

Unit 1: Simple Neural Networks

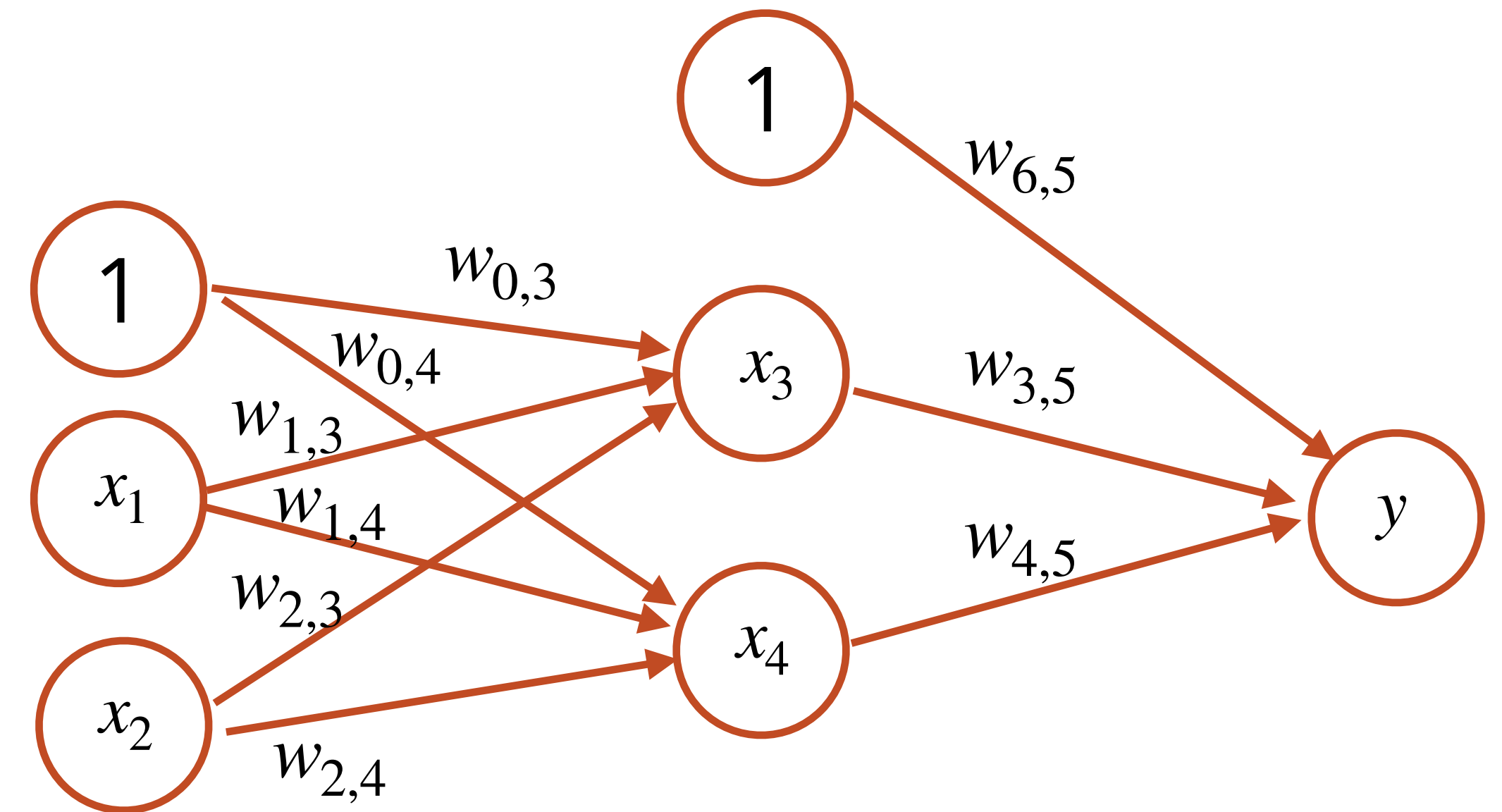
9. Recurrent neural networks

9/29/2020

- 1. Recurrent neural networks can discover structure in time**
- 2. Connectionism recap**
- 3. Voting resources**
- 4. Homework 2**

What these networks can do

Networks like this one can solve problems where there is structure in co-occurrence



With a little modification, they
can also find structure in space
(as you'll see in the Homework 2)

[illegible]

What about structure in time?

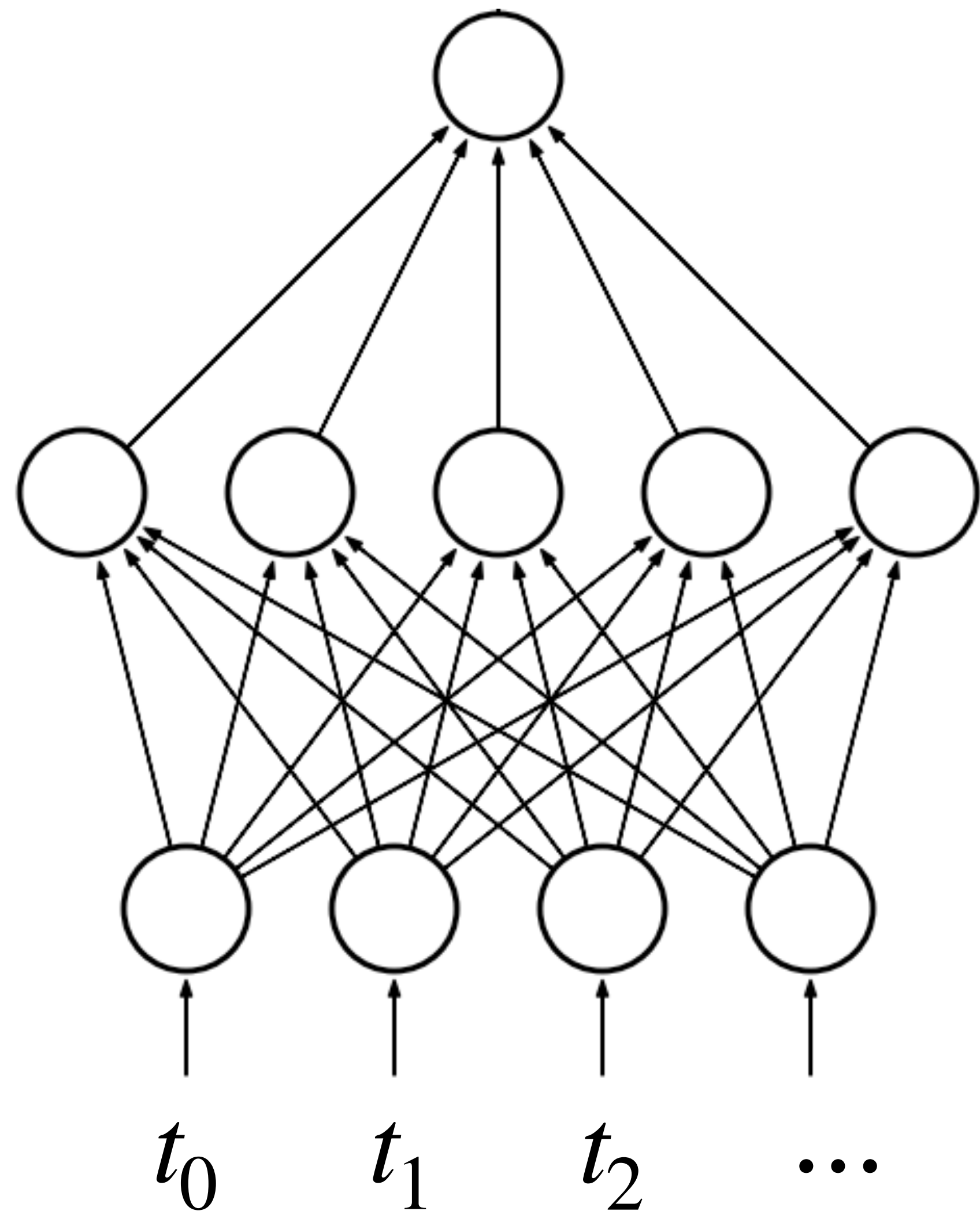
Many of the things people learn, and we want machines to learn are about **structure in time**

From our affinity diagram:

- Learning a song
- Learning to knit
- Playing video games
- Learning a dance routine
- Driving
- Cooking
- ...

How would we represent structure in a network?

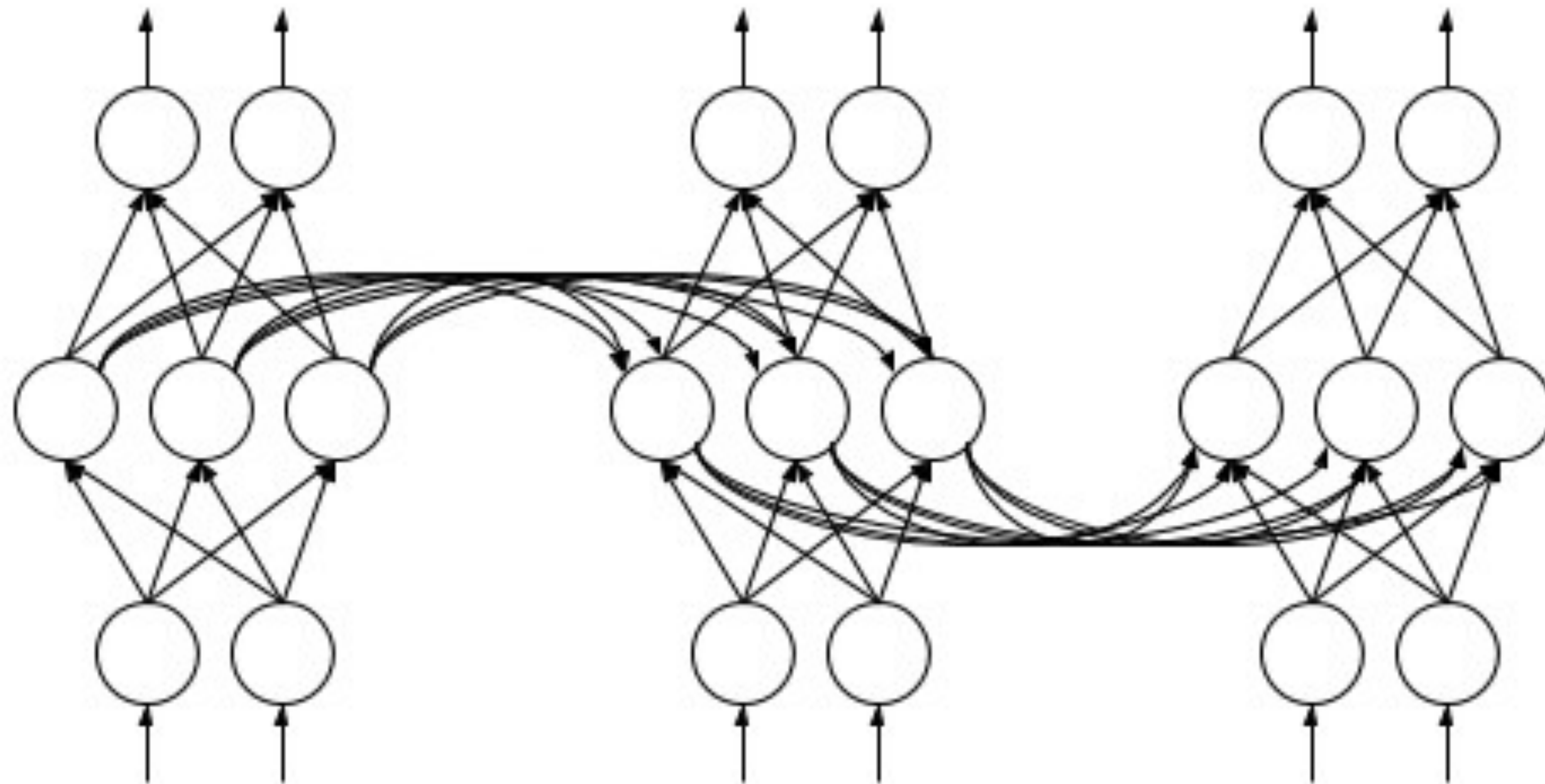
Naive approach: Time as space



Problems:

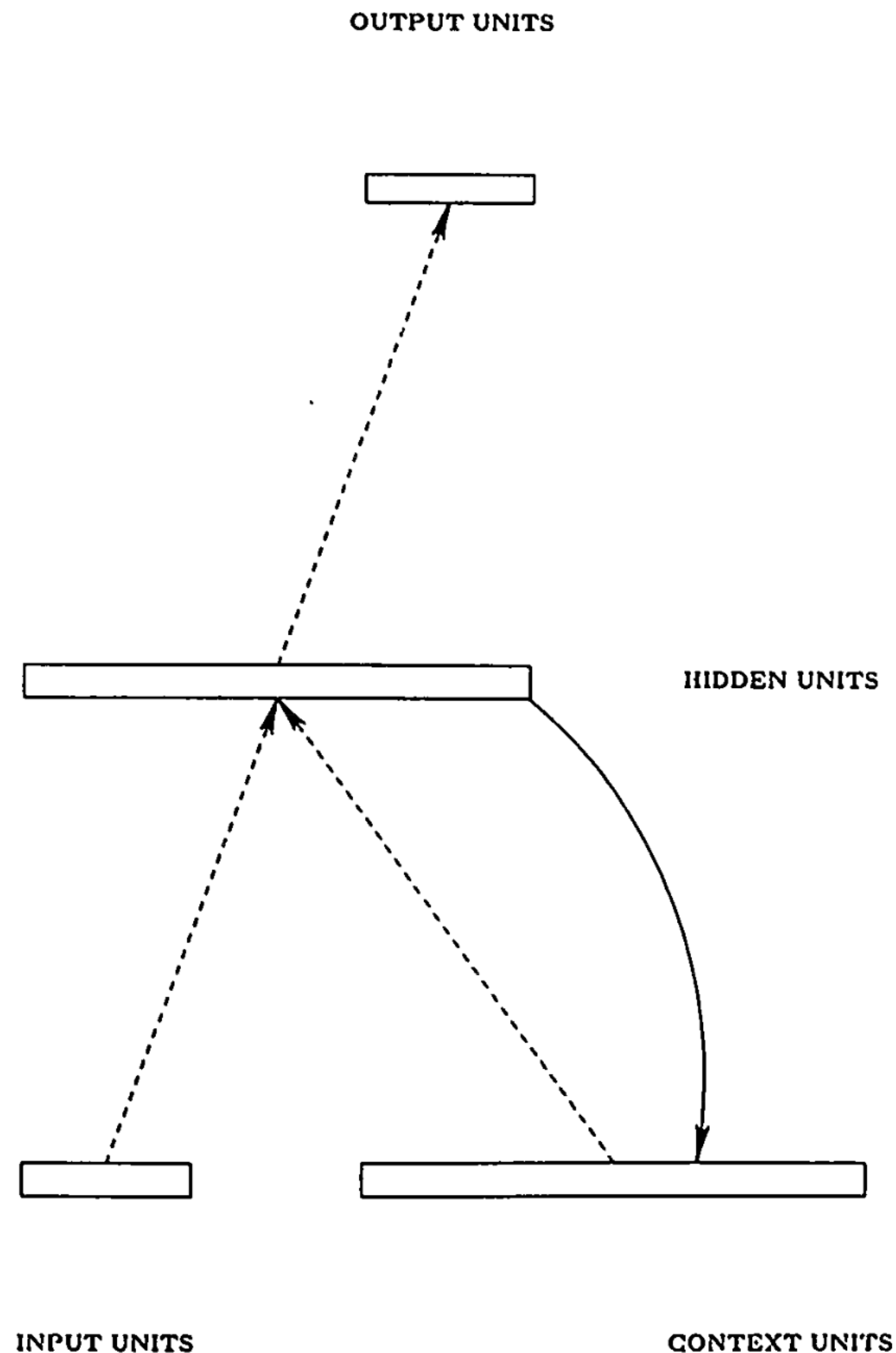
- How big do you make the buffer?
- Two identical patterns translated in time have no natural overlap, e.g. **[0 1 1 0 0 0]** and **[0 0 0 1 1 0]**

