

Neural networks and Chinese character recognition

Jeremy Reizenstein

May 2016

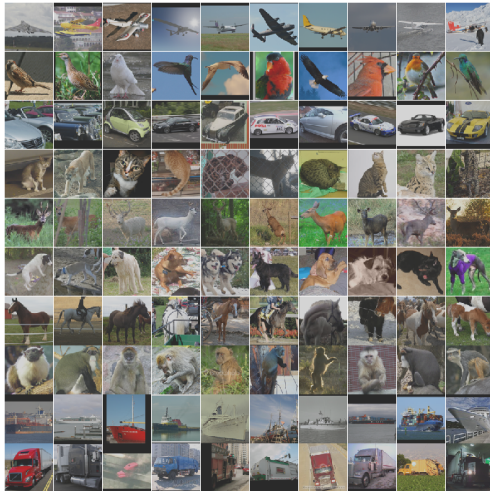
Table of Contents

- 1 Deep learning examples
- 2 Neural networks
- 3 Handwriting recognition
- 4 Signatures

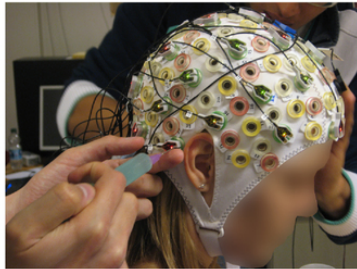
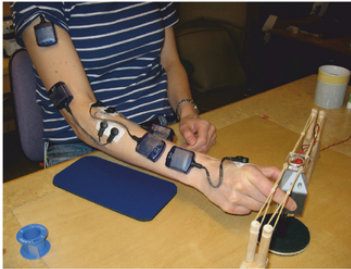
Example: Mnist



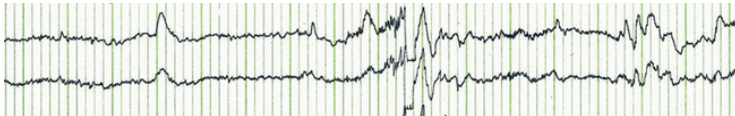
Example: Imagenet



Example: EEG

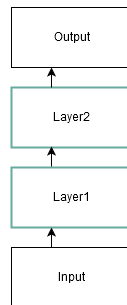
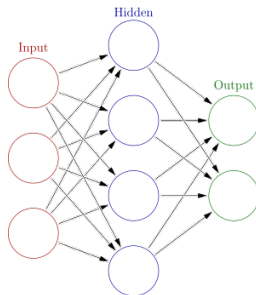


1



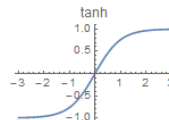
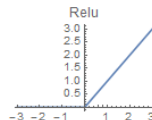
¹Multi-channel EEG recordings during 3,936 grasp and lift trials, Luciw, Jarocka and Edin, Scientific Data 1, 2014; kaggle 2014

Neural Networks



$$H = \sigma(\mathbf{W}_1 I + \mathbf{b}_1)$$

$$O = \sigma(\mathbf{W}_2 H + \mathbf{b}_2)$$



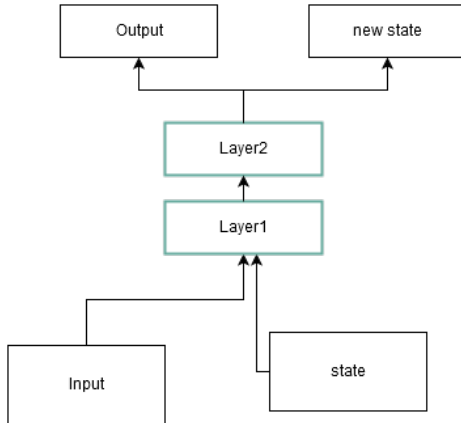
Typical approach to supervised machine learning



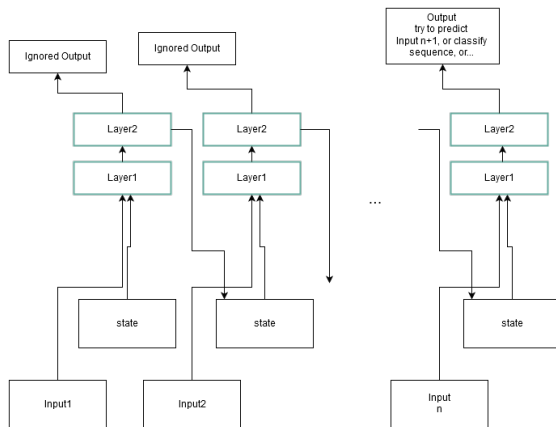
Timeseries and paths

Data which varies in time doesn't live in a fixed \mathbb{R}^k .
Time-structure of the data matters

