

Intro to Recurrent Neural Networks

Tue 13th November 2018 5pm - 6:30pm

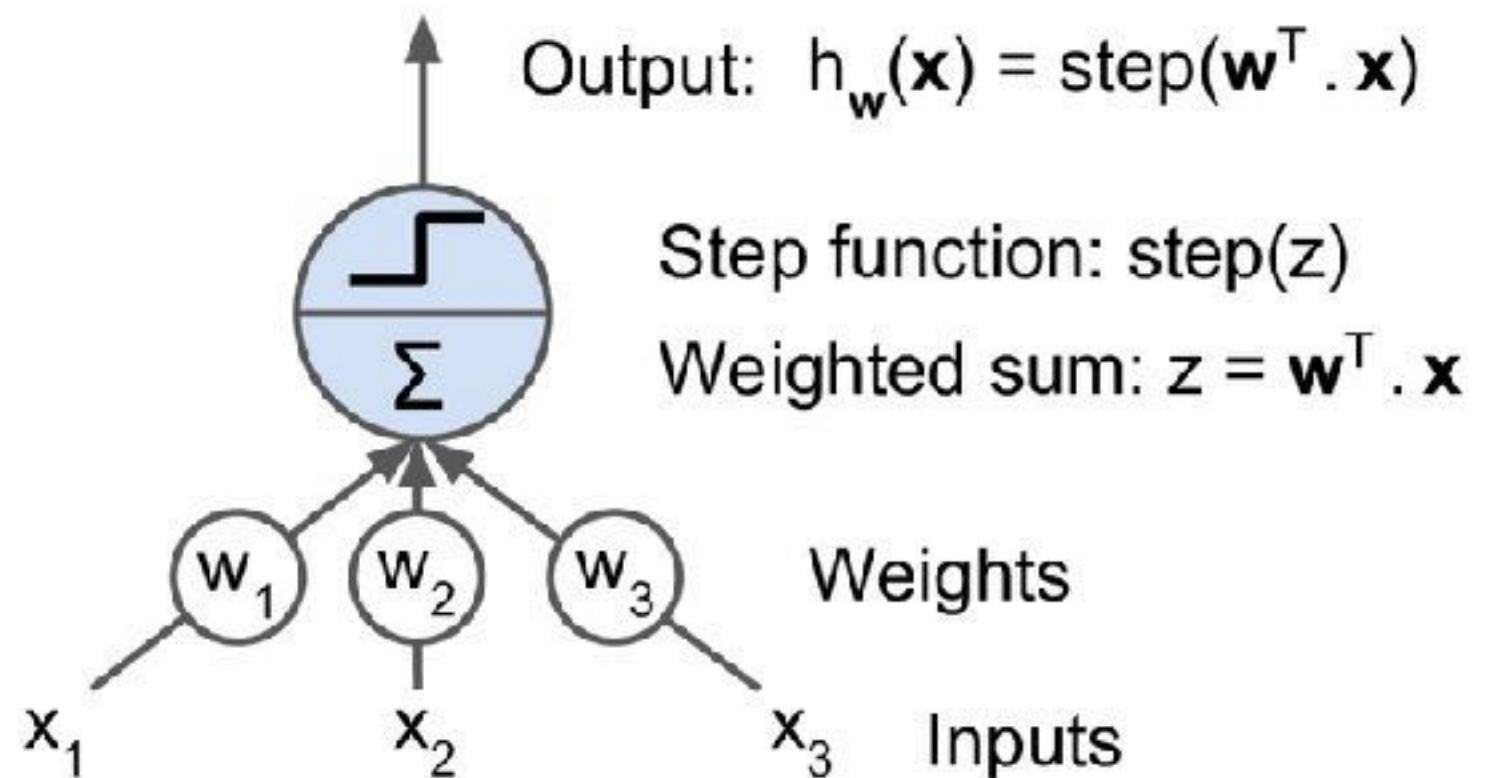


RECAP:
ARTIFICIAL
NEURAL NETWORKS

The Linear Threshold Unit

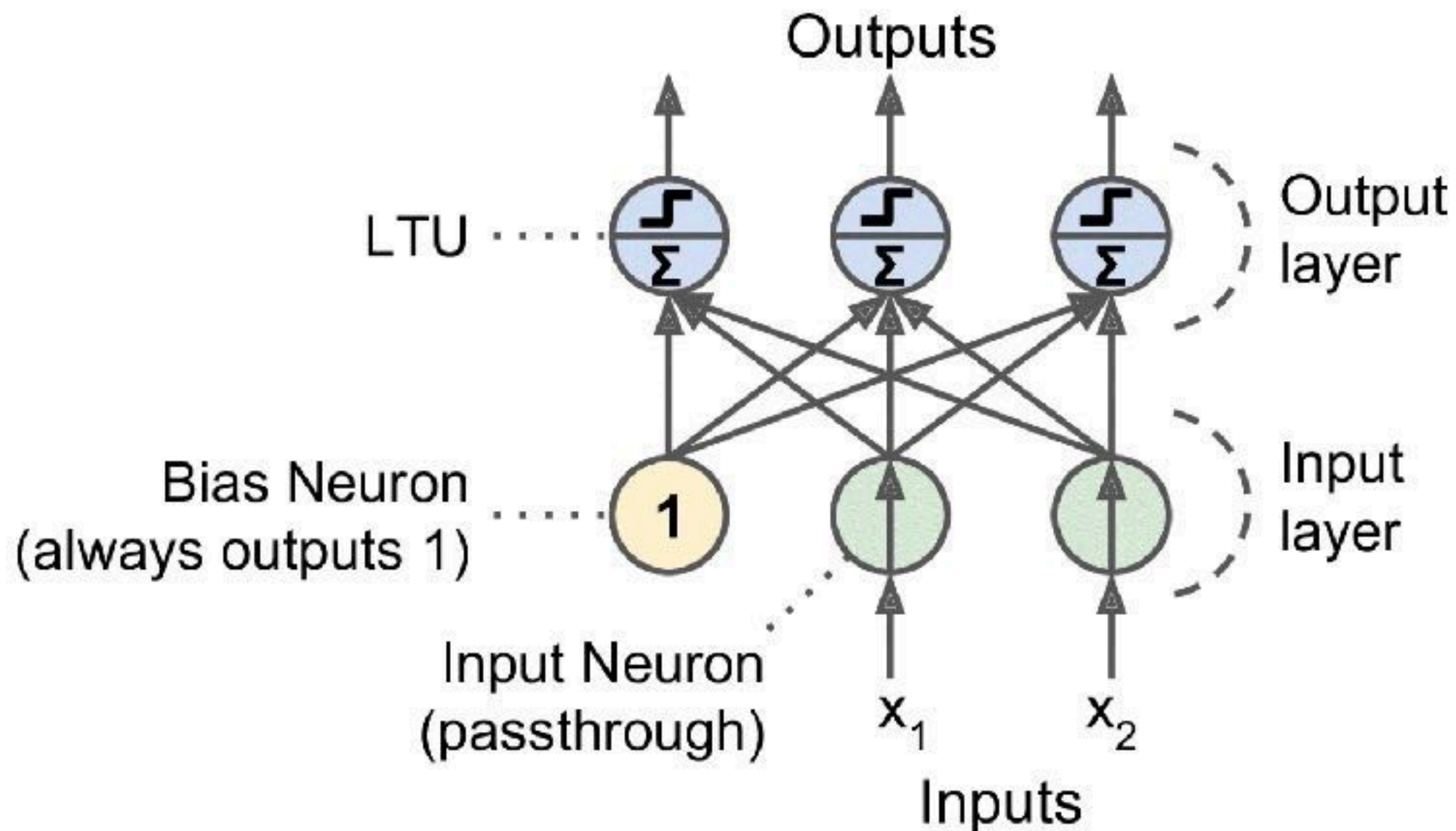
- ▶ 1. Compute **weighted sum of inputs** (linear, i.e. dot product)
- ▶ 2. Apply **step function** (e.g. *heaviside* or *sign*)

$$\text{heaviside}(z) = \begin{cases} 0 & \text{if } z < 0 \\ 1 & \text{if } z \geq 0 \end{cases}$$



The Perceptron

- ▶ A **Perceptron** is simply a **layer of LTUs** plus a **bias** term.
- ▶ Every input feature flows into every LTU.
- ▶ The Perceptron shown below can model three-class binary classification problems.