Recurrent Neural Networks



Time

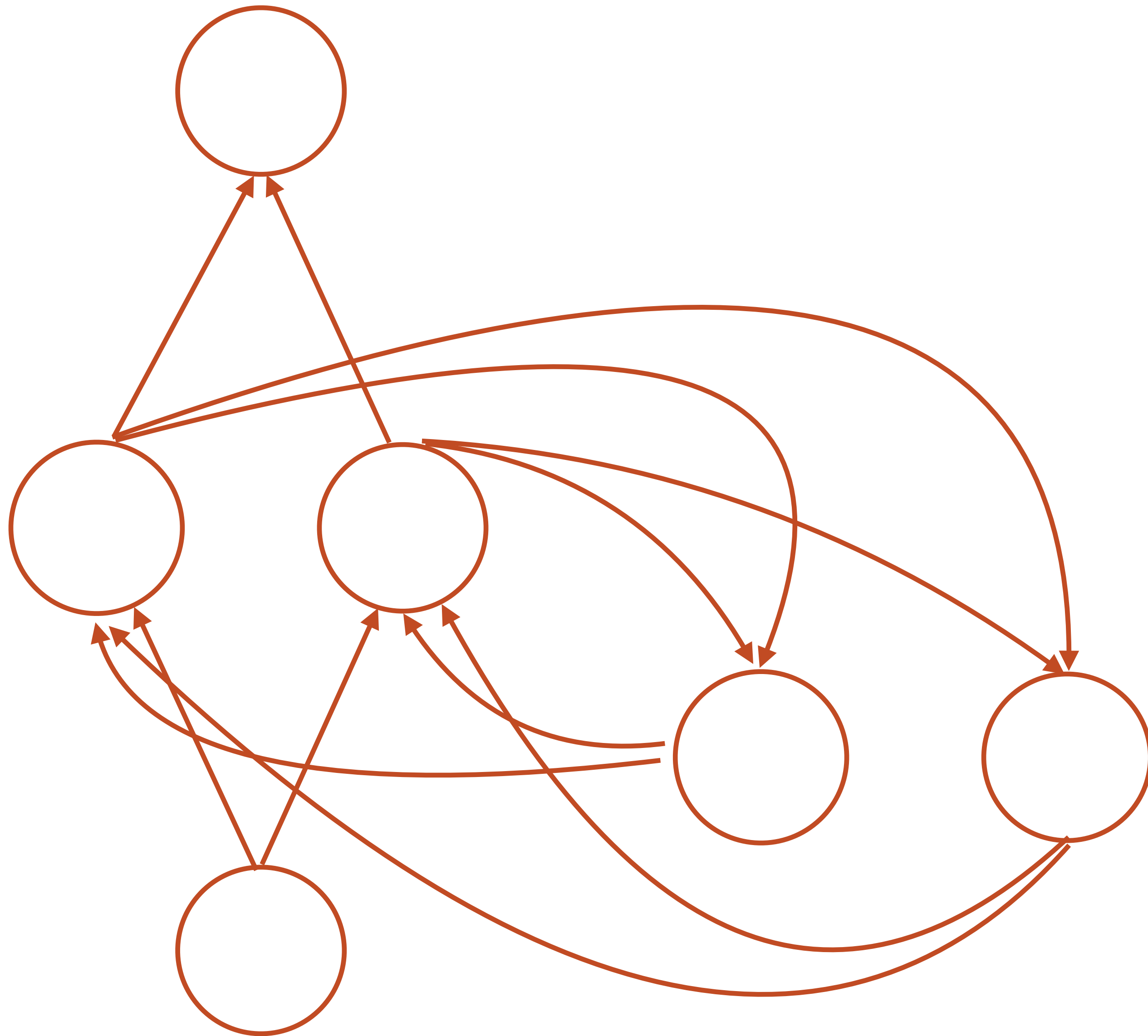
Simple Recurrent Networks (Elman Networks - Elman, 1990)



A set of **context units** that are an exact copy of the hidden layer at $t - 1$

The hidden layer at time t gets input from both the **input units** and the **context units**

XOR in an Elman network



Output

0 1 1 0 0 0 1 1 1 1 0 0 0 1 ?

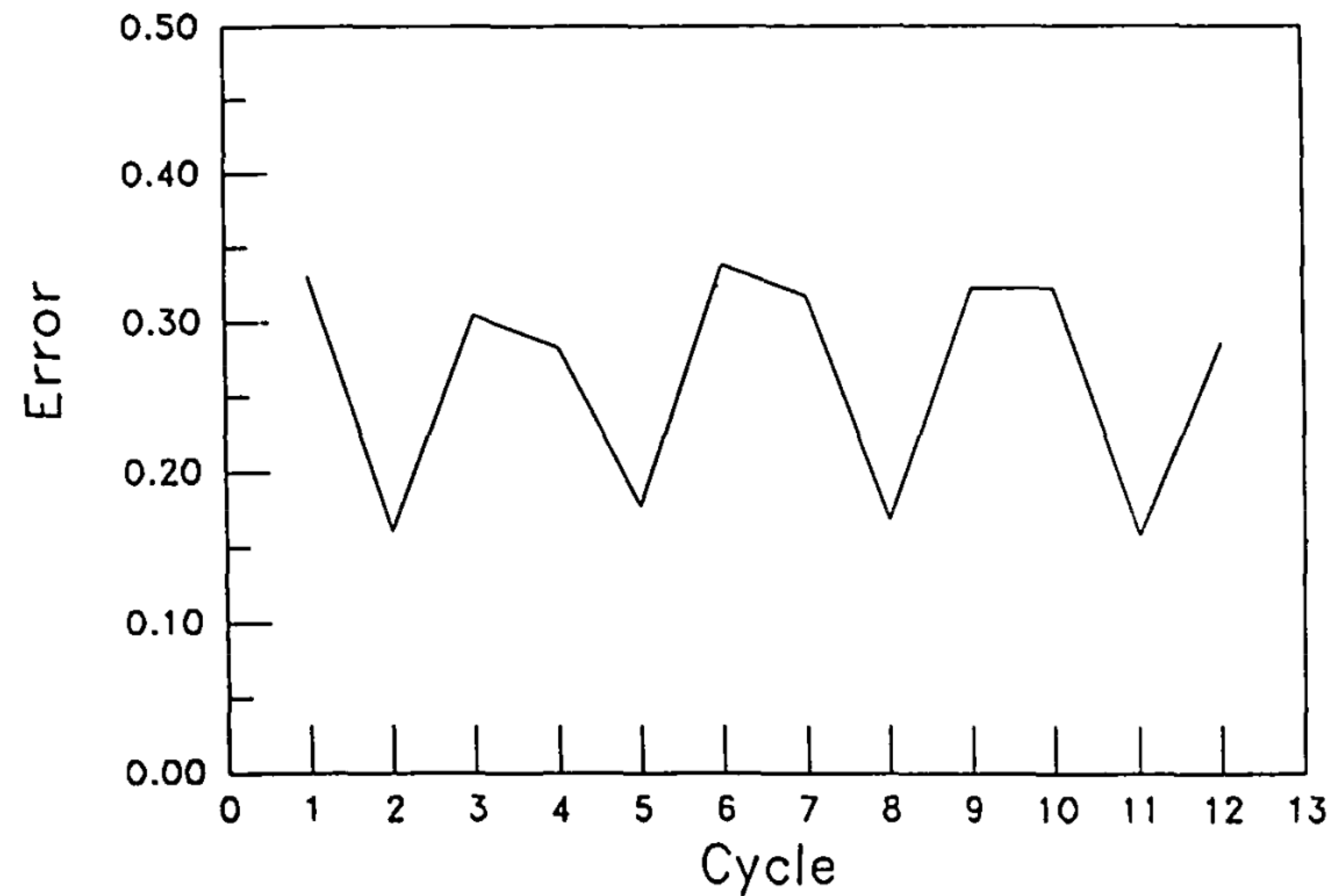
1 0 1 0 0 0 0 1 1 1 1 0 1 0 1...

Input

Goal: Output is XOR of
previous 2 inputs

Error after training

What is happening here?



Output

0 1 0 0 0 0 1 1 1 1 0 1 0 1 ?

1 0 1 0 0 0 0 1 1 1 1 0 1 0 1...

Input

How far back does the network remember?

Input

A random concatenation of the words

ba, dii, and guuu

badiibaguuubadiguuguudiba ...