Recurrent Neural Networks

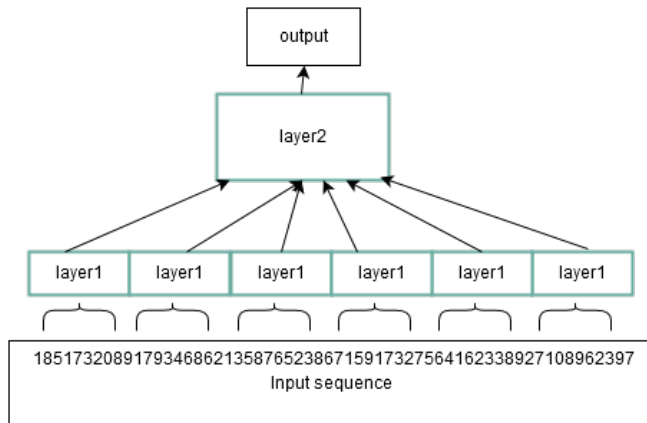


Recurrent Neural Networks



<http://karpathy.github.io/2015/05/21/rnn-effectiveness/>

Convolutional Neural Networks



Chinese - the Casia online dataset

苞胞包褒剥薄雹堡饱宝抱报暴豹鲍爆杯
碑悲卑北斐背贝狈倍狈备惫焙被奔苯本笨
崩绷甬泵蹦迸逼鼻比鄙笔彼碧蓖蔽毕毙毙
币庇痹闭敝弊必辟壁臂避陛鞭边编贬扁便
变卞辨辨辨遍恹彪膘表瞥瞥剔瘰淋斌濒滨
宾揆兵冰柄丙秉饼炳瞭病并玻菠播播钵液
博勃搏铂箔伯帛舶膊膊渤泊驳埔卜哺补埠
不布步簿部怖擦猜裁材才财睬睬采彩菜蔡

Online handwriting



Typical approaches

input $\xrightarrow{\text{representation}}$ \mathbb{R}^k $\xrightarrow{\text{classification}}$ labels

input $\times t$ $\xrightarrow{\text{representation}}$ $\mathbb{R}^{k \times t}$ $\xrightarrow{\text{RNN}}$ labels

Signatures

The signature of a path is a set of iterated integrals.

Consider a path in \mathbb{R}^3 parameterised by the variable t ranging from 0 to 1, given by

$$t \mapsto (f_1(t), f_2(t), f_3(t))$$

Then, for example, the element 2,3 of the signature is

$$\int_0^1 \left[\int_0^t f_2'(s) ds \right] f_3'(t) dt = \int_0^1 \int_0^t df_2(s) df_3(t)$$

and element 2,1,2 of the signature is

$$\int_0^1 \int_0^t \int_0^s df_2(r) df_1(s) df_2(t).$$

Signatures

The m th level of the signature of a path in \mathbb{R}^d is the d^m values of the elements with m integrated integrals. It takes values in $(\mathbb{R}^d)^{\otimes m}$. Given a piecewise linear path, it is easy to compute the first m levels of its signature using a theorem called Chen's identity.

Log-Signature demonstration

There is redundancy in the signature. The log signature is a transformation of the same information as the signature which is not redundant. For example, in \mathbb{R}^2 , the first four levels of the signature look like this

$$(\cdot) + \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} + \begin{pmatrix} \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \end{pmatrix} + \begin{pmatrix} \begin{pmatrix} \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \\ \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \begin{pmatrix} (\cdot) \\ (\cdot) \end{pmatrix} \end{pmatrix}$$

- that is $2 + 4 + 8 + 16 = 30$ numbers while the log signature is only $2 + 1 + 2 + 3 = 8$ numbers.

Questions

A signature is a nice representation of a path of arbitrary length. When is it good enough? What properties of a complicated path can be derived from some levels of its signature? How to balance a representation using more levels of the signature versus chopping the path up and looking at signatures of the chunks?

Thanks!

EPSRC

Engineering and Physical Sciences
Research Council

centre for
complexity
science



Dr Ben Graham