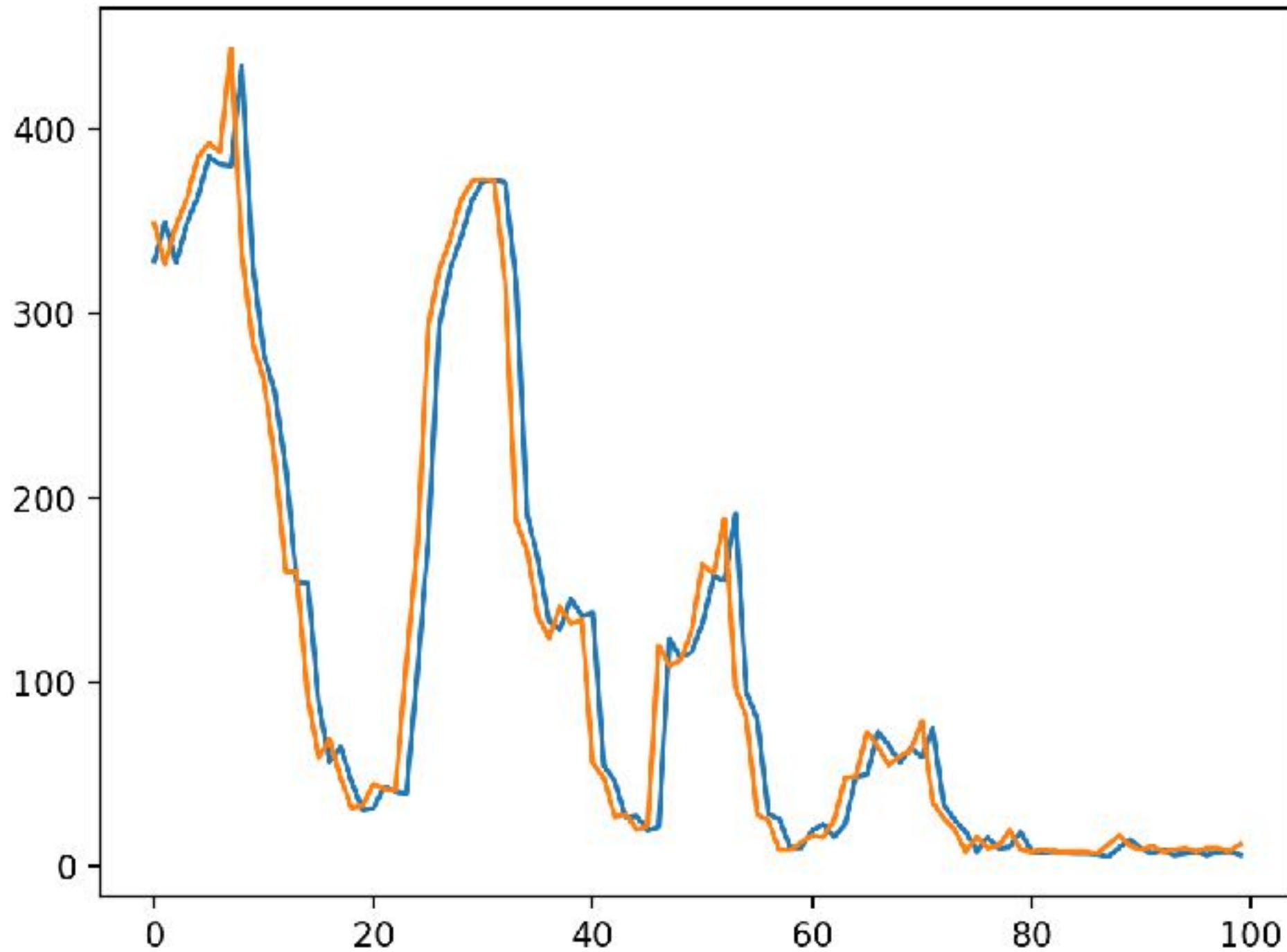


RECURRENT NEURAL NETWORKS

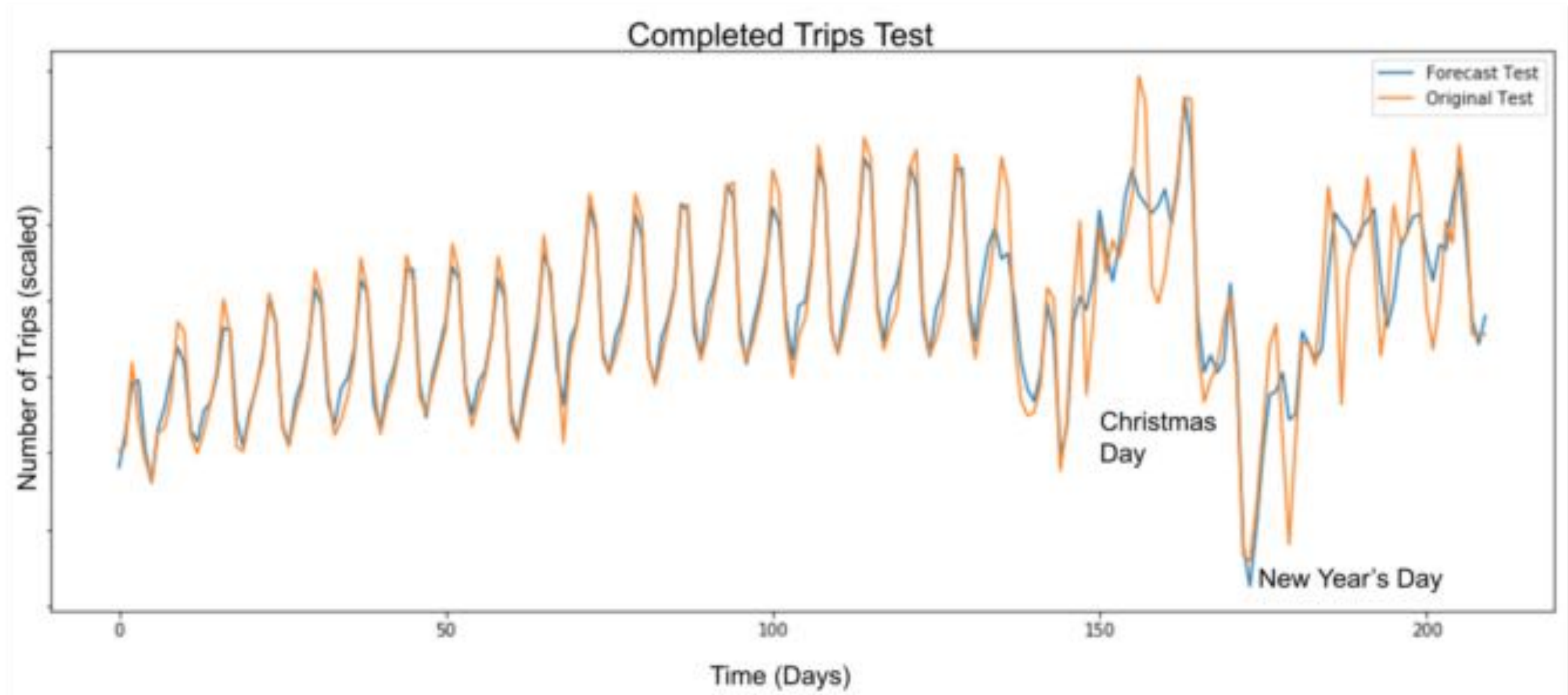
Recurrent Neural Networks

- ▶ Predict the future!
- ▶ Forecast stock prices, anticipate car trajectories in autonomous vehicles, translate words into other languages.
- ▶ Many applications in **time series** and **NLP** (natural language processing), e.g. machine translation, speech-to-text, sentiment analysis and more!
- ▶ **Creative applications:** generating words, sentences, music, knitting patterns, image captions...