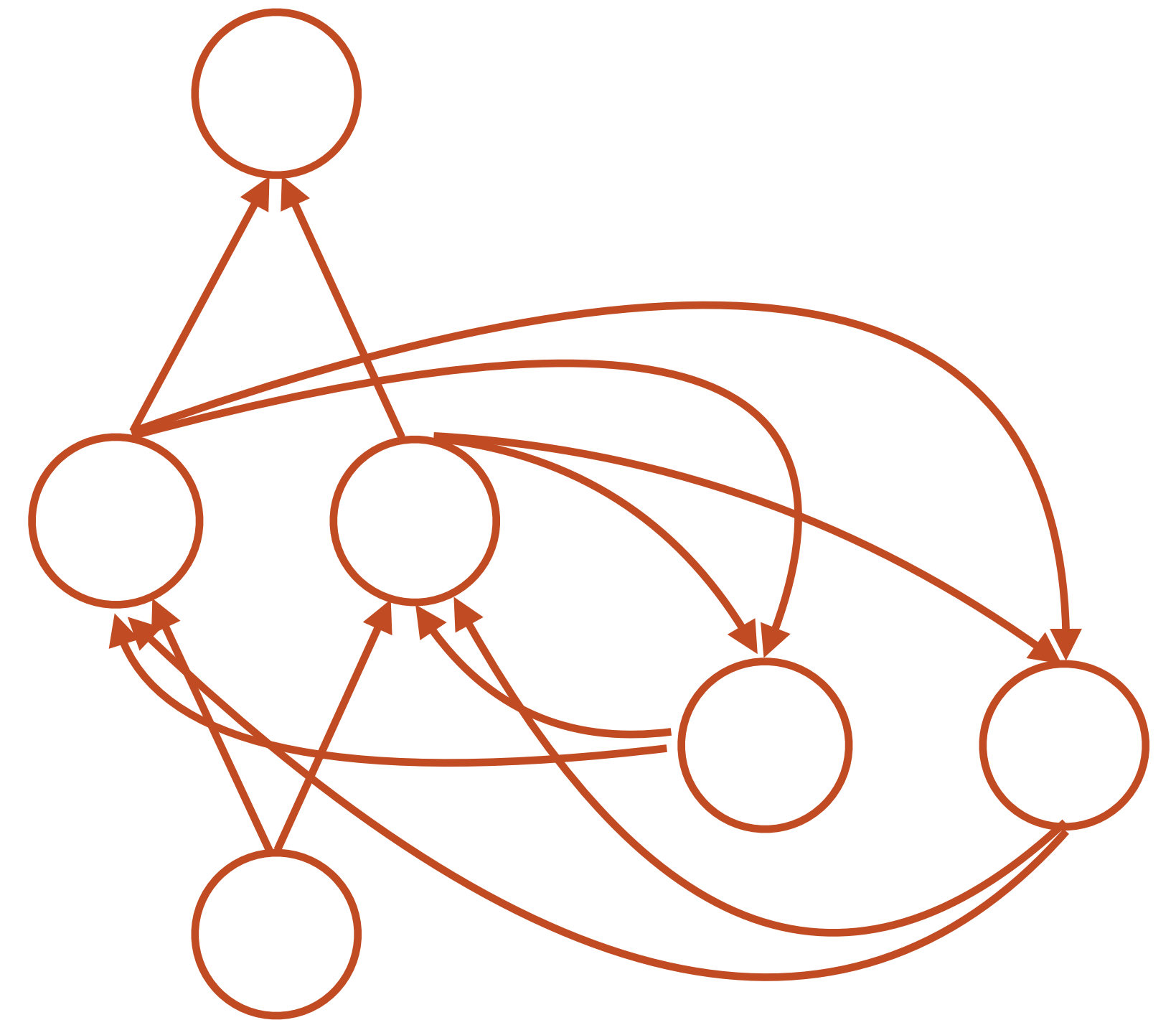
Output

The next letter in the sequence

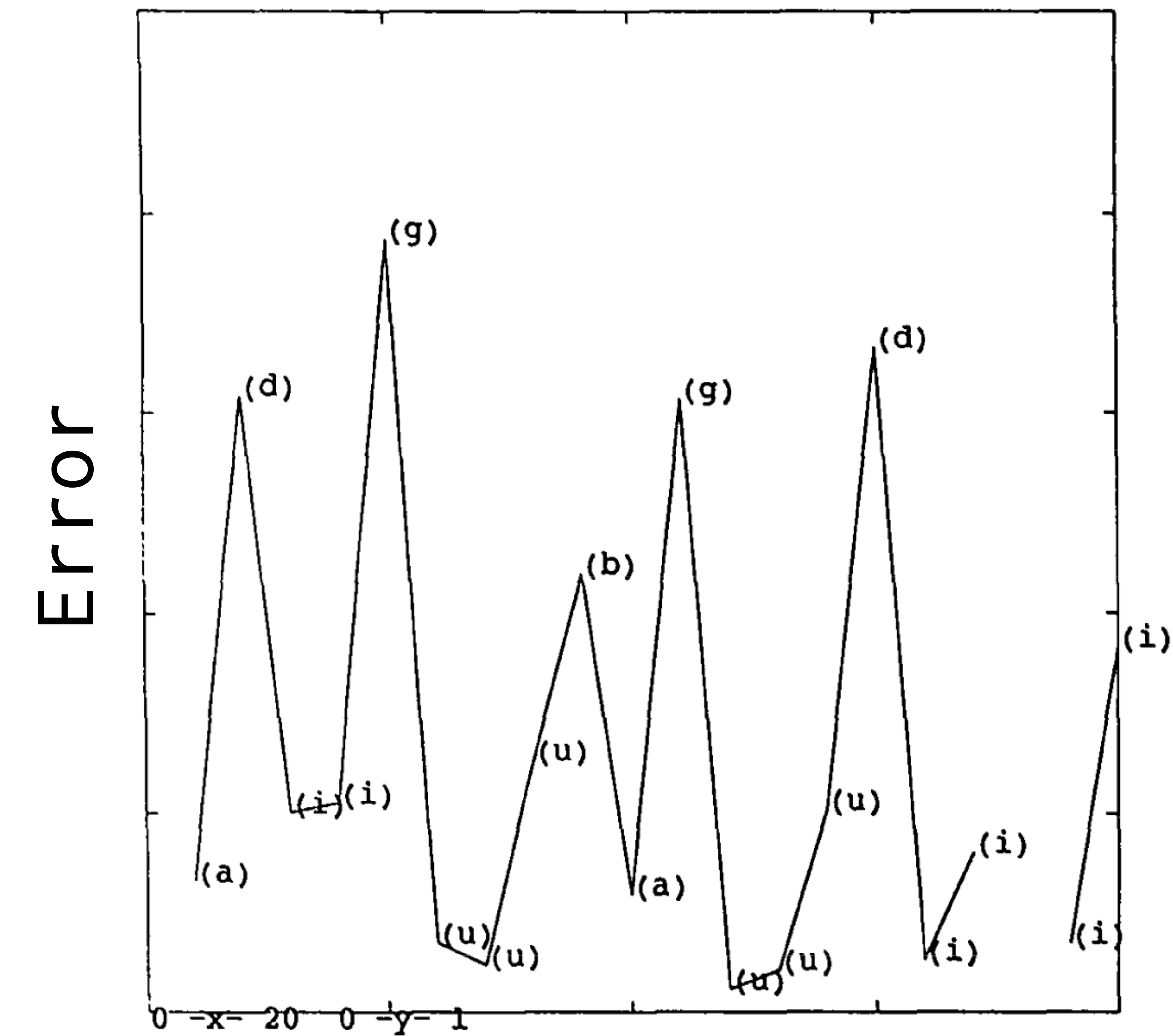
a d i i b a g u u...

Vector Definitions of Alphabet

	Consonant	Vowel	Interrupted	High	Back	Voiced
b	[1	0	1	0	0	1]
d	[1	0	1	1	0	1]
g	[1	0	1	0	1	1]
a	[0	1	0	0	1	1]
i	[0	1	0	1	0	1]
u	[0	1	0	1	1	1]



What has the network learned?

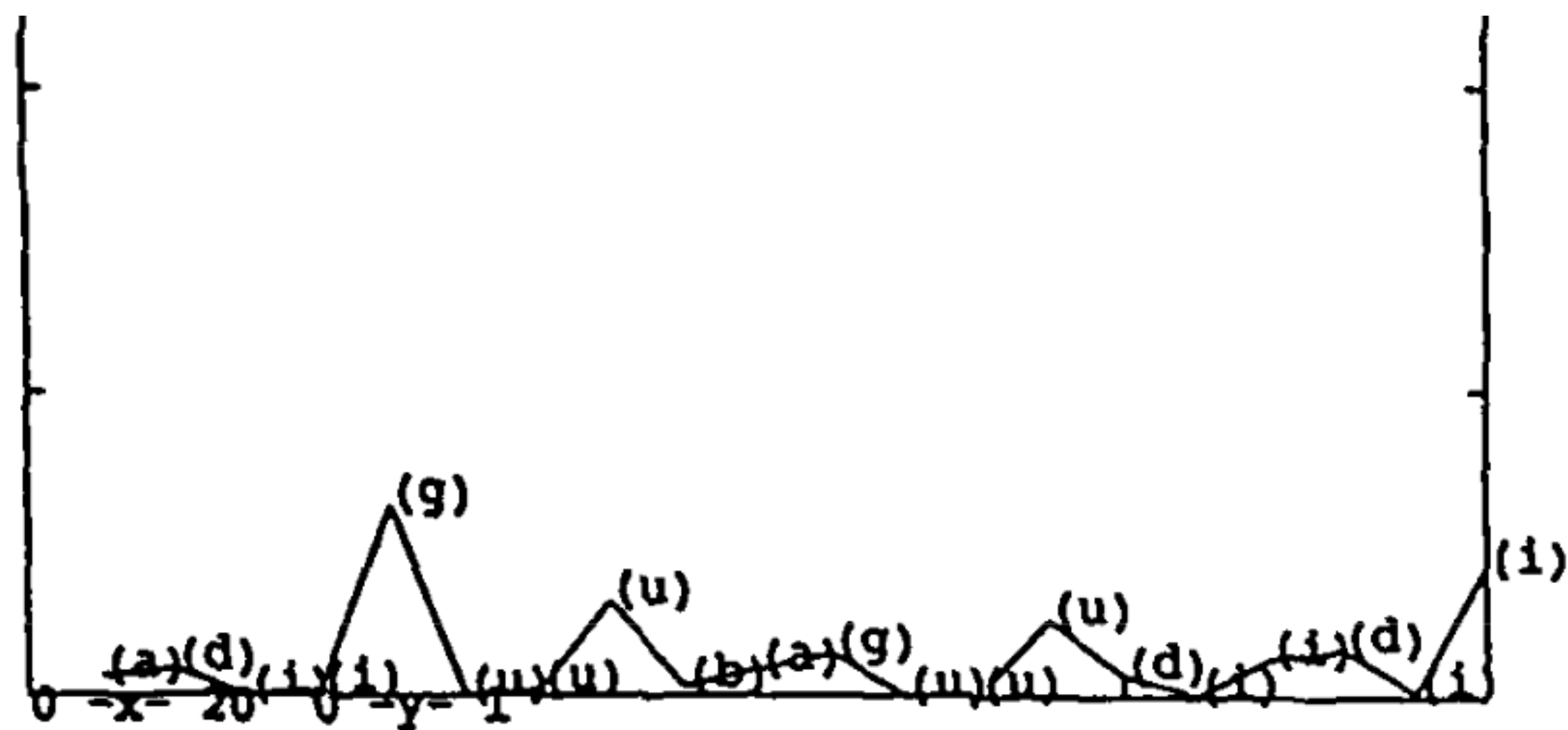


1. Averaging over the bits,
it learns **which letters
form words**

Vector Definitions of Alphabet						
	Consonant	Vowel	Interrupted	High	Back	Voiced
b	[1	0	1	0	0	1]
d	[1	0	1	1	0	1]
g	[1	0	1	0	1	1]
a	[0	1	0	0	1	1]
i	[0	1	0	1	0	1]
u	[0	1	0	1	1	1]

What has the network learned?

Error on bit 1



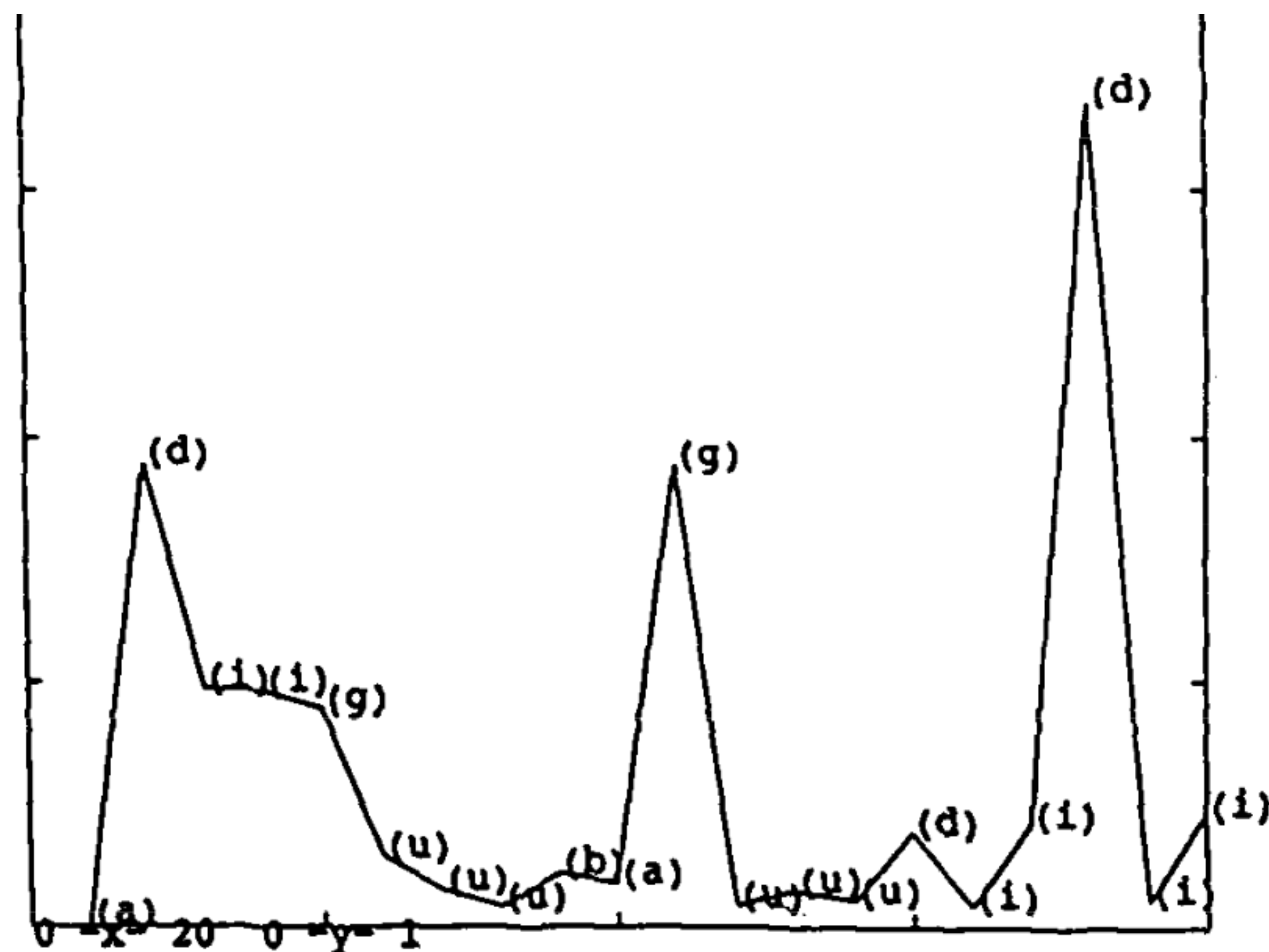
2. **Vowels follow consonants**

Vector Definitions of Alphabet

	Consonant	Vowel	Interrupted	High	Back	Voiced
b	[1	0	1	0	0	1]
d	[1	0	1	1	0	1]
g	[1	0	1	0	1	1]
a	[0	1	0	0	1	1]
i	[0	1	0	1	0	1]
u	[0	1	0	1	1	1]

What has the network learned?

Error on bit 4



2. Because **consonants** differ on the **High** feature, it knows that a consonant is coming but not **which one**

Vector Definitions of Alphabet

	Consonant	Vowel	Interrupted	High	Back	Voiced
b	[1	0	1	0	0	1]
d	[1	0	1	1	0	1]
g	[1	0	1	0	1	1]
a	[0	1	0	0	1	1]
i	[0	1	0	1	0	1]
u	[0	1	0	1	1	1]

What is a “word”?

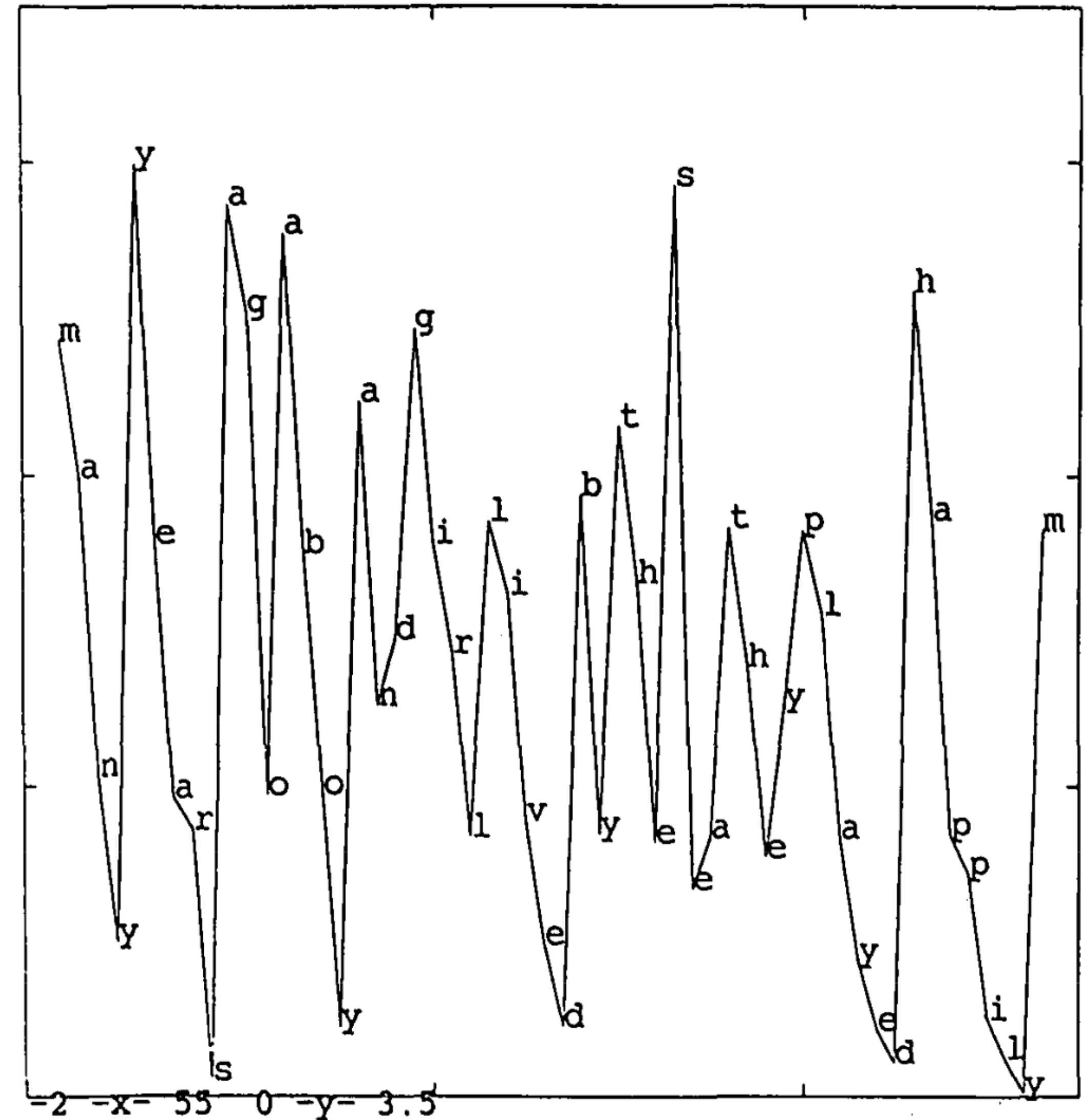
Input

A concatenation of words in English Sentences

manyyearsago**boy**and**girl** ...

