

GOOD FEATURES

ENGINEERING TRICKS

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WiFi : SG-Guest

Problems with Installation? **ASK!**

PLAN OF ACTION

TODAY

- Math (slower)
- Creating Good Features
- Intros and Final Project scope discussion
- Play with Project 01

PLAN OF ACTION

NEXT MONDAY

- Start with project wrap-up
- CNNs - the basics

GOOD FEATURES

- Magic of DL :
 - hand-crafted features not essential
- But : Does no harm to provide something sensible

OVERVIEW

- Numerical inputs
- Categorical inputs
- 'Going deeper'
- Applications

NUMERICAL INPUTS

- Going to multiply by weights
- Helpful to have them similar in size
- Easy choice is identically distributed :
 - $[0 \dots 1]$
 - $[-1 \dots 1]$
 - $\sim \text{NormDist}(\text{mean}=0, \text{sigma}=1)$
- Forcefully rescale them :
 - By known factor (or rule-of-thumb)
 - By learned factor, eg: `BatchNorm()`

NUMERICAL INPUTS

- Also helpful to have them non-collinear :
 - Ideal : Independent, identically distributed (IID)

CATEGORICAL VARIABLES

- True or False
- spring / summer / autumn / winter
- ** or **** reviews
- 2 or 4
- 18 or 34
- country:usa or country:singapore
- country x age

BOOLEAN

- False \rightarrow 0, True \rightarrow 1

or

- False \rightarrow 10, True \rightarrow 01

ONE-HOT

- spring → 1000
- summer → 0100
- autumn → 0010
- winter → 0001

ORDERED

- '*' → 0000
- '**' → 1000
- '***' → 1100
- '****' → 1110
- '*****' → 1111

LINEAR-LIKE

- Use as-is; or
- $3 \rightarrow 000100$ (one-hot, sparse)
- $3 \rightarrow 111100$ (as before)

FLOAT-LIKE

- Bucketing : convert to categorical
 - 0-12 → 1000000
 - 12-15 → 0100000
 - 15-18 → 0010000
 - 18-21 → 0001000
 - 21-25 → 0000100
 - 25-99 → 0000001
- ... and use original too

MANY CLASSES

- one-hot is Ok
 - eg: 100s
- But are you throwing away known structure?

VERY MANY CLASSES

- Hashing trick (for, say, 1000 buckets):
 - Convert label into a 'hash'
 - ... something like MD5sum
 - One-hot category is ($\text{hash} \% 1000$)
- 'infinitely extensible'
- Is collision a problem?

COMBINATIONS OF CATEGORICALS

- Apply hashing trick to cojoined labels :
 - country x age →
 $\text{hash}(\text{"country:usa~age:18-21"}) \% 1000$
- Again : 'infinitely extensible'
- Can store lots of coincidence information 'linearly'

NOW WHAT?

- Each feature position ...
 - ... contributes (or not) the corresponding weight to several nodes in the next layer
- Can't do many layers :
 - 1000 categories \Rightarrow ~1MM parameters per dense layer

EMBEDDING

- Instead of $\{0,1\}$ in a single location (sparse):
 - Use location as index into 'dictionary'
 - That location has a learnable vector ($\sim 8d$)
 - Use the vector as inputs to next layer
- Can learn arbitrary input features
- All trainable via Backpropagation

WHO CARES?

- Applications would have to have :
 - Lots of kinds of data
 - Lots of interactions between factors
 - Lots of data
 - Need for speed
- Suggestions?

RECOMMENDATION

- Scoring of different recommendations
- Presentation :
 - Most likely next action
 - Second-best alternative
 - People-like-you also-like ...

ADVERTISING

- Predict click-through rates
- Intense area of study
- Lots of this is not published, though

PUBLISHED RESULTS

- Recommender system (Google Play store) :
 - [Deep & Wide networks](#) (June-2016)
 - ... implementations available
- Ad Click Predictions (Google) :
 - [Deep & Cross networks](#) (August-2017)
 - ... too new

FURTHER READING

- Wide & Deep :
 - [TensorFlow tutorial](#)
 - [Kitchen-sink notebook](#)
 - [Presentation](#) with even more stuff (up to TFserving)
 - [Keras code](#) (but advanced feature creation is 'TODO')
- Cross & Deep :
 - ... your company?
 - ... your project?

- QUESTIONS -

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