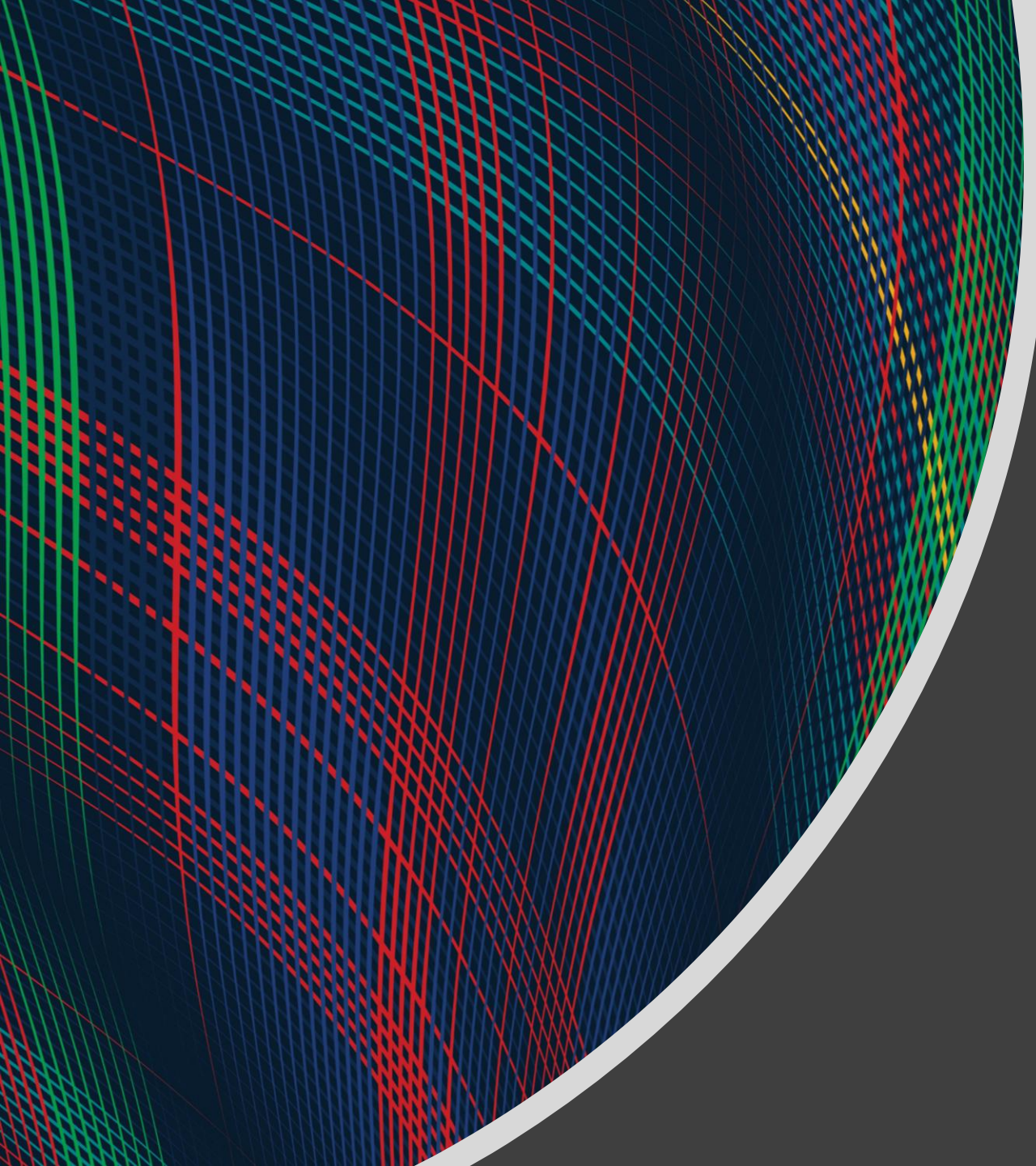


An abstract graphic on the left side of the slide, featuring a sphere-like shape composed of a dense grid of intersecting red, green, and blue lines. The lines are curved and follow the contours of the sphere, creating a complex, woven pattern. The sphere is set against a dark gray background.