A lot of the previous work **assumed** structure in Language (e.g., phonemes, morphemes, words)

But what if “**words**” are just sequences of low prediction error



Syntactic structure through prediction error

Input

A concatenation of triplets

subject - verb - object

womansmashplatecatmovemanbreak

WORD 1	WORD 2	WORD 3
NOUN-HUM	VERB-EAT	NOUN-FOOD
NOUN-HUM	VERB-PERCEPT	NOUN-INANIM
NOUN-HUM	VERB-DESTROY	NOUN-FRAG
NOUN-HUM	VERB-INTRAN	
NOUN-HUM	VERB-TRAN	NOUN-HUM
NOUN-HUM	VERB-AGPAT	NOUN-INANIM
NOUN-HUM	VERB-AGPAT	
NOUN-ANIM	VERB-EAT	NOUN-FOOD
NOUN-ANIM	VERB-TRAN	NOUN-ANIM
NOUN-ANIM	VERB-AGPAT	NOUN-INANIM
NOUN-ANIM	VERB-AGPAT	
NOUN-INANIM	VERB-AGPAT	
NOUN-AGRESS	VERB-DESTROY	NOUN-FRAG
NOUN-AGRESS	VERB-EAT	NOUN-HUM
NOUN-AGRESS	VERB-EAT	NOUN-ANIM
NOUN-AGRESS	VERB-EAT	NOUN-FOOD

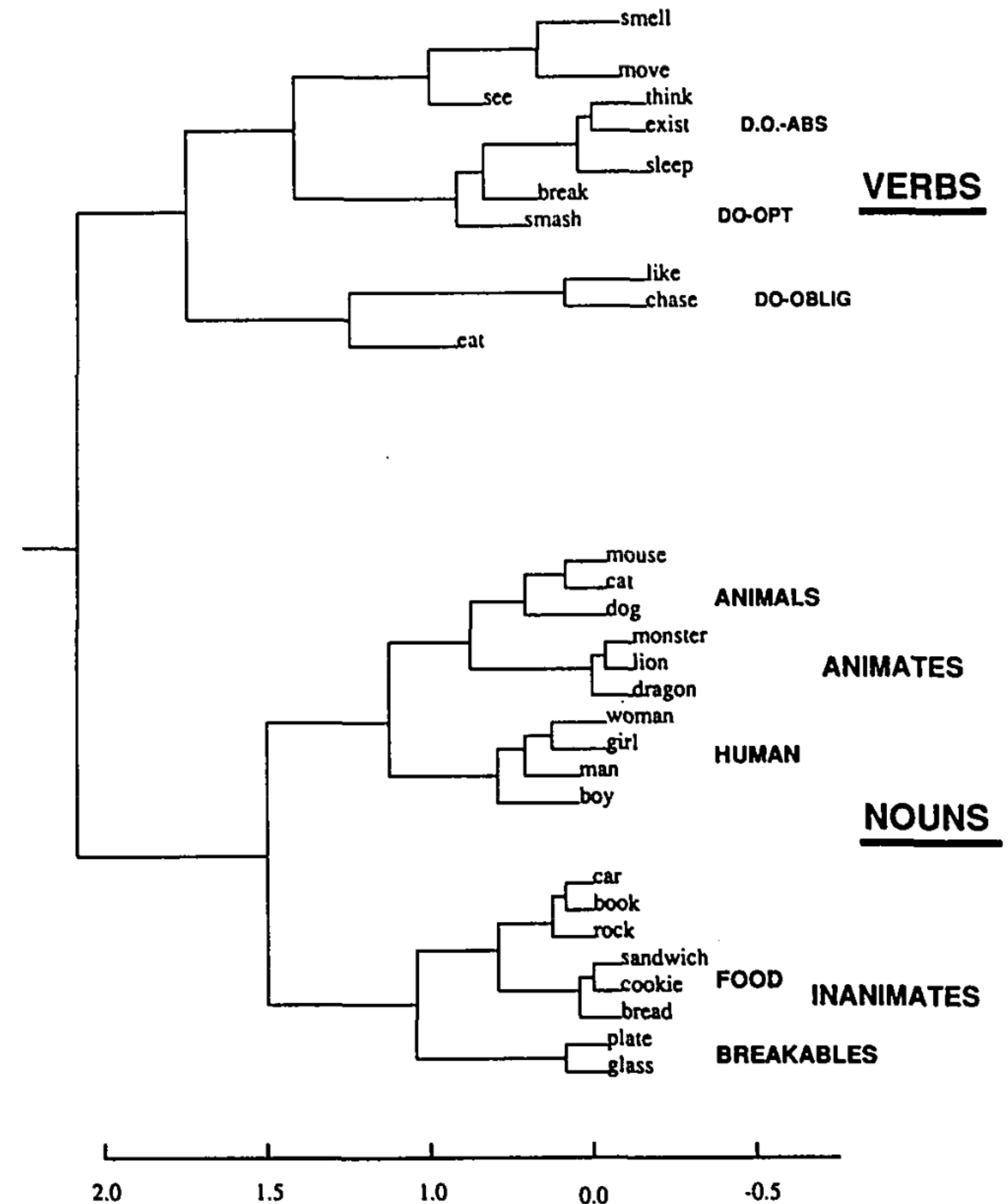
Input	Output
000000000000000000000000000010 (woman)	000000000000000000000000000010000 (smash)
000000000000000000000000000010000 (smash)	00000000000000000000000000001000000000 (plate)
00000000000000000000000000001000000000 (plate)	0000010000000000000000000000000000 (cat)
0000010000000000000000000000000000 (cat)	0000000000000000000000000000100000000000 (move)
0000000000000000000000000000100000000000 (move)	0000000000000000000000000000100000000000000 (man)
0000000000000000000000000000100000000000000 (man)	00010000000000000000000000000000000000 (break)
00010000000000000000000000000000000000 (break)	00001000000000000000000000000000000000 (car)
00001000000000000000000000000000000000 (car)	01000000000000000000000000000000000000 (boy)
01000000000000000000000000000000000000 (boy)	000000000000000000000000000000000000001000000000000 (move)
000000000000000000000000000000000000001000000000000 (move)	00000000000000000000000000000000000000100000000000000 (girl)
0000000000000000000000000000000000000010000000000000 (girl)	0000000000000000000000000000000000000010000000000000000 (eat)
00000000000000000000000000000000000000100000000000000 (eat)	00100 (bread)
00100 (bread)	0000000000100 (dog)
0000000000100 (dog)	0001000000000000 (move)
0001000000000000 (move)	000100000000000000 (mouse)
000100000000000000 (mouse)	000100000000000000 (mouse)
000100000000000000 (mouse)	000100000000000000 (move)
000100000000000000 (move)	100 (book)
100 (book)	00010000000000000000 (lion)

Syntactic structure through prediction error

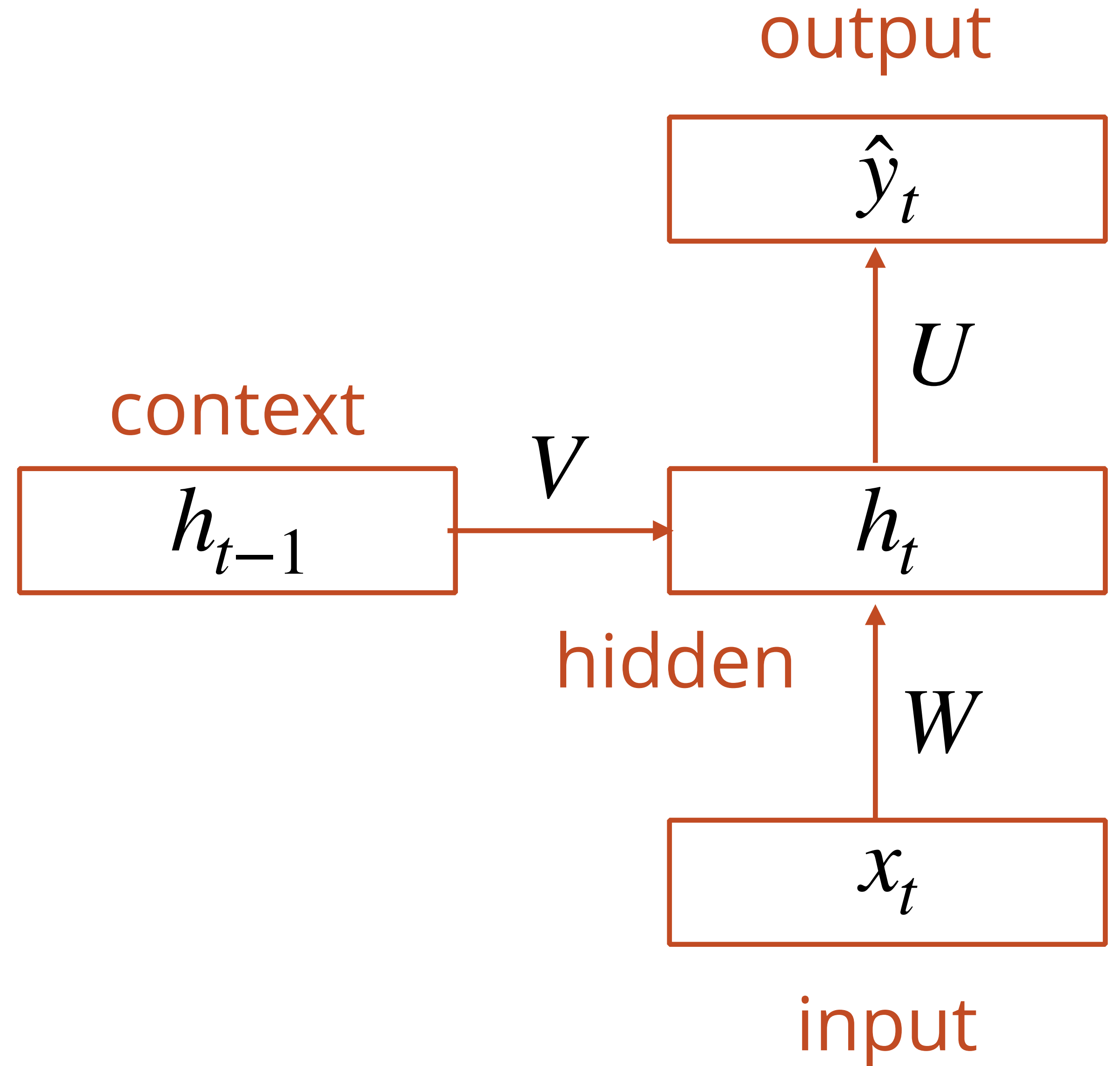
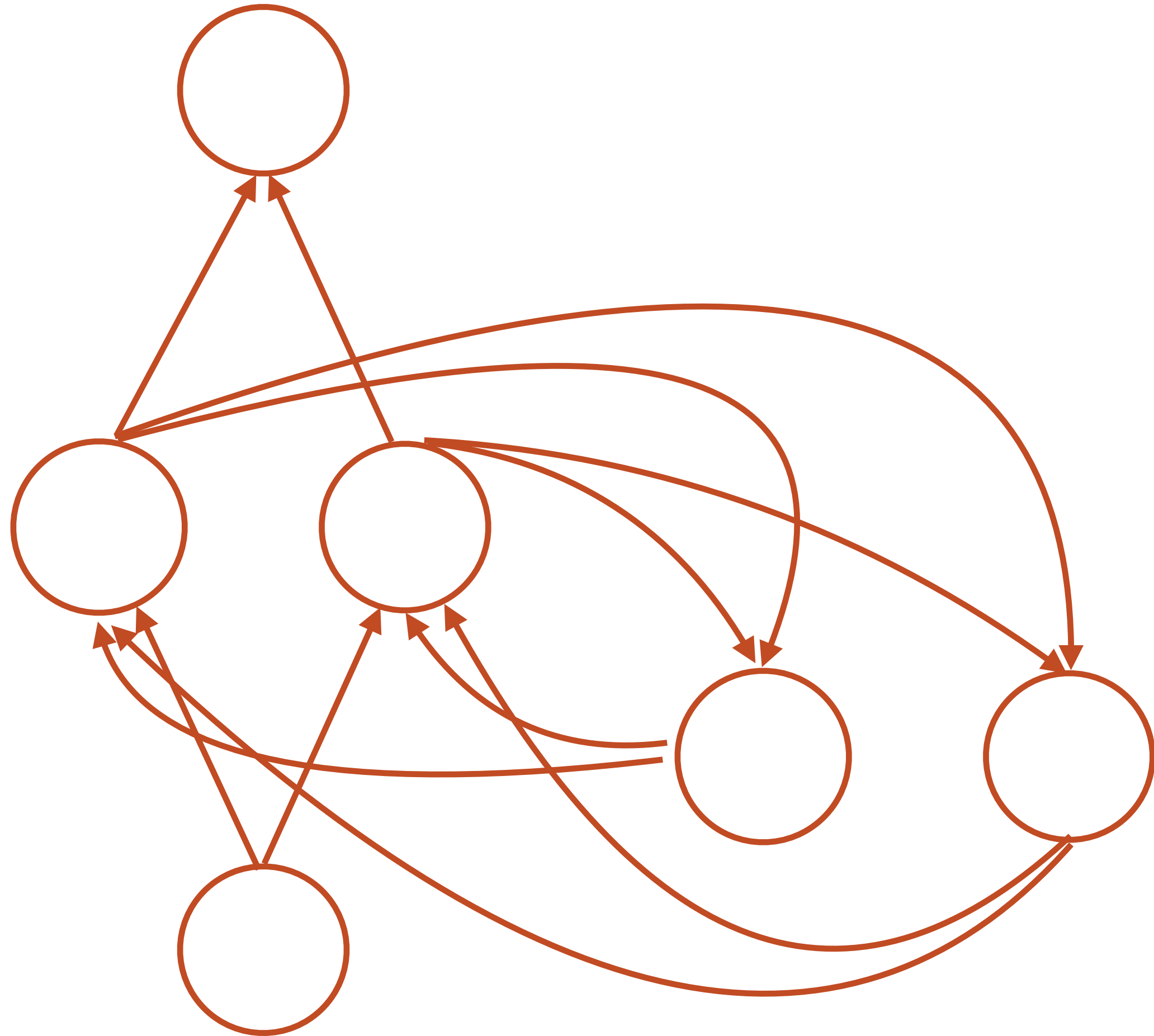
Hierarchically clustering the hidden layer activations for words reveals structure!