Applications: Time Series Forecasting



Applications: Time Series Forecasting



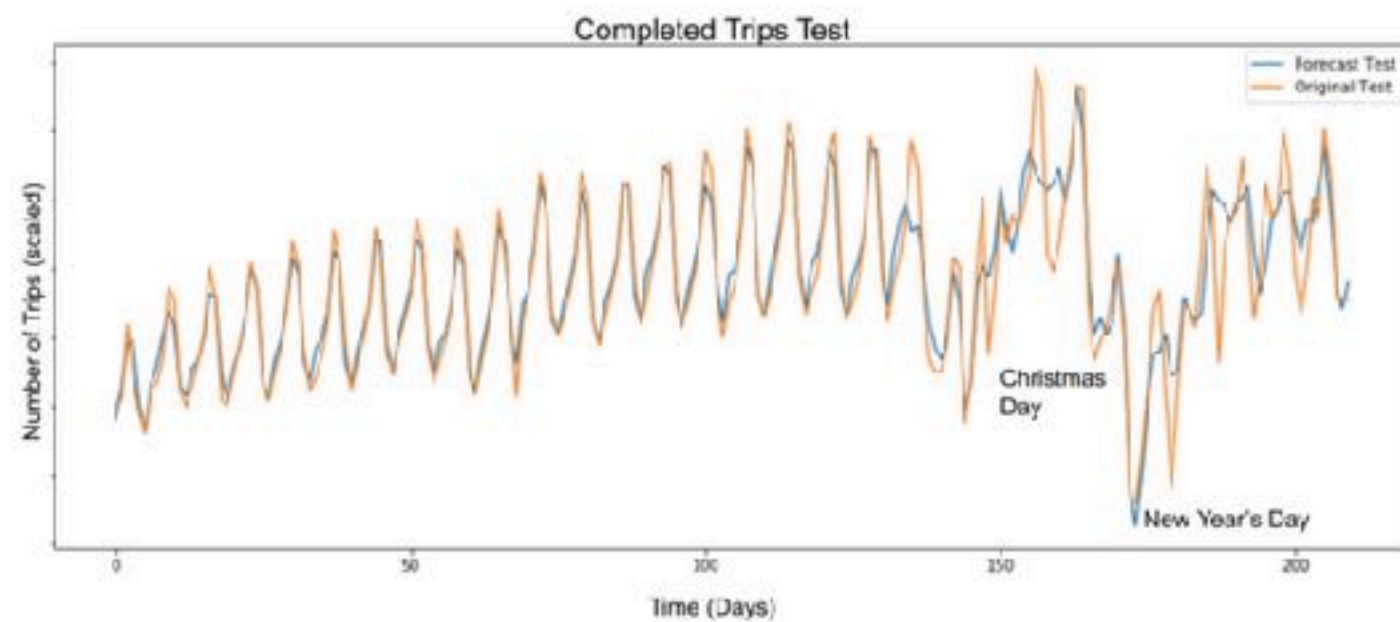
Applications: Time Series Forecasting

Uber Engineering

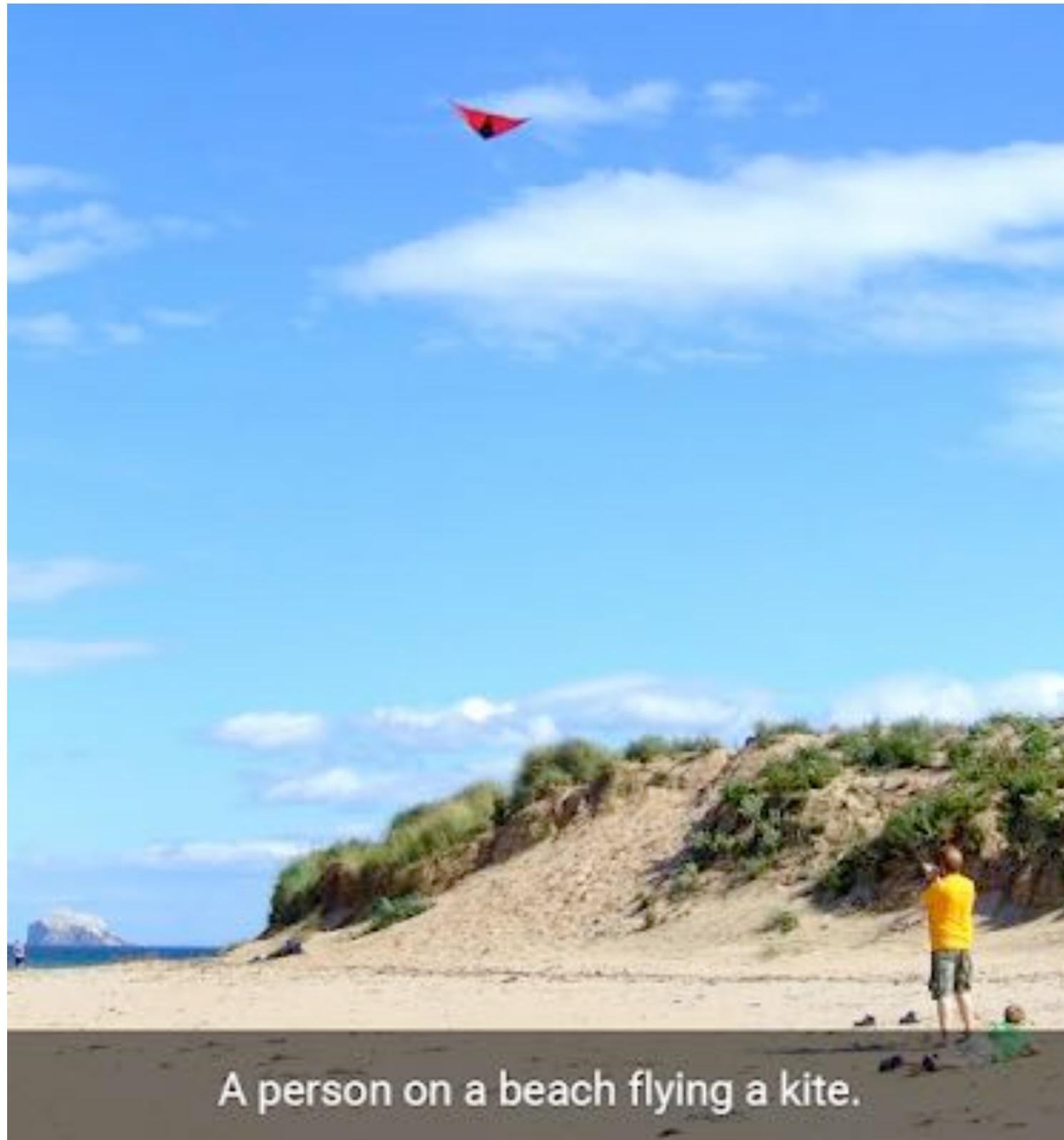
Uber Data

Engineering Extreme Event Forecasting at Uber with Recurrent Neural Networks

June 9, 2017



Applications: Image Captioning



A person on a beach flying a kite.

Applications: Text Summarisation

▶ Original text

- ▶ Alice and Bob took the train to visit the zoo.
They saw a baby giraffe, a lion, and a flock of
colourful tropical birds.

▶ Summary

- ▶ Alice and Bob visited the zoo and saw animals
and birds.

Applications: Generating English Towns



<https://medium.com/@hondanhon/i-trained-a-neural-net-to-generate-british-placenames-9460e907e4e9>

Applications: Generating Shakespeare

PANDARUS:

Alas, I think he shall be come approached and the day
When little strain would be attain'd into being never
fed,

And who is but a chain and subjects of his death,
I should not sleep.

Second Senator:

They are away this miseries, produced upon my soul,
Breaking and strongly should be buried, when I perish
The earth and thoughts of many states.

Applications: Generating Knitting Patterns



Janelle Shane

@JanelleCShane

Follow



So I'm doing a project where I'm training a neural network to generate new knitting patterns. It's going very well. This is "Mystery lace", test-knit by DataSock.

ravelry.com/discuss/lazy-s ...

I like the mystery is tentacles!



ly is sort of seed stitches with some random holes:



6:58 PM - 17 Feb 2018

1,667 Retweets 4,489 Likes



124

1.7K



4.5K



Applications: Generating Knitting Patterns



Janelle Shane @JanelleCShane · Feb 17

Nobody attempted to test-knit this one, for some reason.