10-315
Introduction to ML

Decision Trees

Instructor: Pat Virtue

Today

Autoencoders

More course info

Announcements

ML data and notation

Using data to learn

Decision trees



Autoencoders

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Machine Learning Systems

Three components $\langle T, P, E \rangle$:

1. Task, T
2. Performance measure, P
3. Experience, E

Definition of learning:

A computer program **learns** if its performance at tasks in T , as measured by P , improves with experience E .

Input/Output Tasks

Task

numerical data → Iris classification → category

hour/time → Traffic prediction → num

img → Image classification → category

img → Image denoising → img

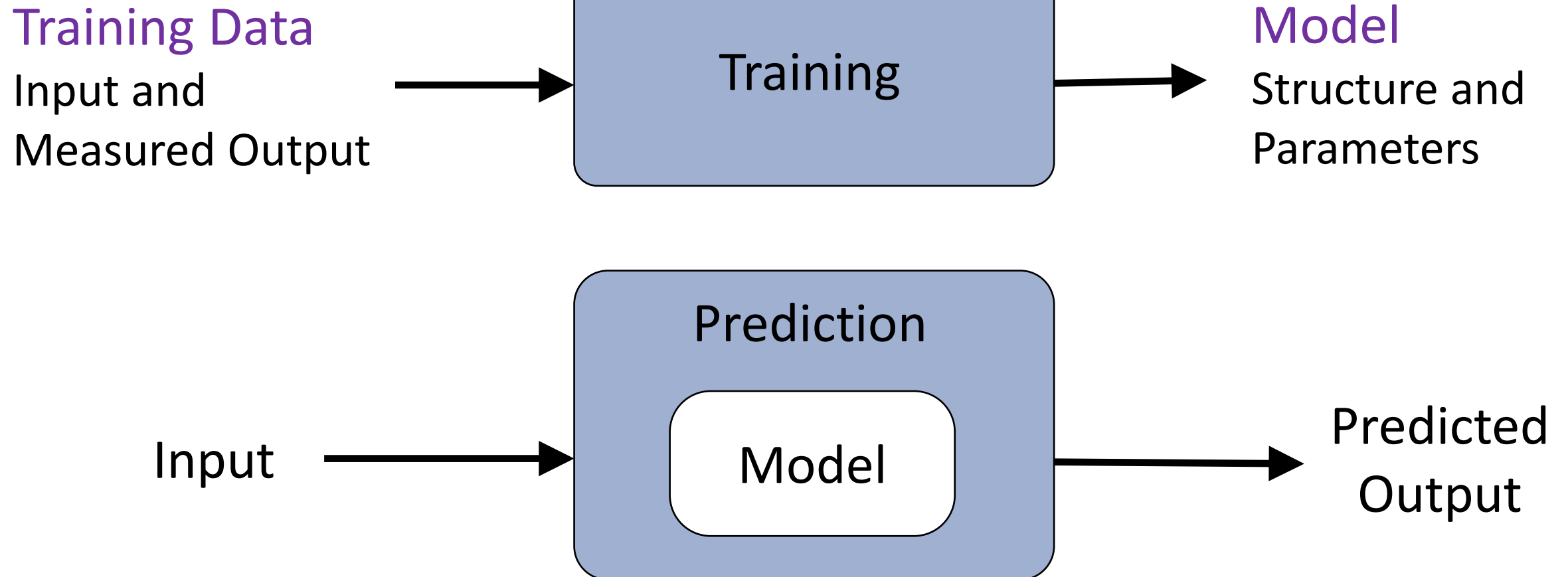
img → Medical image recon → img

text → Text to image generation → img

rand num → Face generation → img

Machine Learning

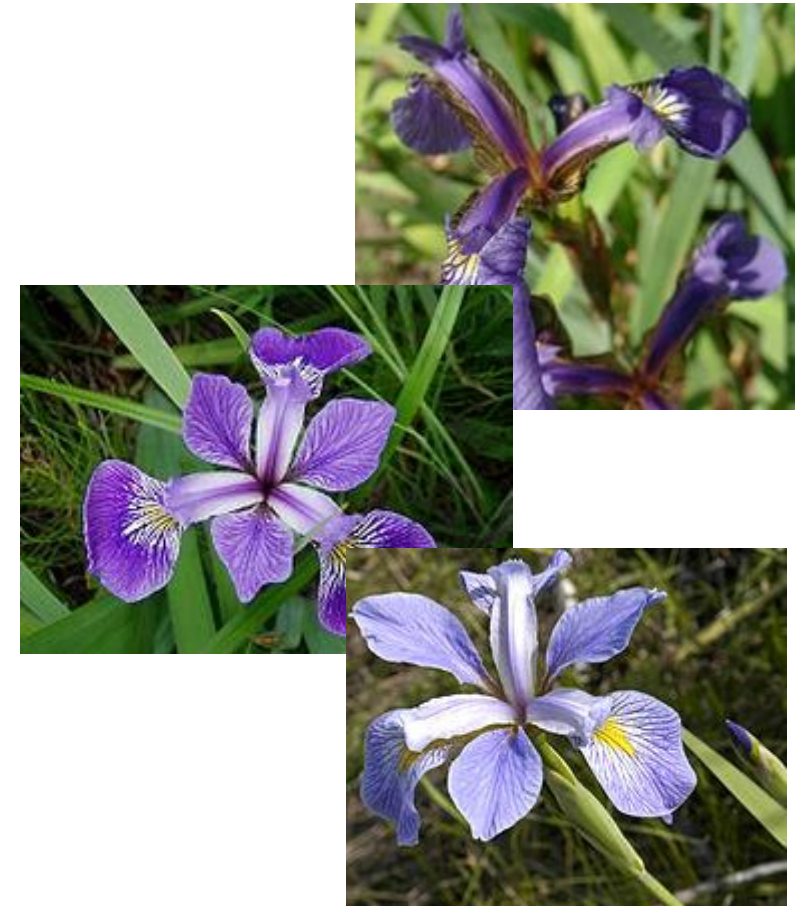