The network learns **syntactic** and **semantic** roles

Why?

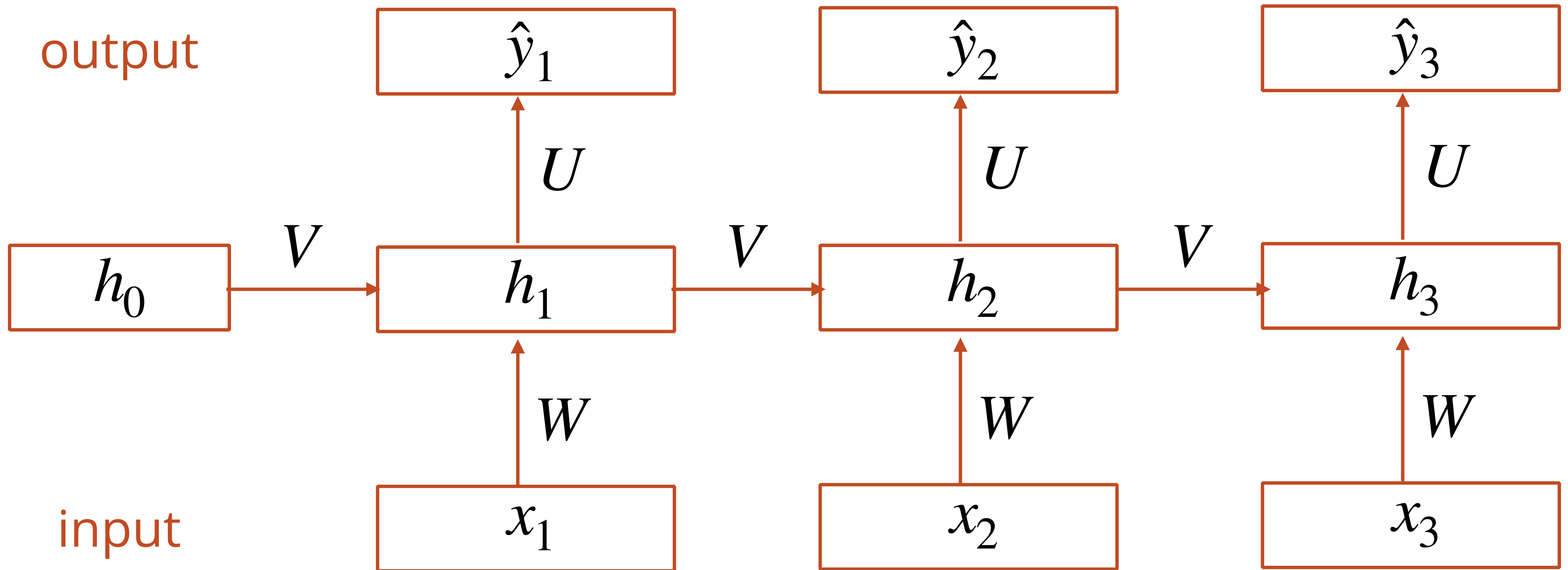


Training a Recurrent Neural Network



Unrolling a network in time

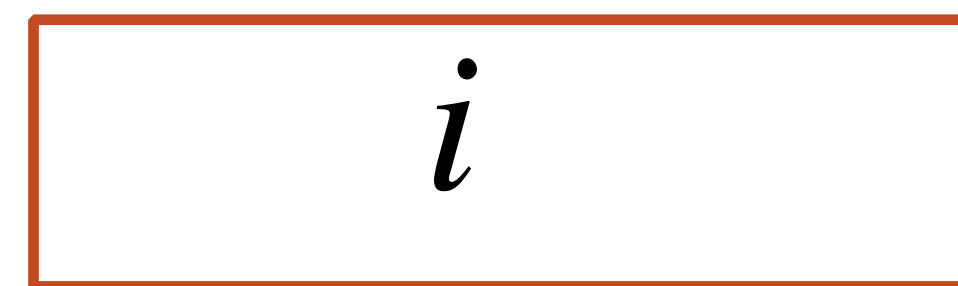
Global Error: $E = \sum_t E_t$



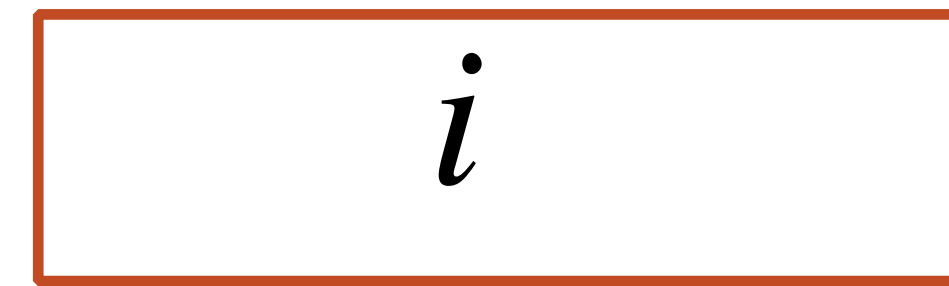
Unrolling a network in time

Global Error: $E = \sum_t E_t$

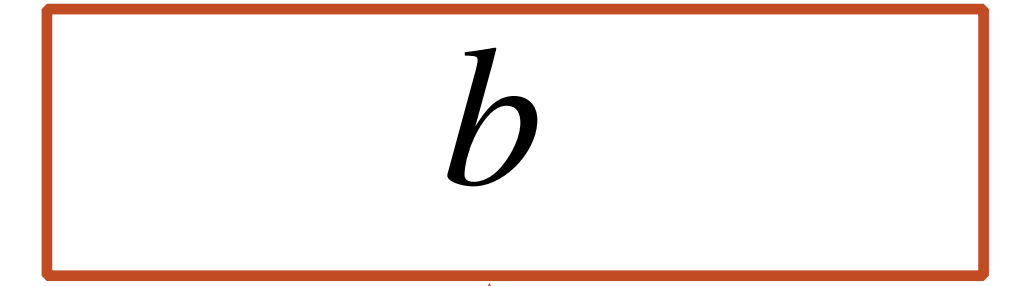
output



U

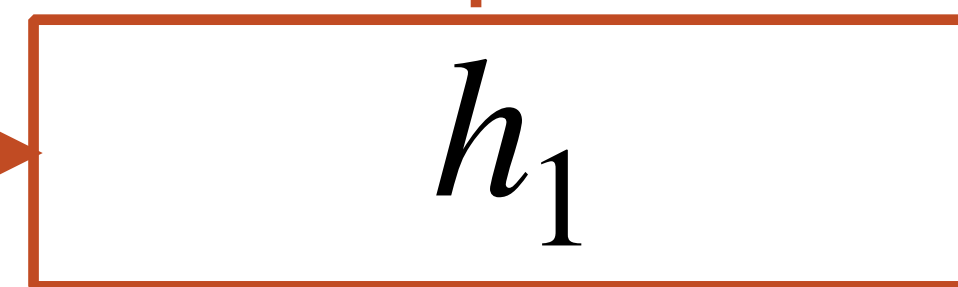


U

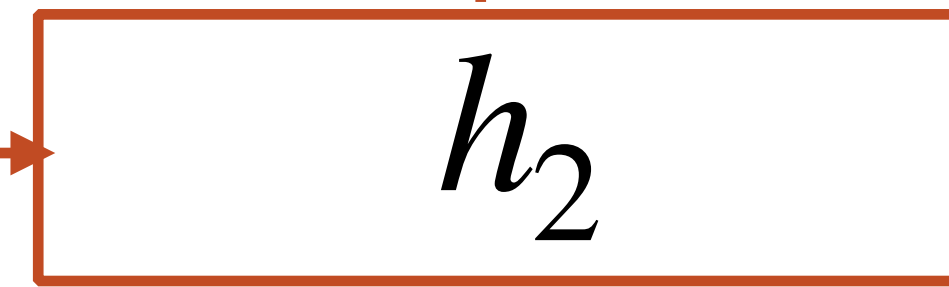


U

V



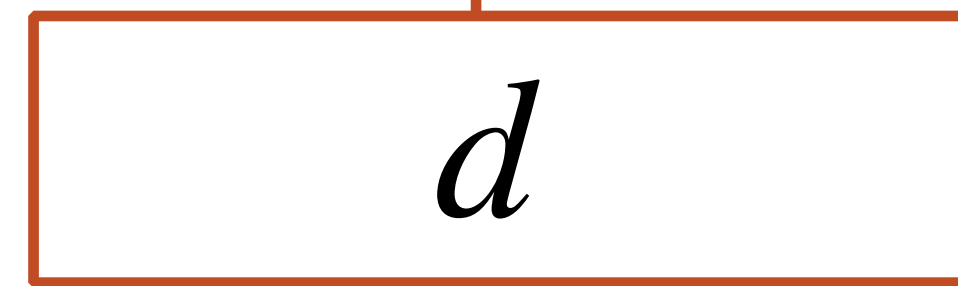
V



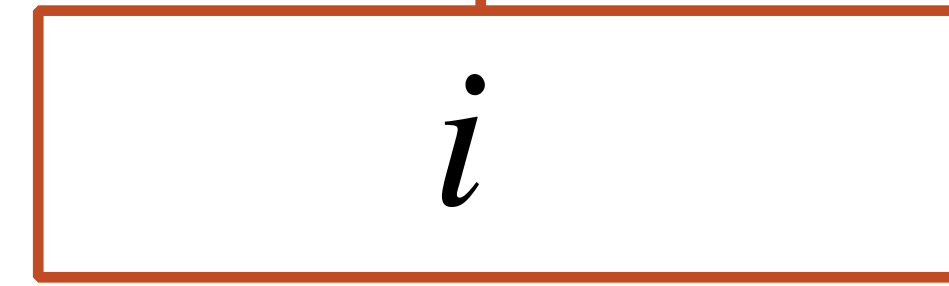
V



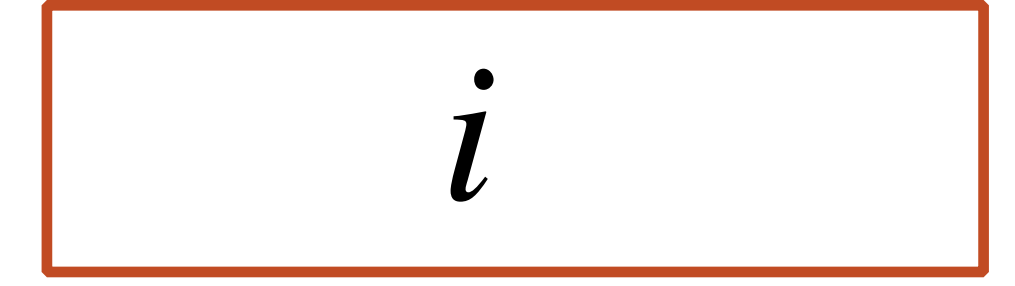
W



W



W

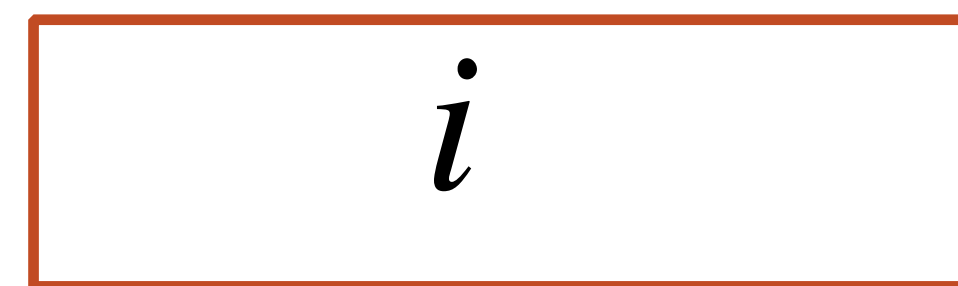


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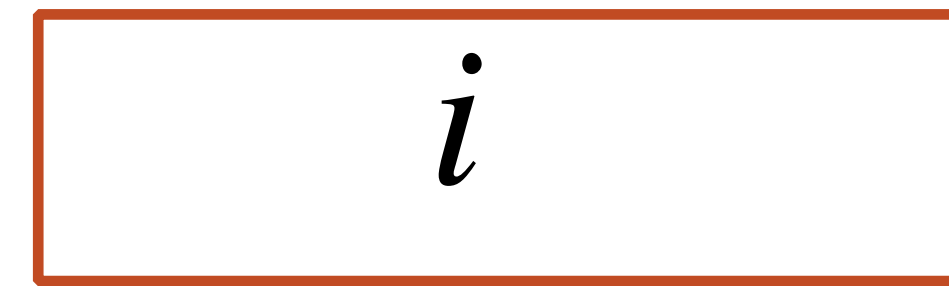
Backpropagation through time

Global Error: $E = \sum_t E_t$ $\frac{\partial E}{\partial W} = \sum_t \frac{\partial E_t}{\partial W}$