Waist Row RS: K1. Row 1 RS: P1, k2tog, yo twice more times. 10 sts). Rows 3, 4, 6, 3, 4, 4 times.
Row 2 (RS): K1, work in pattern to end. Row 3 RS: Work in pattern to end.
Row 4 RS: K1, M1, k1, p1, repeat from * to end.
Repeat these 2 rows in pattern.
Rows 3 and 8 WS: Work in pattern as set. Row 8 WS: P1. 6395, 71, 70, 77 sts.
Row 3 WS: K across. Row 4 (WS): K1, p1, work in pattern to end. Rows 4 and 5 WS: K1, (k1, p1] to-end.
Row 4 (WS): K2tog, yo, k to last st, k1. 5797, 73. Row 3 WS: P1, k5. Row 3 (WS): K3, p1, k1, p1. 7752, 55, 164, 146 sts.

Applications: Generating Knitting Patterns



Janelle Shane

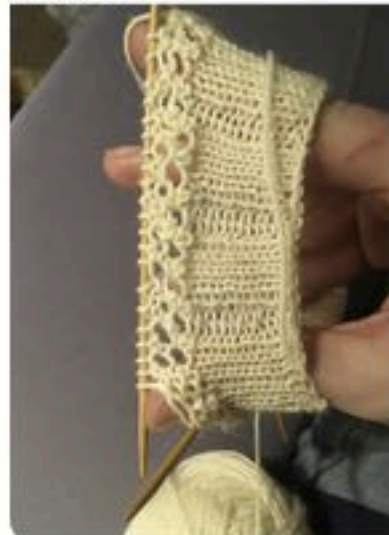
@JanelleCShane

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I learned knitters are amazing at debugging, used to fixing all sorts of pattern problems. They took most of the neural net's mistakes in stride.

Test-knitting: CorvusAlatus [ravelry.com/discuss/lazy-s ...](https://www.ravelry.com/discuss/lazy-s)

so I'm playing with the Winder Socks "pattern", and it has potential. It also has problems. I cast on 73 stitches, worked the row 1 ribbing for a bit before starting with the purl and mesh sections. It's 2x2 ribbing as written, but 4x2, with the exception of where the beginning/end of the round would be. I started the mesh, and...either I've screwed up, or it's naturally increasing by 1 each round. I have 31 stitches currently. - So, it could potentially be used (with some modifications/additional instructions) to actually knit a sock.



*2, repeat from to
rl.*

*o, ssk, repeat from
2tog, yo, k1, repe*

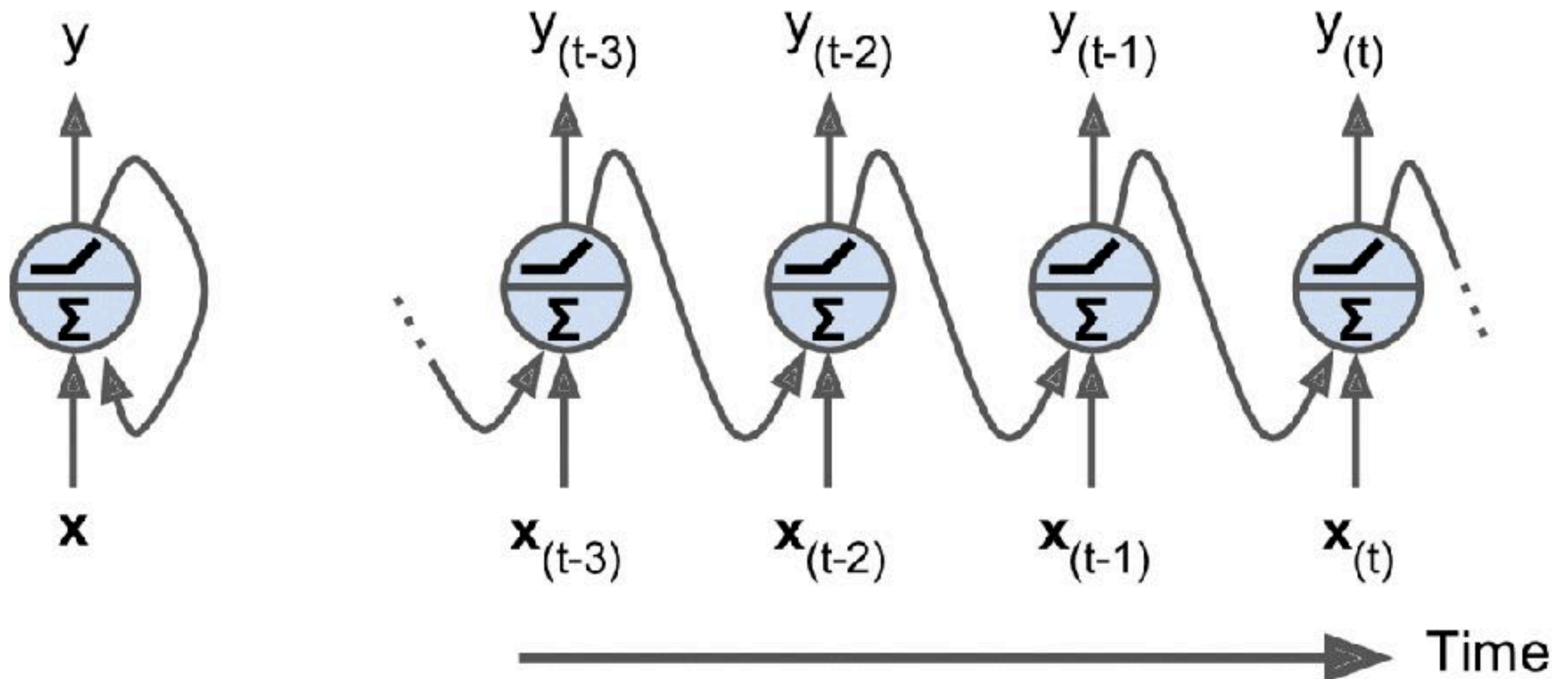
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58 Retweets 313 Likes