Using (training) data to learn a model that we'll later use for prediction



Example Dataset: Fisher Iris Dataset

Fisher (1936) used 150 measurements of flowers from 3 different species: *Iris setosa* (0), *Iris virginica* (1), *Iris versicolor* (2) collected by Anderson (1936)

Species	Sepal Length	Sepal Width	Petal Length	Petal Width
0	4.3	3.0	1.1	0.1
0	4.9	3.6	1.4	0.1
0	5.3	3.7	1.5	0.2
1	4.9	2.4	3.3	1.0
1	5.7	2.8	4.1	1.3
1	6.3	3.3	4.7	1.6
2	5.9	3.0	5.1	1.8



Images and full dataset: https://en.wikipedia.org/wiki/Iris_flower_data_set

Example Dataset: Fisher Iris Dataset

Assume samples in data are i.i.d.

```
from sklearn import datasets

iris = datasets.load_iris()
X = iris.data
y = iris.target
```

Dataset notation

$$\mathcal{D} = \left\{ \left(y^{(i)}, \mathbf{x}^{(i)} \right) \right\}_{i=1}^N$$
$$= \left\{ \left(y^{(i)}, x_1^{(i)}, x_2^{(i)}, x_3^{(i)}, x_4^{(i)} \right) \right\}_{i=1}^N$$

