

A Machine Learning Perspective on Predictive Coding with PAQ by Knoll & de Freitas

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December 3, 2019

Overview

- 1 Introduction to PAQ
- 2 PAQ8L
 - Architecture
 - Neural Network
 - Model Mixer
 - Adaptive Probability Maps
- 3 Applications for PAQ8
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Introduction to PAQ

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Introduction to PAQ

What is PAQ8

- What is it?
- How does it work?
- What makes it so famous?

What is PAQ?

- A lossless, open-source compression algorithm
- Brings high performance at the cost of increased memory usage and time consumption
- Related to PPM, is envisioned as PPMs improvement

Principles of PAQ

- Modeling combined with adaptive arithmetic encoding
- Open to additions and improvements
- Improves performance of PPM by including several predictors (i.e. models of data)
- Combines the result of the predictors

Introduction to PAQ

Exemplary Predictors

The order- n context predictor

- Examines the last n bits and counts the 1's and 0's
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Whole word order- n context

- Context is the latest n whole words
- Non-alphabetical characters are ignored and upper- or lower case letters are viewed as the same
- Very useful for text files

PAQ & Predictors

- PAQ encoder looks at the beginning of input file for deciding which predictors are used
- Ways to combine predictions change through with the different versions
- Each predictor outputs a pair of bit counts (n_0, n_1)
- Counts of each predictor are weighted with context length
- Those counts get summed up

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PAQ8 - What's new?

- Predictors don't produce a pair of bit counts anymore
 \hookrightarrow those counts get weighted and normalized into the interval $[0, 1] \subset \mathbb{R}$
- Instead each predictor already outputs a probability
- *paq8l* is a stable version of paq8, released by Matt Mahoney

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PAQ8L - Machine Learning Perspective

- paq8l is the version of PAQ used by *Byron Knoll & Nando de Freitas*
- They try to show the possibilities of PAQ beyond data compression

Architecture of PAQ8

- Uses weighted combination of predictions from Large number of models
- Allows no-contiguous context matches
- paq8l uses **552** prediction models
- Combines the output of them into a single one
 - ↪ Passes this through an *adaptive probability map* (APM) before using the arithmetic coder

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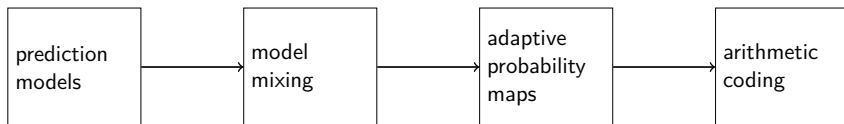


Figure: PAQ8 Architecture

Neural network

Neurons of a neural network

A **neuron** takes one or more **inputs** and gives an **output**.

Within the topic of machine learning, the neuron can be understood as a **function**.

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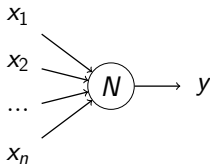


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Layer in neural network

A layer is a group of neurons.

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A neural network

Neural networks is defined by its layers:

- 1 input layer with n inputs
- 1 output layer with k outputs
- M layers between input and output layer (i.e. hidden layers)
- Layers can consist of different amounts of neurons

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General structure of neural network

Let it be an generic neural network with:

- x_1, \dots, x_n inputs and y_1, \dots, y_k outputs
- There are M different layers between input and output

Neural network

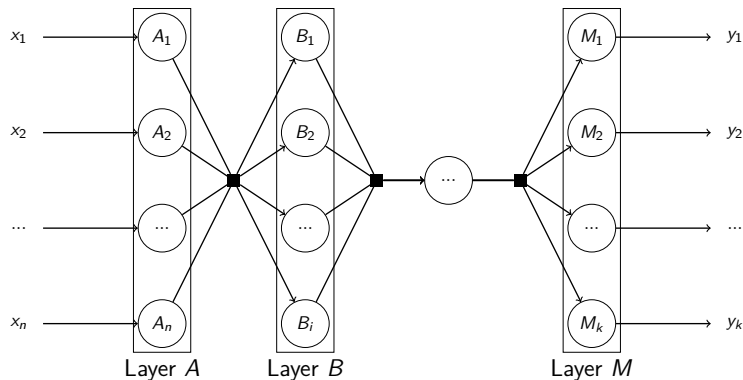


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Model Mixer of paq8l

- Resembles a neural network with one hidden layer
- One hidden layer is between input and output layer
- Subtle differences from a standard neural network

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Differences between paq8l and neural networks

- 1 Weights for first and second layers are learned online and independently for all nodes:
 - Each node trained separately
 - reduces predictive cross-entropy error (unlike back propagation)
- 2 Hidden nodes are partitioned into seven sets

Hidden Node Partitioning

- For every bit of data 1 node from each set
- Only edges of selected nodes are updated
- $552 \times 7 = 3,864$ weights updated per bit

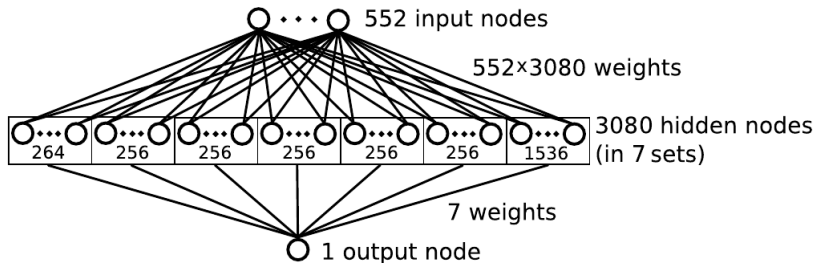


Figure: Model Mixer architecture (Graphic by Knoll & De Freitas)

Node selection

- Sets 1,2,4 and 6 choose node based on single byte in context
↔ 256 nodes per set
- Sets 3 and 7 use combination of several bytes
- Depending on the context, a specific node is selected

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Mixtures of Experts

- Technique published by *Jacobs et al.*
- Used for neural network training
- Requires a gating network to select expert model

Definition APM

- Takes prediction from model mixer
- is an two dimensional table and low order context as input
- Outputs a new prediction on non-linear scale
- Table entries adjusted after each bit is coded

Applications for PAQ8

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Application for PAQ8

What applications?

PAQ8 is useful even beside compressing files.

- Adaptive Text Prediction
- Text categorization
- Shape recognition
- Lossy compression (i.e. JPEG)

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Adaptive Text Prediction

- PAQ8 can be used to find string x for some training string y
- Can be set to work in speech recognition and text prediction (for typing)

Lossy compression

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