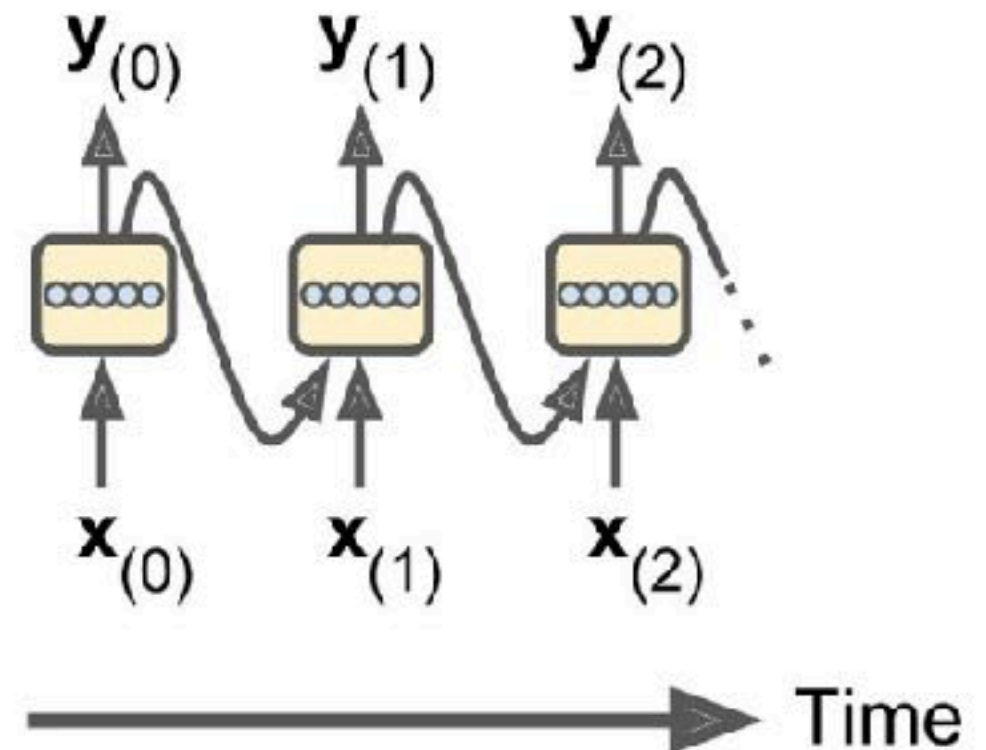
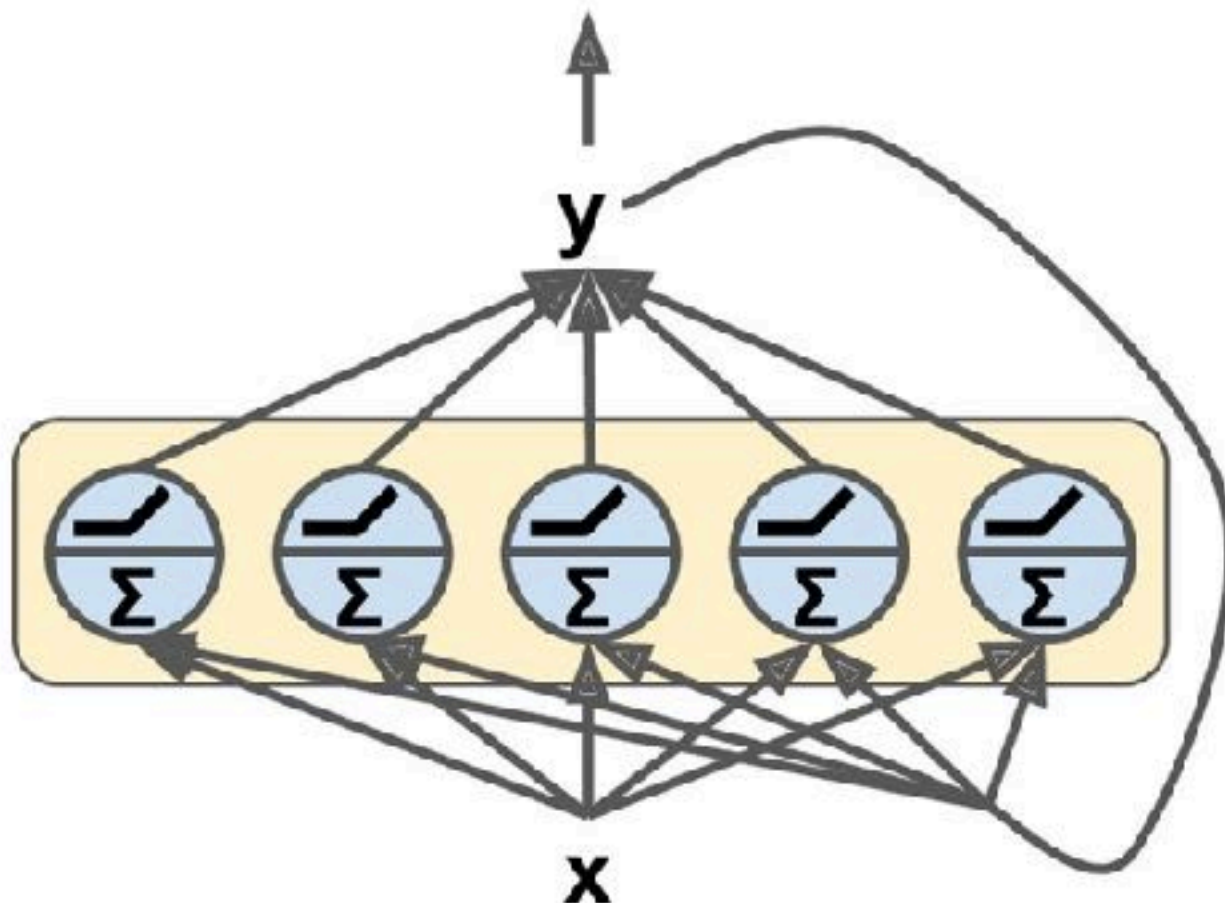
Architecture of RNNs