Linear algebra can represent all data

$$\mathbf{y} \in \{0, 1, 2\}^N$$

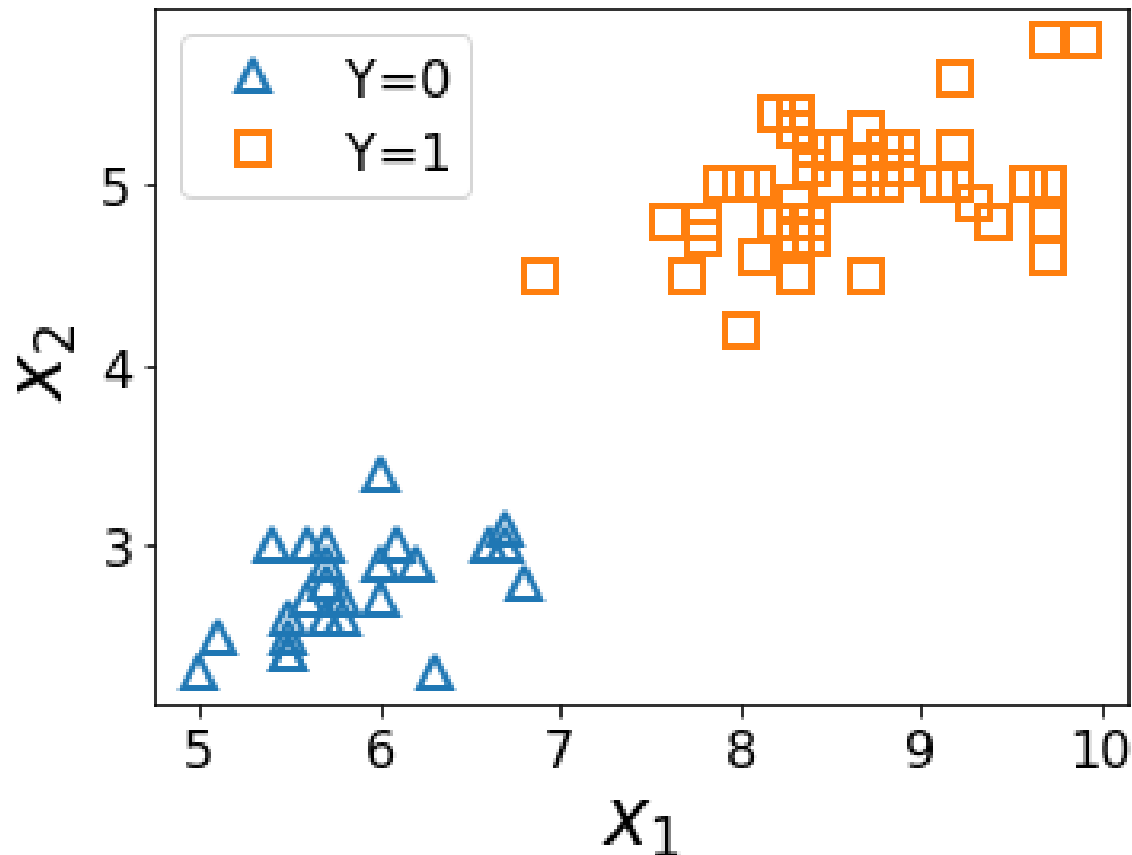
$$\mathbf{X} \in \mathbb{R}^{N \times 4} \quad (\text{design matrix})$$

Species	Sepal Length	Sepal Width	Petal Length	Petal Width
0	4.3	3.0	1.1	0.1
0	4.9	3.6	1.4	0.1
0	5.3	3.7	1.5	0.2
1	4.9	2.4	3.3	1.0
1	5.7	2.8	4.1	1.3
1	6.3	3.3	4.7	1.6
2	5.9	3.0	5.1	1.8

ML Task: Classification

Predict species label from first two input measurements

$$h(\mathbf{x}) \rightarrow \hat{y}$$



Species	Sepal Length	Sepal Width
0	4.3	3.0
0	4.9	3.6
0	5.3	3.7
1	4.9	2.4
1	5.7	2.8
1	6.3	3.3
2	5.9	3.0

ML Tasks

Supervised learning

$$\mathcal{D} = \{(y^{(i)}, \mathbf{x}^{(i)})\}_{i=1}^N \quad h(\mathbf{x}) \rightarrow \hat{y}$$

Classification

- Output labels
- $y \in \mathcal{Y}$, where $\mathcal{Y}$ is discrete and order of values has no meaning

