

Artificial Neural Network: Orientation & Introduction

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- Introduction & Sweet Treats
- Optimization & Machine Learning
- 2-Layer MLP & Backpropagation
- Selected Topic: Dimension Reduction
- Selected Topic: CNN
- Selected Topic: LSTM

} Tentative

A Rought Intro to Artificial Neural Network

Artificial Neural Network (ANN)

- A computing (math) model
- Prediction: given x , predict (compute) y
- Simply, ANN $\equiv f : x \mapsto y$.
- x is a question, y is an answer, and f is a model
If an answer is an egg, a model is a hen.
If you want good eggs, take good care of a hen.
- They can be “anything”.
E.g., categorical data, numerical data, text, image, audio, vdo, etc.
 \Rightarrow (This is not completely true,
but we learn this complicated fact as we go along.)

This Course (I)

- We won't learn everything. (Nobody can!)
- We will cover the essential.
- We may learn some related topics that are not “ANNs” in the conventional meaning.
- ^{Hand-on} It is about MATH and MATH PROGRAMMING.
- It requires TENACITY, curiosity, creativity, and skills, in this order.

This Course (II)

The Essential

- Optimization. (NOT AN OPTION! IT'S THE ENGINE.)
- Model selection and Evaluation.
- Simple model, e.g., polynomial (basic modeling)
- 2-Layer ANN
- Other models and applications, e.g., PCA, CNN, LSTM, etc.
- Reading an academic paper (object recognition, etc.)
Learn from the source and be able to stand on your own feet

Grading is based on:

Participation in class, assignments, ~~and~~ class project, Ex + Ex.

- Machine Learning (ML): a broader term covers other families of models, e.g., SVM, PCA, LDA, LSTM, CRF, etc.
- Data Science: a fancy term combining mainly ML and techniques, e.g., MapReduce, to handle big data.
- Artificial Intelligence (AI): an even broader term covers various approaches including heuristic search and logic based approaches.
⇒ ‘A’ = Artificial is in the sense of “not biological”.
It does not mean “fake” or “charade”.
AI \neq I can be lazy and stupid and let the machine does the job.

Common Theme (I)

- Predictive model $f : x \mapsto y$
- Answer y for question x .
- We do not find answer y directly.
- We find a good f . \Rightarrow We do not lay eggs y ourselves.
 \Rightarrow Take good care of our hen f .
- f is a tunable function: f has parameters
 - Linear model: $f(x) = mx + c$
 - Cubic model: $f(x) = a_3x^3 + a_2x^2 + a_1x + a_0$
 - Polynomial: $f(x) = a_nx^n + \dots + a_1x + a_0$
 - Perceptron: $f(x_1, x_2) = h(w_1x_1 + w_2x_2 + b)$, where $h(a) = 1$ for $a > 0$ and $h(a) = 0$ otherwise.

Common Theme (II)

- Predictive model $f : x \mapsto y$
- Answer y for question x .
- We do not find answer y directly.
- We find a good f .
- f is a tunable function, e.g., $f(x) = mx + c$.
- Find a good set of parameters $\Rightarrow \sim$ good f .
- “Two-Phase” Strategy
 - Find a good f (\sim find a good set of param's)
 \Rightarrow “Training” / “Learning” (Take care of the hen)
 - Use good f for prediction.
 \Rightarrow “Testing” / “Predicting” / “Using” (Get the eggs)

Numerical Programming I

- Equality comparison

Naive match: $x == y$ vs Tolerance: $y - e < x < y + e$

- Computation sequence, e.g., $y = \frac{a}{b} \cdot \log_{\exp c} \exp b$

Naive vs Refined $\frac{a}{c}$

- $y = \sum_i w_i x_i$

Conventional loop vs Vectorization: $y = \mathbf{w}^T \cdot \mathbf{x}$

- $[\mathbf{y} : y_n = \sum_i w_i x_{i,n}]$

Loop vs Vectorization

- Probabilities

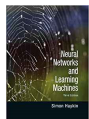
Probability: $\prod_i p_i$ vs Log Probability: $\sum_i \log p_i$

- Softmax $y_k = \frac{\exp(a_k)}{\sum_i \exp(a_i)}$
Straightforward vs $y_k = \frac{\exp(a_k - \max(\mathbf{x}))}{\sum_i \exp(a_i - \max(\mathbf{x}))}$
- Cross-Entropy: $\hat{y} \log y + (1 - \hat{y}) \log(1 - y)$,
where $\hat{y} \in \{0, 1\}$ and $y \in [0, 1]$
 \Rightarrow Naive vs Selective
- Multi-Class Cross-Entropy: $\sum_k \hat{y}_k \log y_k$
Naive vs Selective
- NaN : Not A Number

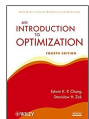
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ธ. กัตัญญกุล, การเรียนรู้ของเครื่องเบื้องต้น. ม. ขอนแก่น 2017



Bruce Lee:
“Don’t pray for an easy life.
Pray for strength to endure the hard one.”