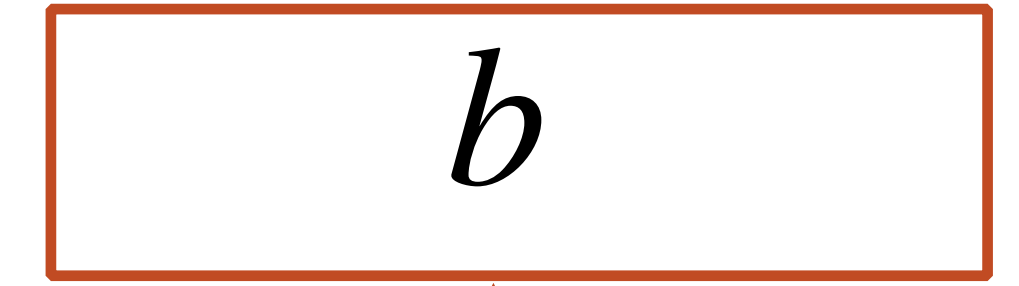
output



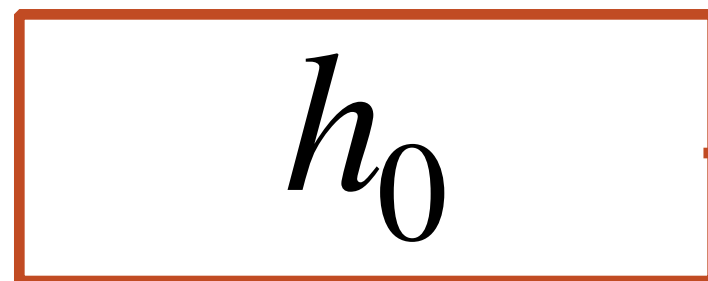
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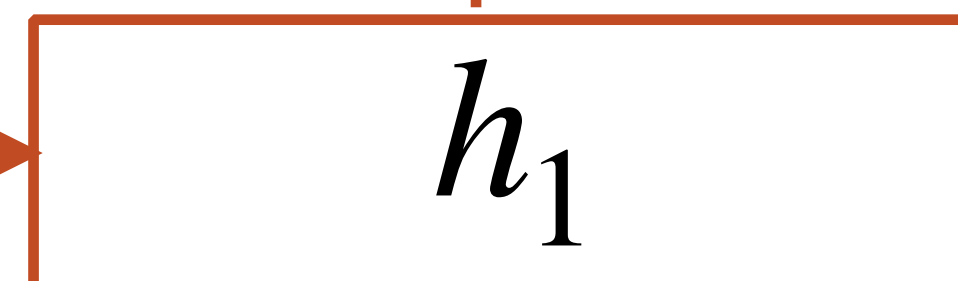
U



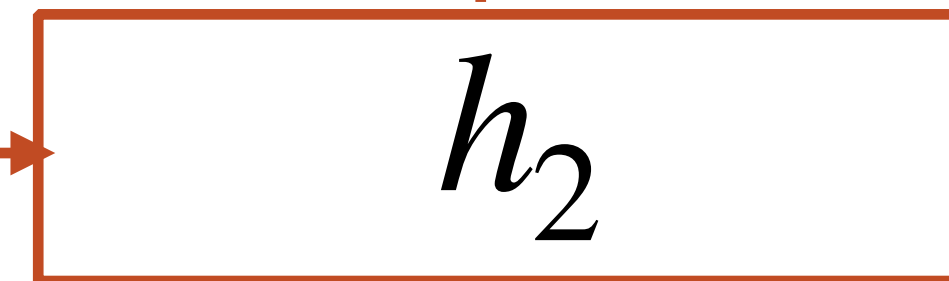
U



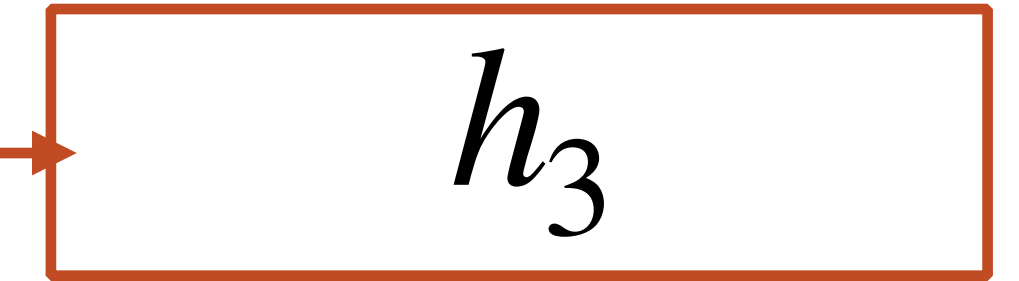
V



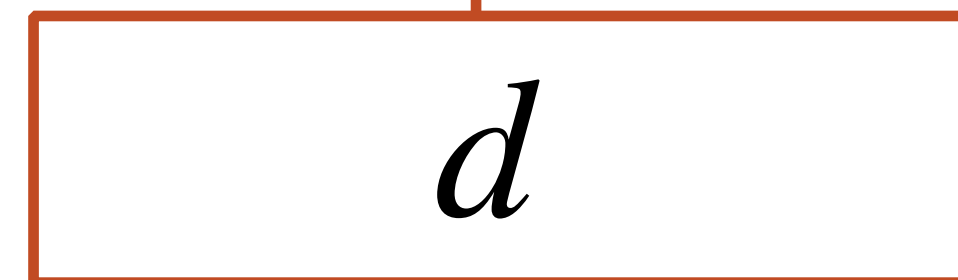
V



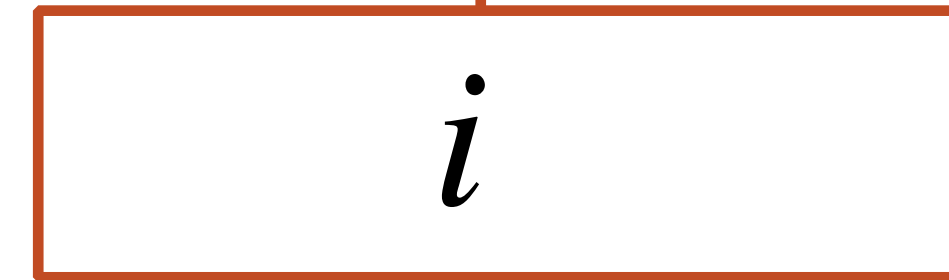
V



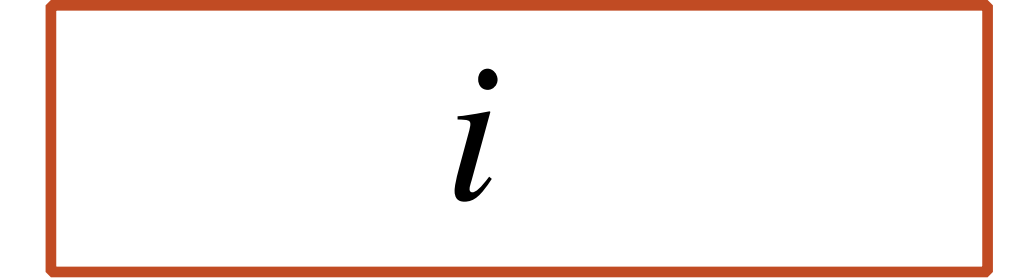
W



W



W



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Backpropagation through time

Global Error: $E = \sum_t E_t$ $\frac{\partial E}{\partial W} = \sum_t \frac{\partial E_t}{\partial W}$

output

i

i

b

U

U

$\frac{\partial E_3}{\partial h_3}$
 U

V

V

V

h_0

h_1

h_2

h_3

W

W

$\frac{\partial h_3}{\partial W}$
 W

diibaguuu

d

i

i

Backpropagation through time

Global Error: $E = \sum_t E_t$ $\frac{\partial E}{\partial W} = \sum_t \frac{\partial E_t}{\partial W}$

output

i

i

b

U

U

$\frac{\partial E_3}{\partial h_3}$
 U

V

V

V

h_0

h_1

h_2

h_3

$\frac{\partial h_3}{\partial h_2}$

W

$\frac{\partial h_2}{\partial W}$
 W

W

diibaguuu

d

i

i

Backpropagation through time

Global Error: $E = \sum_t E_t$ $\frac{\partial E}{\partial W} = \sum_t \frac{\partial E_t}{\partial W}$

output

i

i

b

U

U

$\frac{\partial E_3}{\partial h_3}$
 U

V

V

V

h_0

h_1

h_2

h_3

$\frac{\partial h_1}{\partial h_0}$
 W

$\frac{\partial h_2}{\partial h_1}$

W

$\frac{\partial h_3}{\partial h_2}$

W

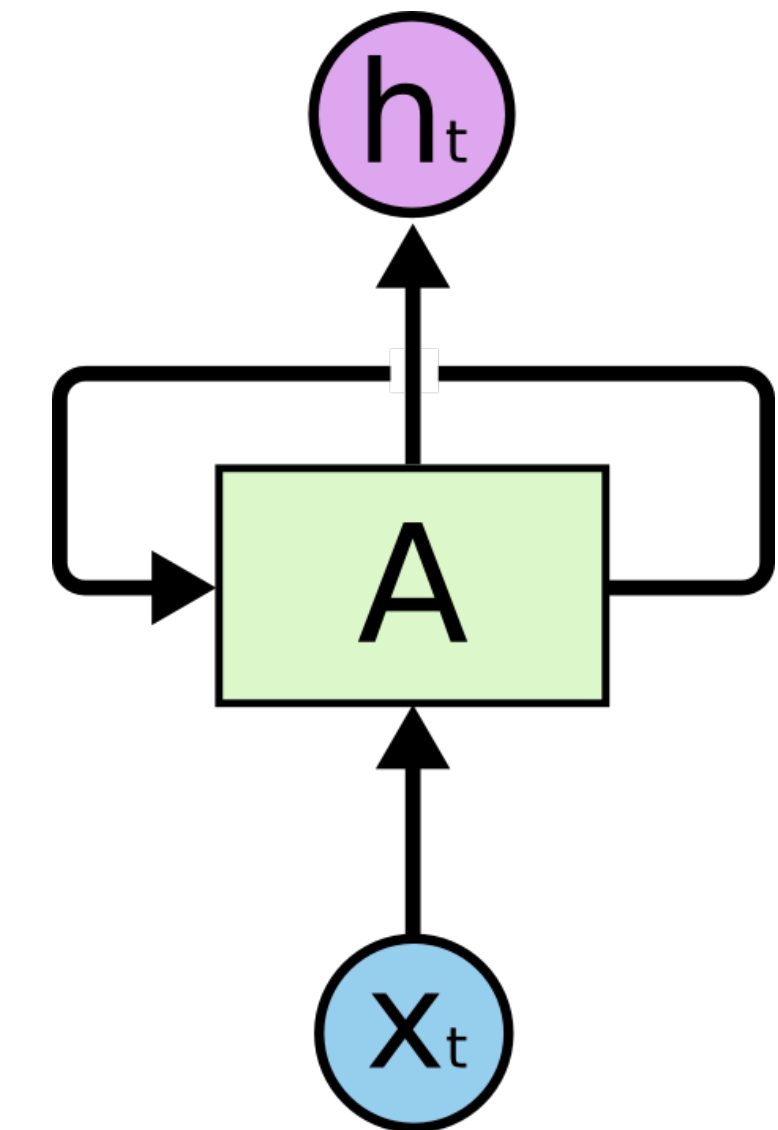
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d