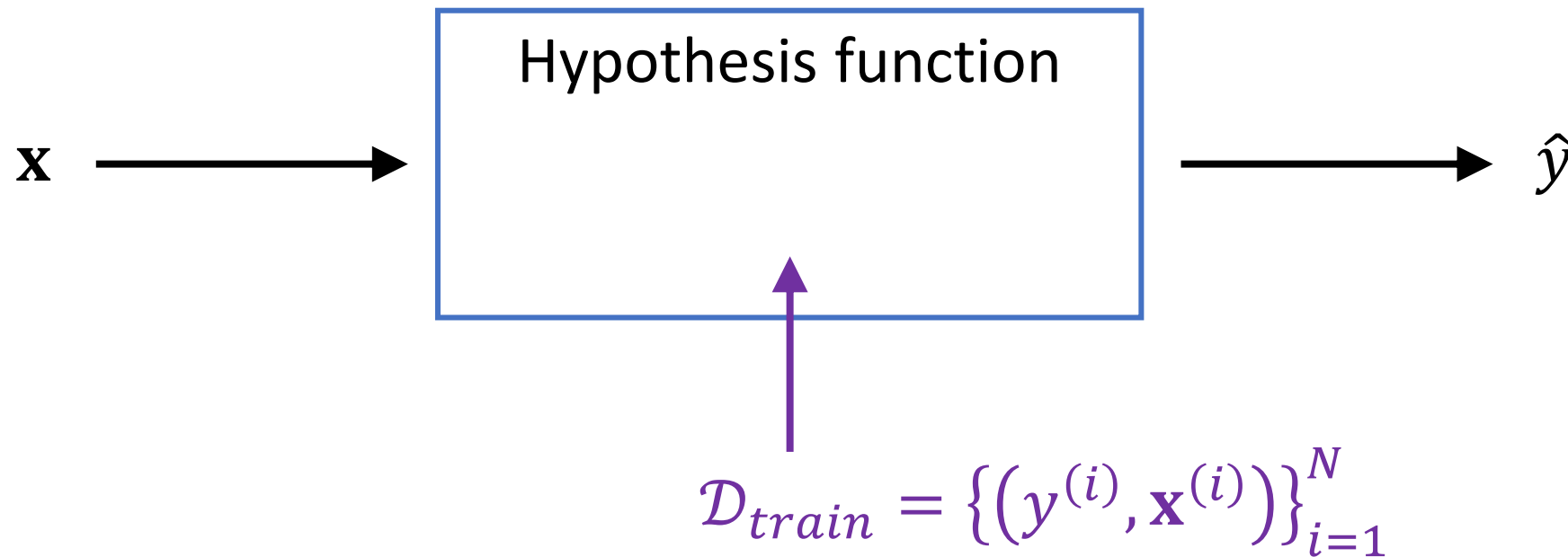
Regression

- Output values
- $y \in \mathcal{Y}$, where $\mathcal{Y}$ is usually continuous, order of values has meaning