- ▶ RNNs are neurons (e.g. LTUs) that feed back into themselves.
- ▶ Simple RNN
- ▶ For a time series, we have multiple \mathbf{x}_t over time, also outputs \mathbf{y}_t
- ▶ Here's what the network looks like "**unrolled through time**"



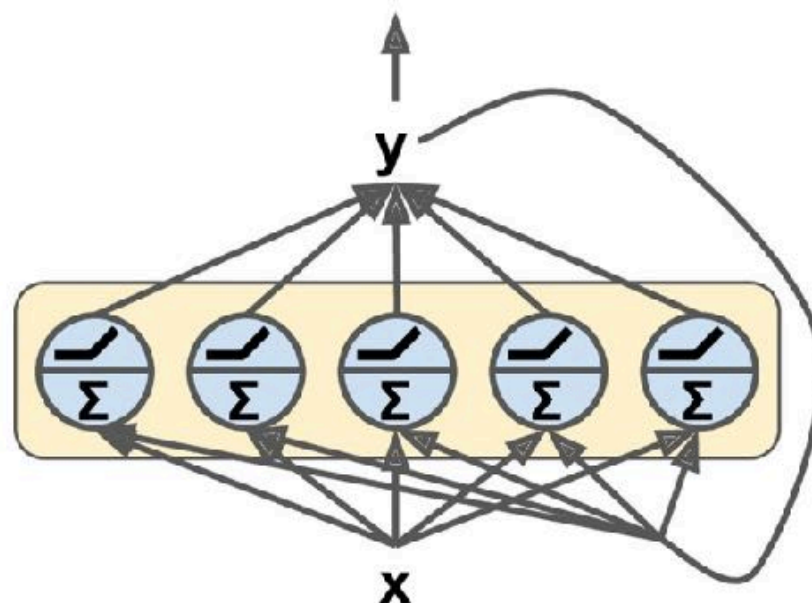
Cells are layers of recurrent neurons

- ▶ A **cell** contains multiple recurrent neurons.
- ▶ Every neuron receives same input (one \mathbf{x}_t and one \mathbf{y}_{t-1})
- ▶ Every neuron generates an output.
- ▶ Combine all the neurons' output to form \mathbf{y}_t
- ▶ We can also unroll the cell through time.



What are the weights?

- ▶ Our input \mathbf{x}_t can be a vector of multiple features (e.g. using AAPL stock price and S&P500 index gives two features).
- ▶ Our output \mathbf{y}_t depends on how many neurons are in the cell, e.g. 3 neurons gives a vector of length three.
- ▶ Neurons in the cell have following weights:
 - ▶ \mathbf{W}_x links 2-feature \mathbf{x}_t input to 3 neurons, i.e. has shape (2,3)
 - ▶ \mathbf{W}_y links 3-neuron \mathbf{y}_{t-1} input to 3 neurons, i.e. has shape (3,3)
 - ▶ Like LTU, the cell also has a bias vector of shape (3,)





LAB:

BUILD A RECURRENT NEURAL NETWORK