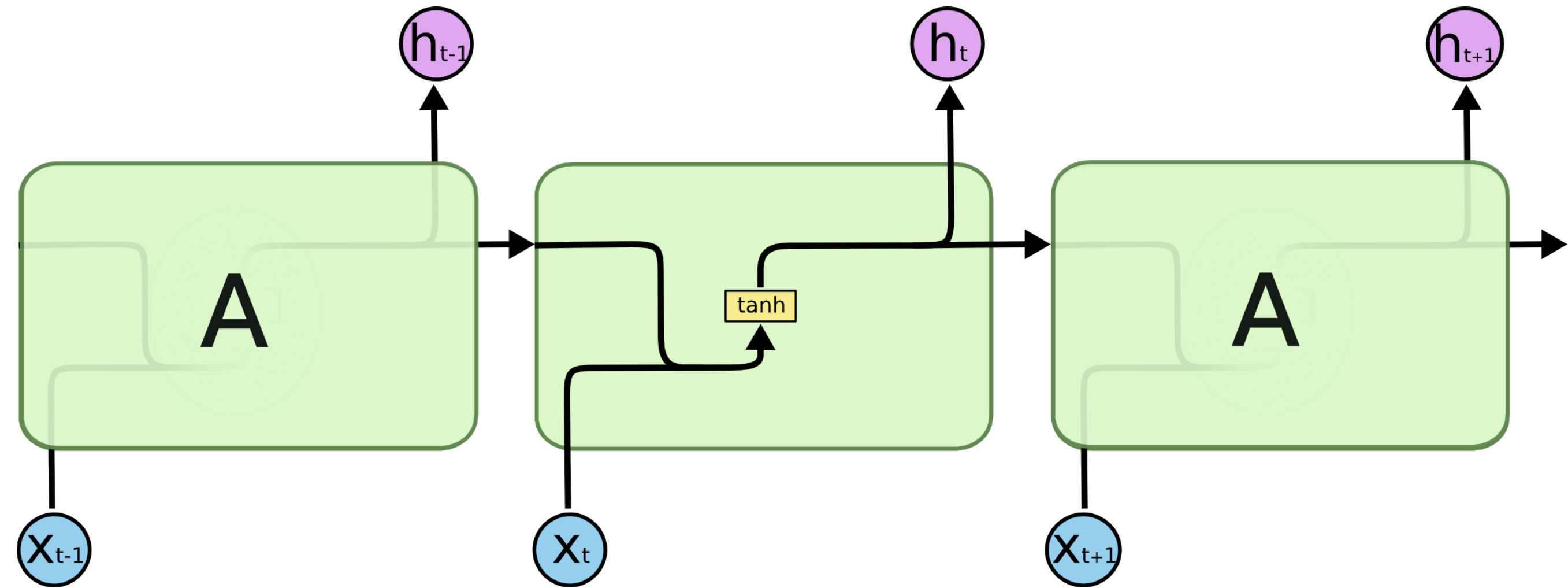
i

i

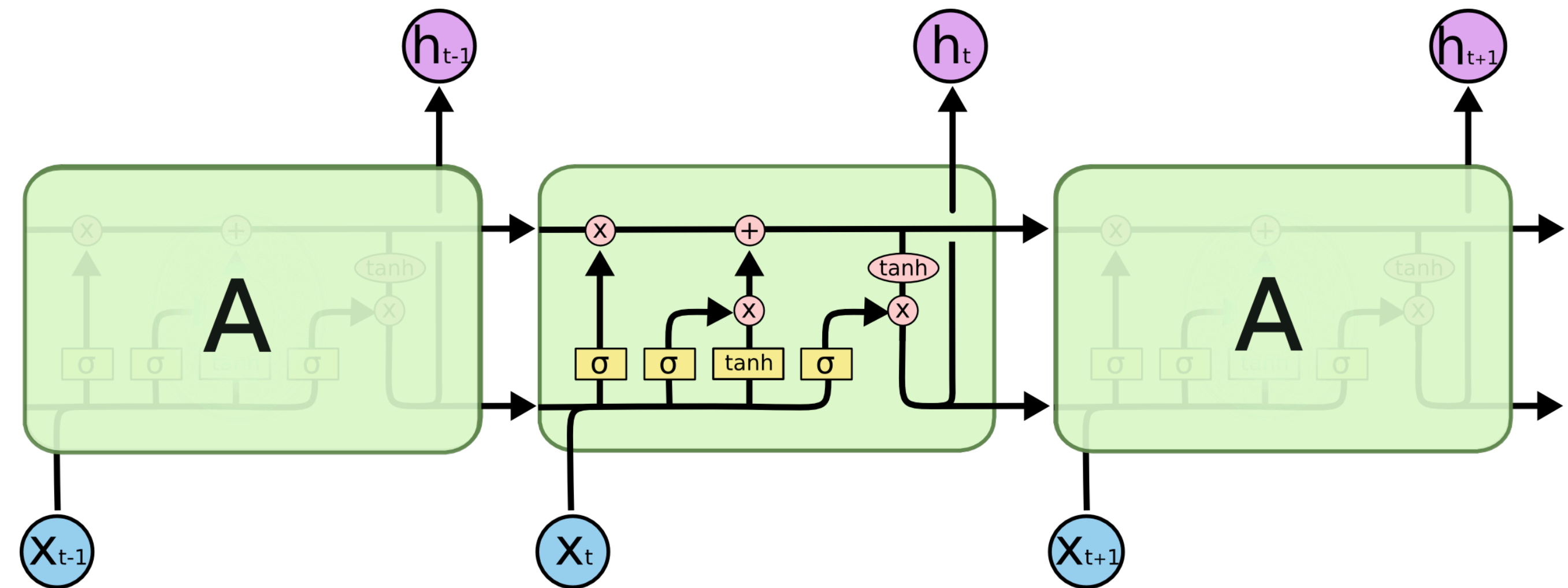
Modern language models extend this idea



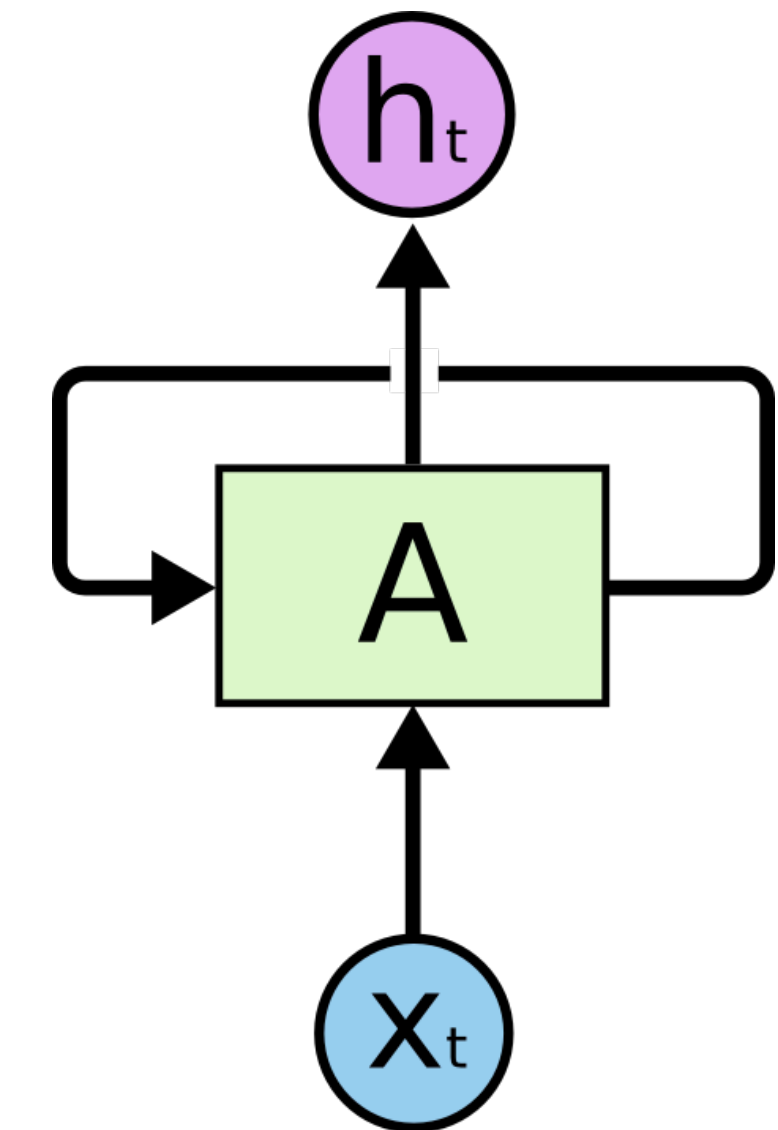
Elman Network



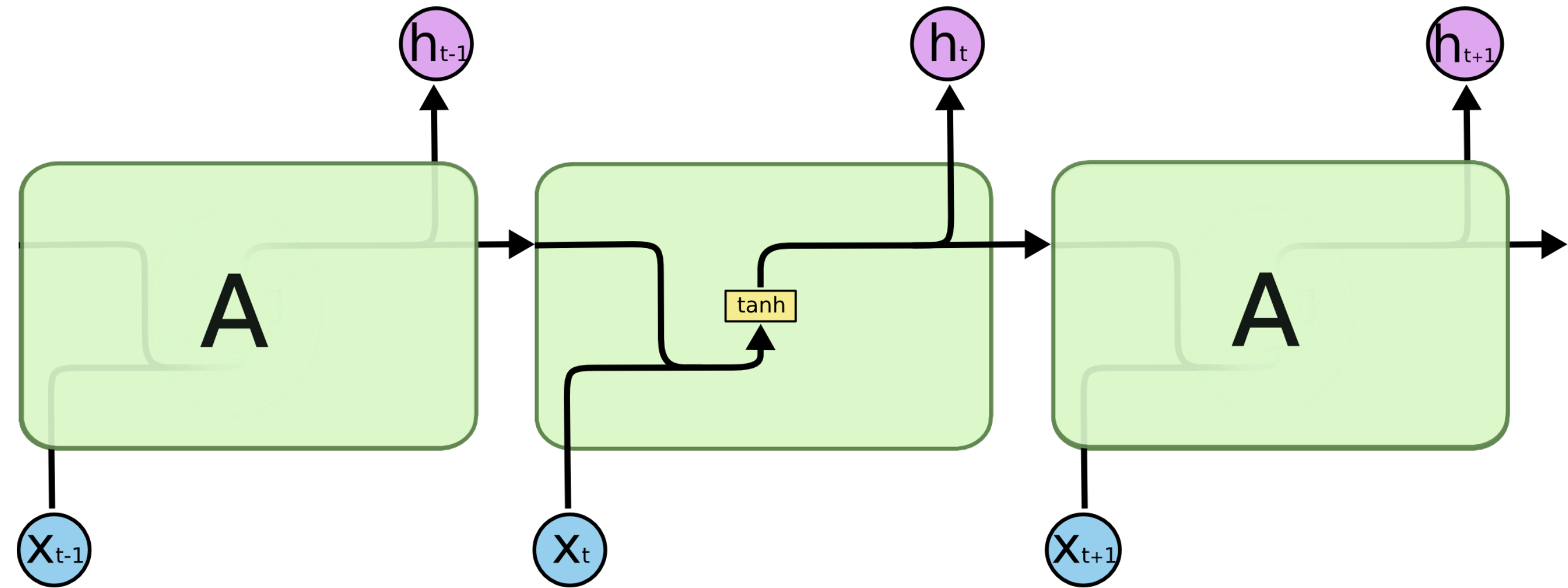
Long Short Term Memory (LSTM)



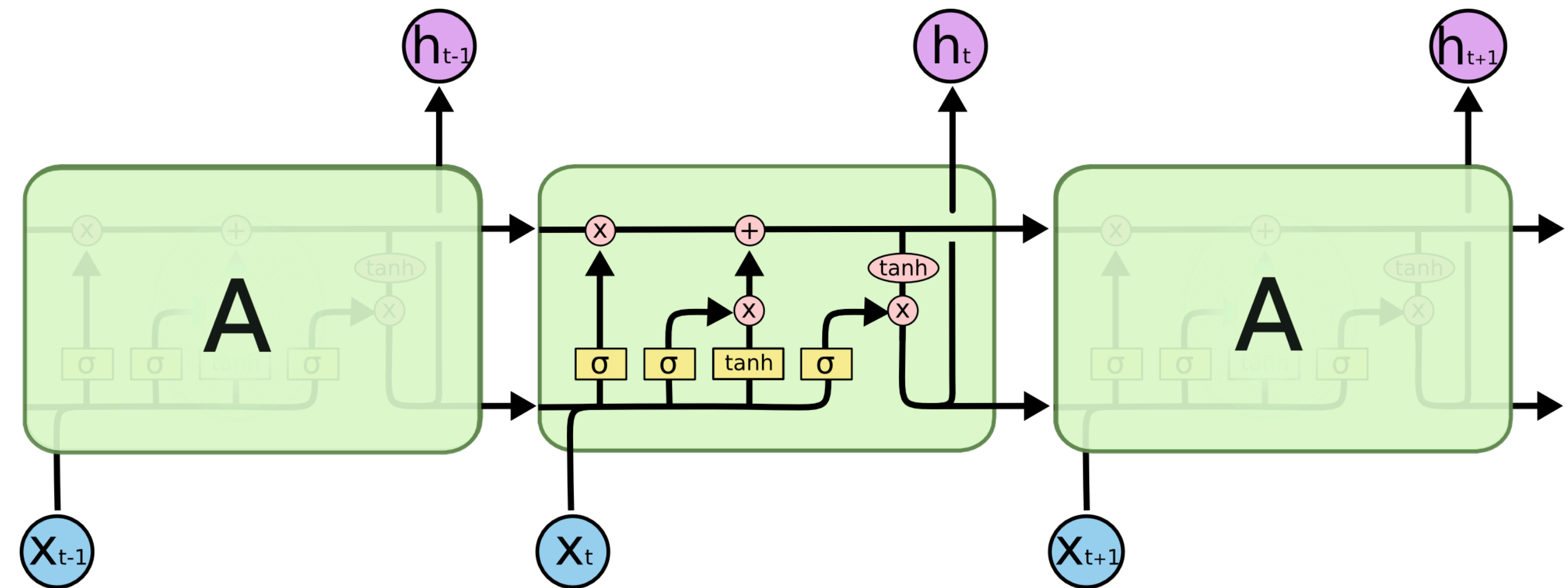
Modern language models extend this idea



Elman Network



Long Short Term Memory (LSTM)



Connectionism as a framework for learning

1. Complex behavior can arise from the interaction of simple individual units (emergence)
2. Rule-like behavior need not be controlled by rule-like representations (sub-symbolic computation)
3. You don't need to know the structure of a domain priori to model learning in that domain. (Domain-general)
 1. Although it can help with deciding network architecture
4. The same kind of learning model can work for explicit supervision and prediction

Limits to connectionism

1. Very data-hungry
2. Very computation-intensive
3. No built-in distinctions between relevant and irrelevant information
4. No built-in distinctions between correlation and causation
5. Hard to learn abstractions (or approximate them)
6. Models are large and hard to understand



VOTING RESOURCES

VOTER REGISTRATION

Register! <https://turbovote.org/>

Check your voter registration!

<https://www.vote.org/am-i-registered-to-vote/>

CAN YOU VOTE IN PA?

College students can register to vote in Pennsylvania after living here for 30 days.

Students can give either an on-campus or an off-campus Pennsylvania address.

Students can use a school photo ID as verification.

MAKE A PLAN

Make a voting plan ahead of time. Ask me for help if you have questions.

If you are voting by mail, make sure to request your ballot asap! <https://www.vote.org/absentee-ballot/>

NEW THIS YEAR IN PA: EARLY VOTING!

If you have NOT requested a mail ballot, you can vote early at these satellite locations!
If you HAVE requested your mail ballot, you can drop it off at these locations!

← Shop 'N Save Hill District: 18...

name

Shop 'N Save Hill District: 1850 Centre Ave


description

Saturday, Oct. 17 – Hours: 9 a.m. to 5 p.m.
Sunday, Oct. 18 – Hours: 11 a.m. to 7 p.m.