Supervised Learning

Use training dataset to learn “best” hypothesis function

Predict output from input features

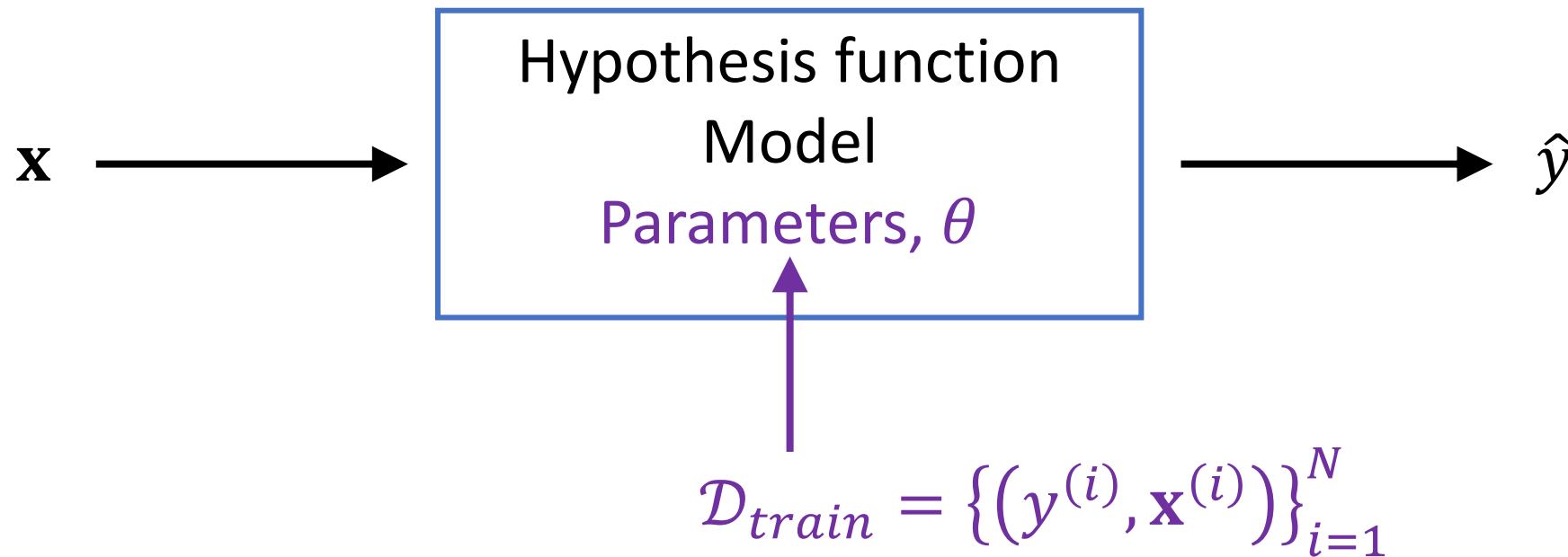


Need **performance measure** to determine meaning of “best” function

Supervised Learning

Use training dataset to learn “best” hypothesis function

Predict output from input features



Need **performance measure** to determine meaning of “best” function

Classification

Iris data example

$$\mathcal{D} = \{(y^{(i)}, \mathbf{x}^{(i)})\}_{i=1}^N, \text{ where } \mathbf{x}^{(i)} \in \mathbb{R}^4, y^{(i)} \in \{0, 1, 2\}$$

Predict species label from input measurements

$$h(\mathbf{x}) \rightarrow \hat{y}$$

Performance measure?

Classification error rate

- Fraction of times $y \neq \hat{y}$ in a given dataset

Species	Sepal Length	Sepal Width	Petal Length	Petal Width
0	4.3	3.0	1.1	0.1
0	4.9	3.6	1.4	0.1
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1	6.3	3.3	4.7	1.6
2	5.9	3.0	5.1	1.8

Classification

Iris data example

Notation alert: Indicator function

$$\mathbb{1}(z) = \mathbf{1}(z) = \begin{cases} 1 & \text{if } z \text{ is true} \\ 0 & \text{otherwise} \end{cases}$$

$$\mathcal{D} = \{(y^{(i)}, \mathbf{x}^{(i)})\}_{i=1}^N, \text{ where } \mathbf{x}^{(i)} \in \mathbb{R}^4, y^{(i)} \in \{0, 1, 2\}$$

Predict species label from input measurements

$$h(\mathbf{x}) \rightarrow \hat{y}$$

Performance measure?

Classification error rate

- Fraction of times $y \neq \hat{y}$ in a given dataset
- $\frac{1}{N} \sum_{i=1}^N \mathbb{1}(y^{(i)} \neq \hat{y}^{(i)})$

Species	Sepal Length	Sepal Width	Petal Length	Petal Width
0	4.3	3.0	1.1	0.1
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Problem Formulation

Medical Prediction

Outcome	Fetal Position	Fetal Distress	Previous C-sec
Natural	Vertex	N	N
C-section	Breech	N	N
Natural	Vertex	Y	Y
C-section	Vertex	N	Y
Natural	Abnormal	N	N

Problem Formulation

Medical Prediction

Y	X_1	X_2	X_3
Outcome	Fetal Position	Fetal Distress	Previous C-sec
Natural	Vertex	N	N
C-section	Breech	N	N
Natural	Vertex	Y	Y
C-section	Vertex	N	Y
Natural	Abnormal	N	N

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = [x_1, x_2, x_3]^T$$

$$x_1 \in \{Vertex, Breech, Abn\}$$

$$x_2 \in \{Y, N\}$$

$$x_3 \in \{Y, N\}$$

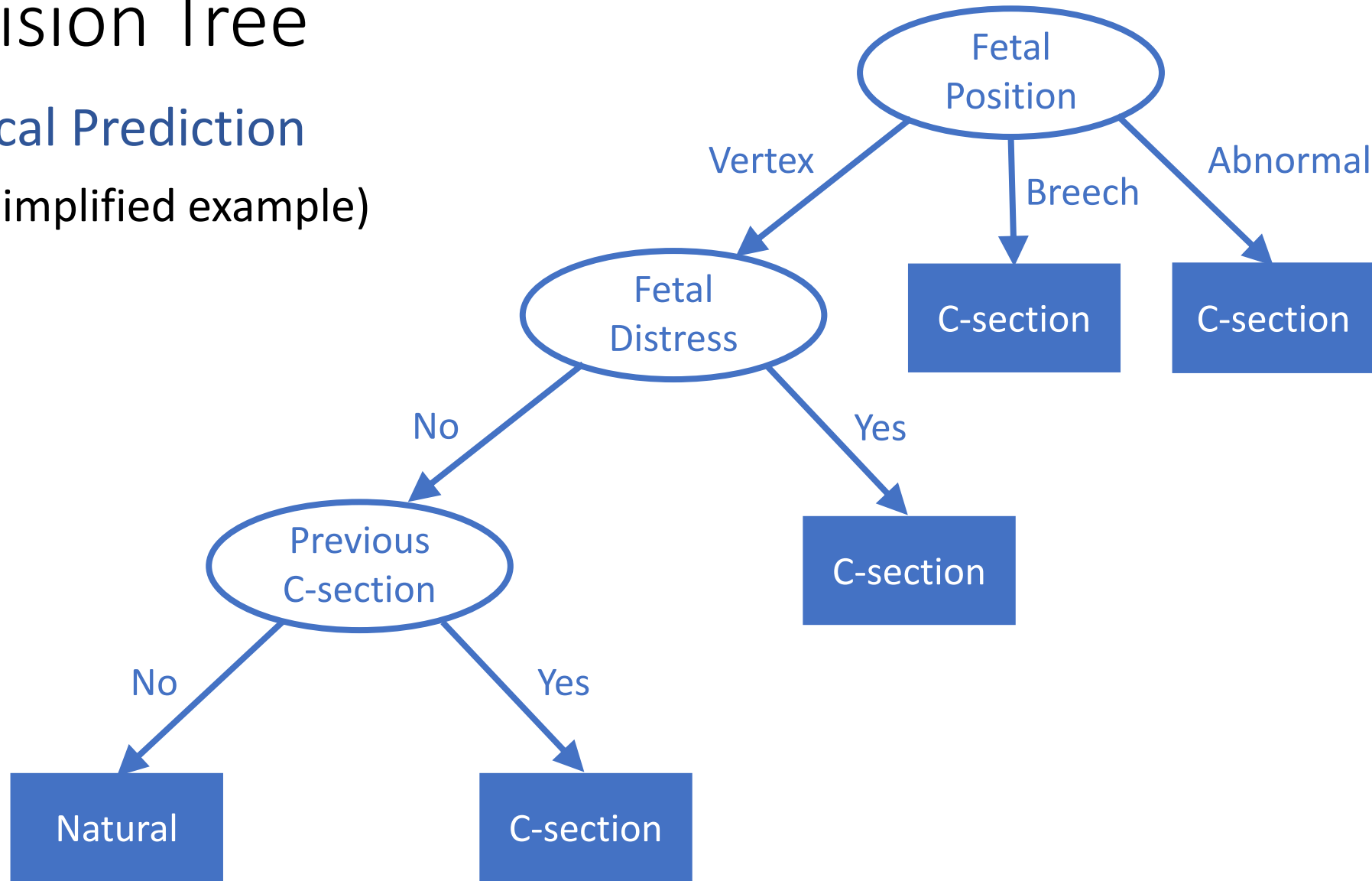
$$y \in \{Csection, Natural\}$$

$$\hat{y} = h(\mathbf{x})$$

Decision Tree

Medical Prediction

(Oversimplified example)



How could we implement training and prediction?

Algorithm 0: Memorization algorithm

Piazza Poll 1

Does the memorization algorithm learn?

A. Yes

B. No

C. I have no clue