Details from Google Maps

1850 Centre Ave, Pittsburgh, PA 15219

[View in Google Maps](#)



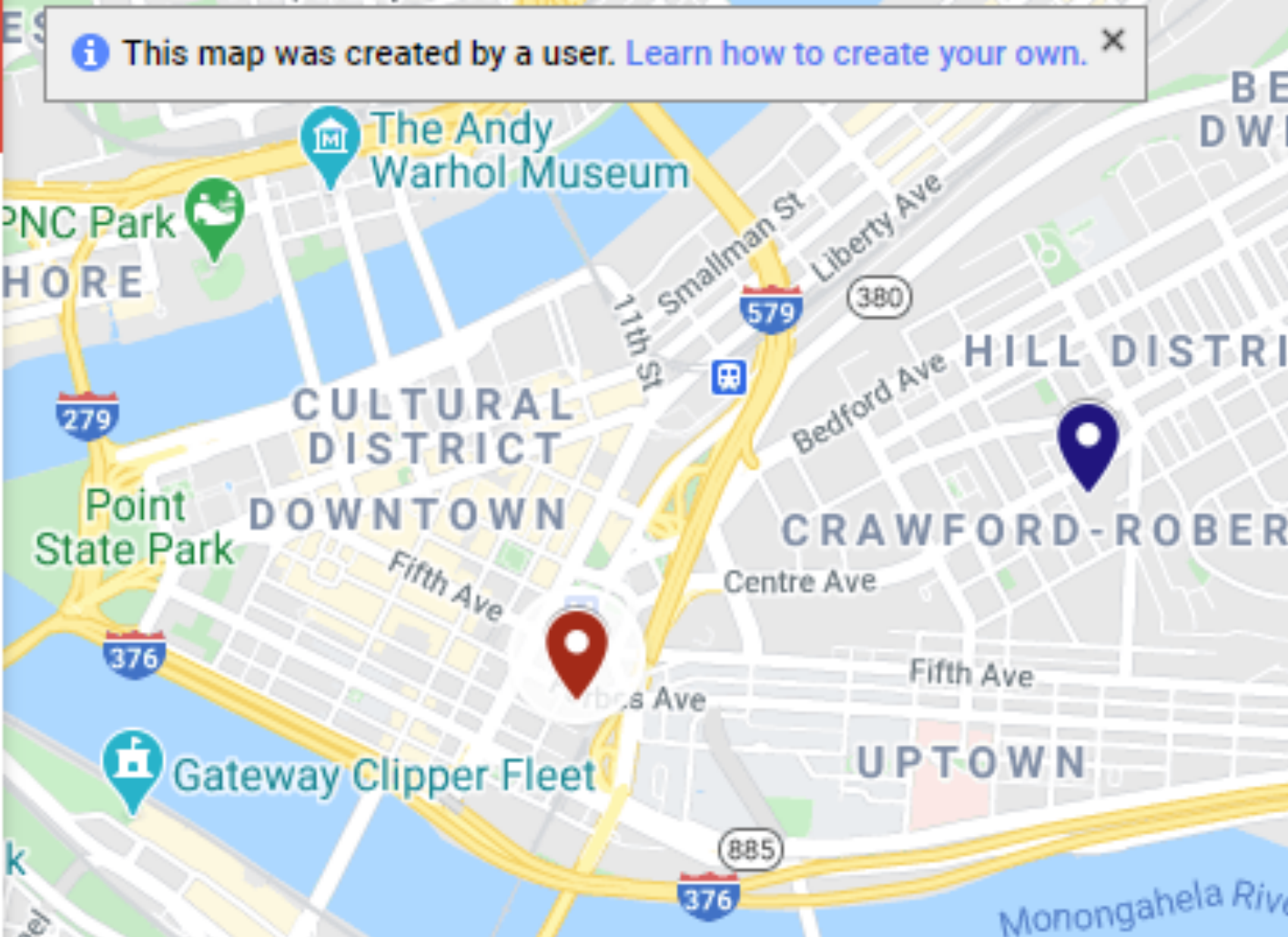
← County Office Building: 542 ...

name

County Office Building: 542 Forbes Ave

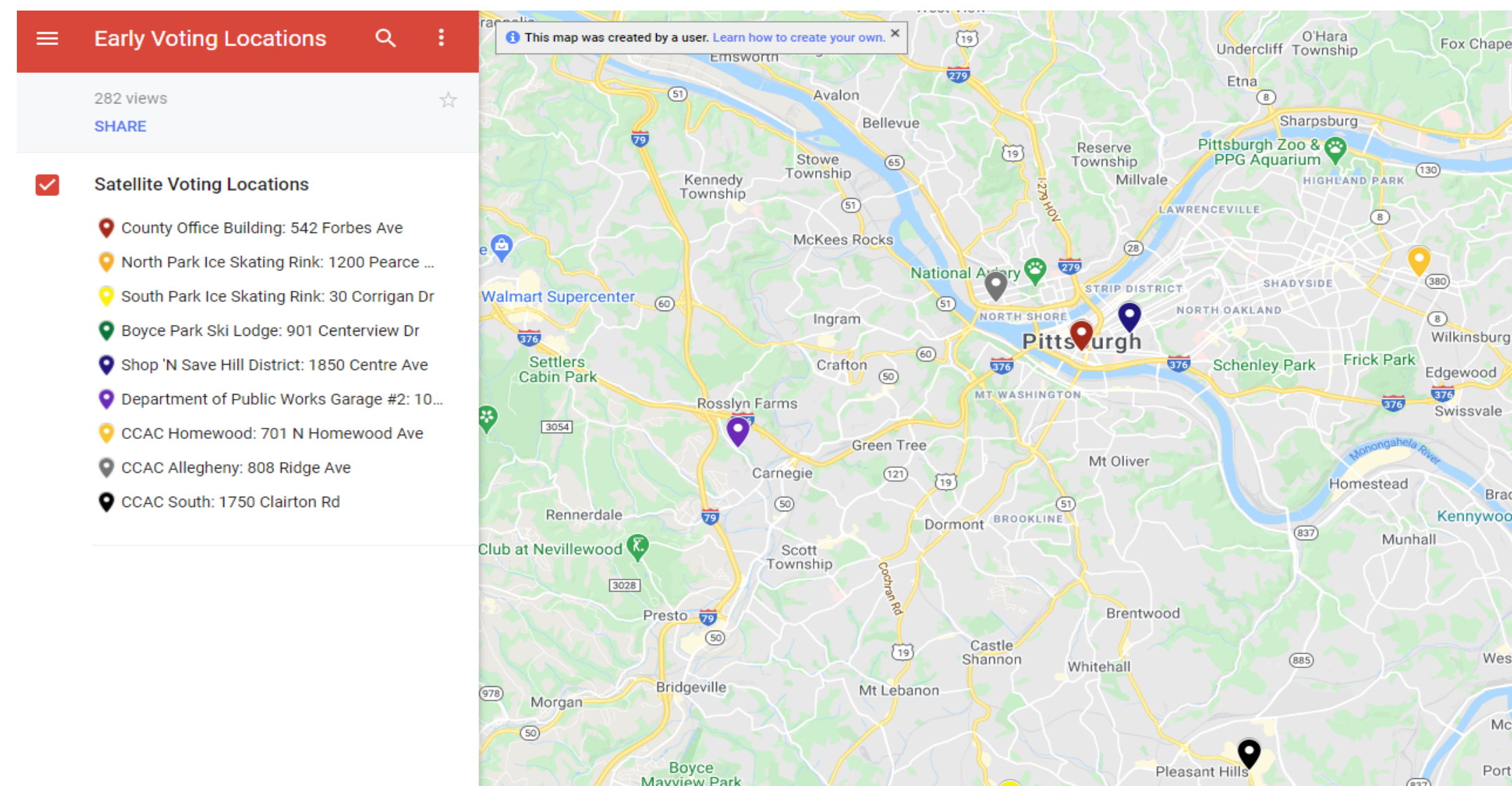
description

Saturday, Oct. 10 – Hours: 9 a.m. to 5 p.m.
Sunday, Oct. 11 – Hours: 11 a.m. to 7 p.m.
Saturday, Oct. 17 – Hours: 9 a.m. to 5 p.m.
Sunday, Oct. 18 – Hours: 11 a.m. to 7 p.m.
Saturday, Oct. 24 – Hours: 9 a.m. to 5 p.m.
Sunday, Oct. 25 – Hours: 11 a.m. to 7 p.m.



NEW THIS YEAR IN PA: EARLY VOTING!

You can see all of the satellite locations on this great map! <https://tinyurl.com/yymhqsph>



Saturday, October 10 – Hours: 9 AM to 5 PM

1. County Office Building
2. North Park Ice Rink
3. South Park Ice Rink
4. DPW Garage # 2 (Carnegie)
5. CCAC Homewood

Sunday, October 11 – Hours: 11 AM to 7 PM

1. County Office Building
2. North Park Ice Rink
3. South Park Ice Rink
4. DPW Garage # 2 (Carnegie)
5. CCAC Homewood

Saturday, October 17 – Hours: 9 AM to 5 PM

1. County Office Building
2. South Park Ice Rink
3. CCAC South
4. Boyce Park Ski Lodge
5. Shop 'n Save Hill District

Sunday, October 18 – Hours: 11 AM to 7 PM

1. County Office Building
2. South Park Ice Rink
3. CCAC South
4. Boyce Park Ski Lodge
5. Shop 'n Save Hill District

Saturday, October 24 – Hours: 9 AM to 5 PM

1. County Office Building
2. Boyce Park Ski Lodge
3. CCAC South
4. CCAC Allegheny
5. North Park Ice Rink

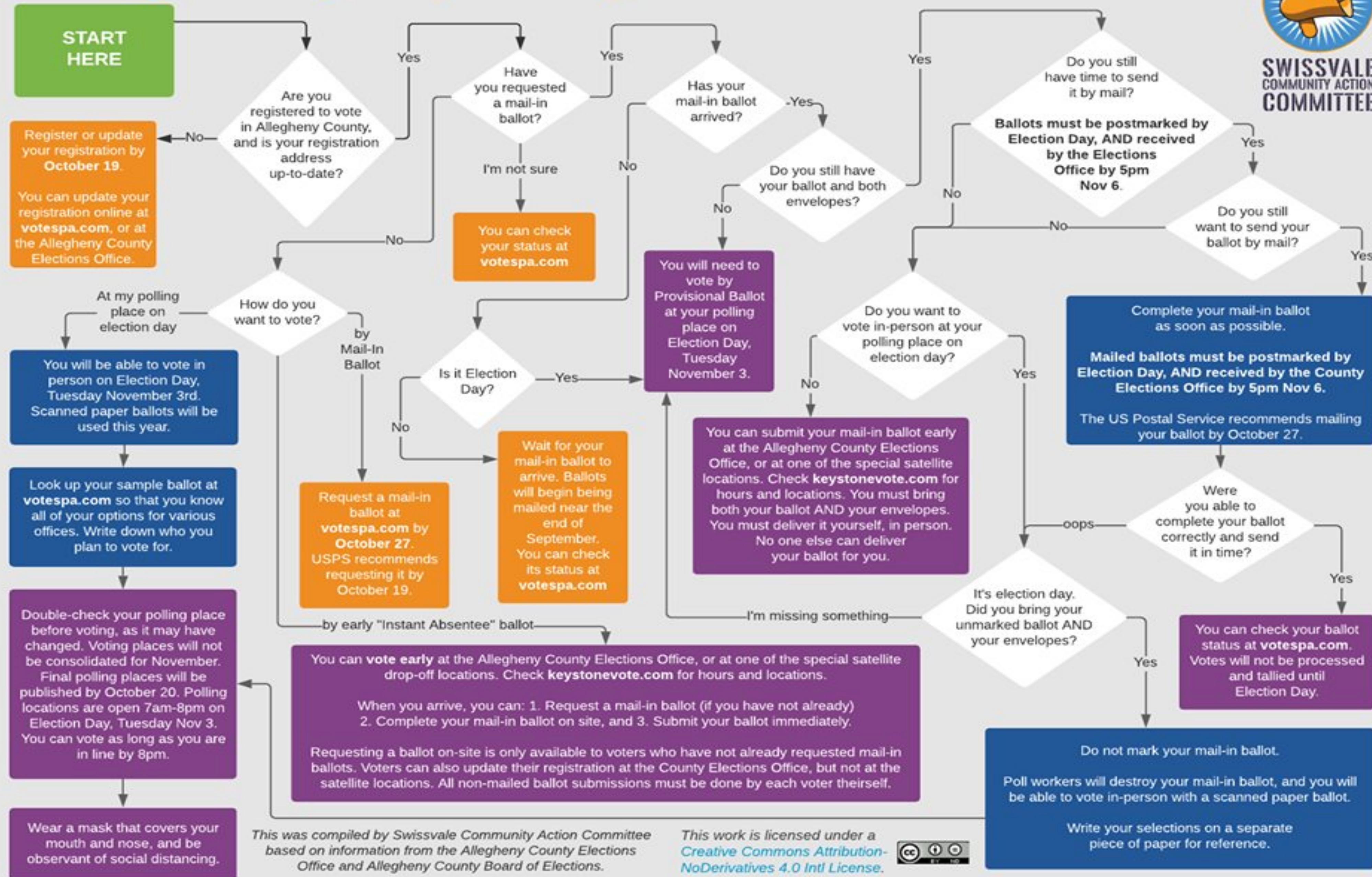
Sunday, October 25 – Hours: 11 AM to 7 PM

1. County Office Building
2. Boyce Park Ski Lodge
3. CCAC South
4. CCAC Allegheny
5. North Park Ice Rink

Voting in Allegheny County, November 2020



SWISSVALE
COMMUNITY ACTION
COMMITTEE

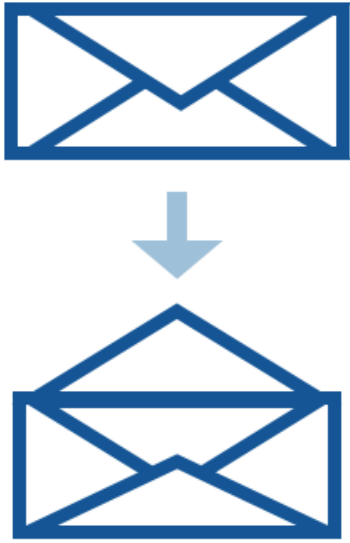


VOTING BY MAIL IN PA!

Watch this 1-min non-partisan video on how to complete your mail ballot correctly! <https://www.youtube.com/watch?v=krhGbx7fA4o&t=0s>

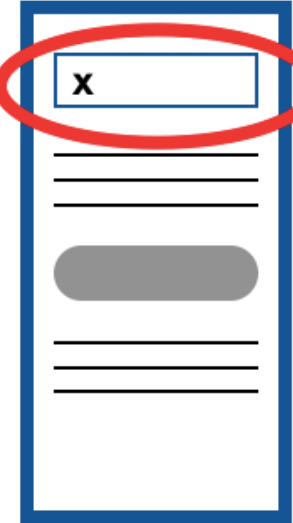
Voting by mail?

Make sure your vote is counted!



First seal the ballot inside of the secrecy envelope that says "Official Election Ballot."


Then place the sealed secrecy envelope inside of your barcoded envelope. Do not make any marks on the secrecy envelope!




Remember to sign and complete the voter's declaration on the back of the barcoded envelope.

Make sure to also seal the outside envelope!

REMEMBER: USE both envelopes, SEAL both envelopes, and SIGN the outer envelope! Then your ballot is ready to go!




PENNSYLVANIA VOTERS:




You may be naked, but your ballot must not.*


How to properly complete your Pennsylvania Mail-in Ballot




1 Complete your ballot in blue or black ink.





2 Insert your completed ballot into the small ballot envelope. If your ballot is not inside the small envelope, it will be considered a "naked ballot" and will **NOT** be counted. (Think of it as underwear for your ballot.)





3 Place your sealed ballot envelope into the larger return envelope. (This would be your ballot's pants.)





4 Seal the return envelope and sign and date the voter declaration. Your ballot is ready to be mailed.

@joshinphl

**Does not apply for in-person voting.*

QUESTIONS?

Get your questions about voting in PA answered here: <https://tinyurl.com/yyyyzuvda>

Need Immediate Support?

Call our Voter Assistance Hotline:

833-PAVOTES (833-728-6837)

- 1. Recurrent neural networks can discover structure in time**
- 2. Connectionism recap**
- 3. Voting resources**
- 4. Homework 2**