

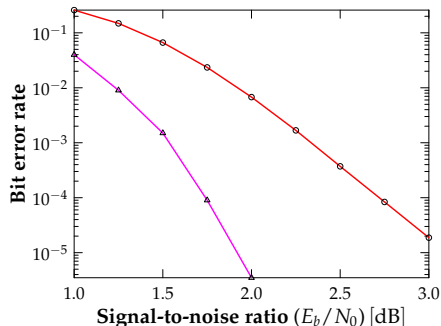
# List-Decoding of Polar Codes

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# Problem and goal

- Channel polarization is slow. For short to moderate code lengths, polar codes have disappointing performance.



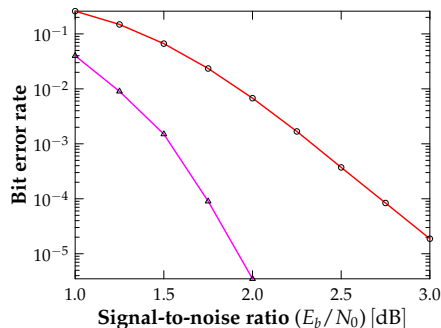
## Legend:

- successive cancellation,  $n = 2048$ ,  $k = 1024$
- △— LDPC (Wimax standard,  $n = 2304$ )

- In this talk, we present a generalization of the SC decoder which greatly improves performance at short code lengths.

# Avenues for improvement

From here onward, consider a polar code of length  $n = 2048$  and rate  $R = 0.5$ , optimized for a BPSK-AWGN channel with  $E_b/N_0 = 2.0$  dB.



Legend:

- successive cancellation,  $n = 2048, k = 1024$
- △— LDPC (Wimax standard,  $n = 2304$ )

- Why is our polar code under-performing?
  - Is the SC decoder under-performing?
  - Are the polar codes themselves weak at this length?

# A critical look at successive cancellation

## Successive Cancellation Decoding

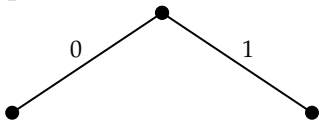
```
for  $i = 0, 1, \dots, n - 1$  do
  if  $\hat{u}_i$  is frozen then set  $\hat{u}_i$  accordingly;
  else
    if  $W_i(\mathbf{y}_0^{n-1}, \hat{u}_0^{i-1} | 0) > W_i(\mathbf{y}_0^{n-1}, \hat{u}_0^{i-1} | 1)$  then
      | set  $\hat{u}_i \leftarrow 0$ ;
    else
      | set  $\hat{u}_i \leftarrow 1$ ;
```

Potential weaknesses (interplay):

- Once an unfrozen bit is set, there is “no going back”. A bit that was set at step  $i$  can not be changed at step  $j > i$ .
- Knowledge of the value of future frozen bits is not taken into account.

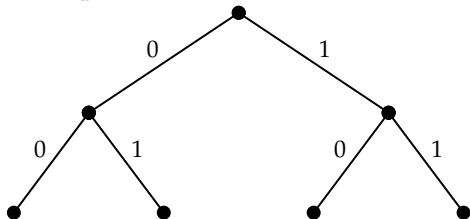
## List decoding of polar codes

**Key idea:** Each time a decision on  $\hat{u}_i$  is needed, split the current decoding path into two paths: **try both  $\hat{u}_i = 0$  and  $\hat{u}_i = 1$ .**



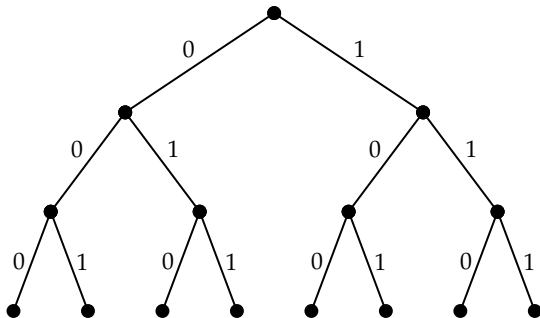
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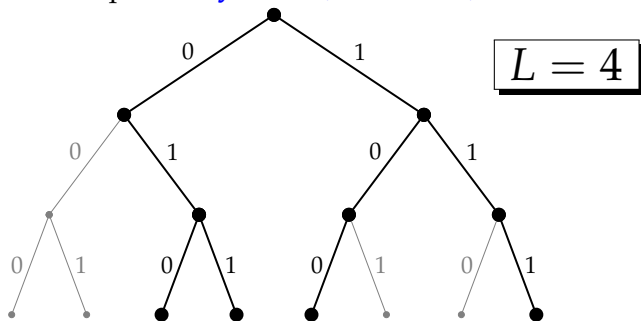
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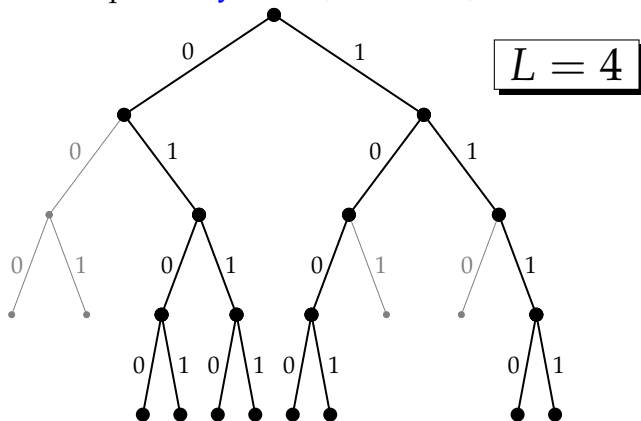


When the number of paths grows beyond a prescribed threshold  $L$ , discard the worst (least probable) paths, and keep only the  $L$  best paths.



## List decoding of polar codes

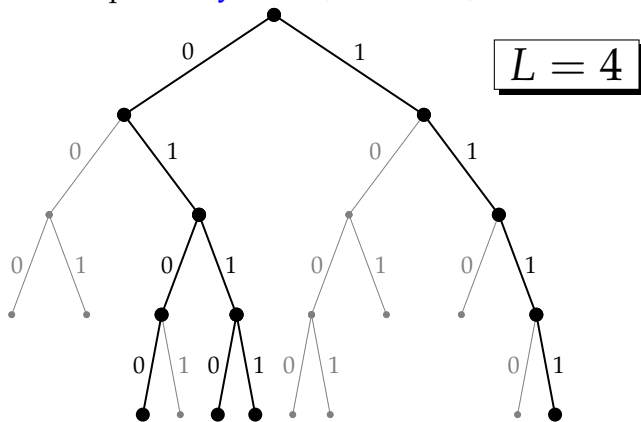
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## List decoding of polar codes

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When the number of paths grows beyond a prescribed threshold  $L$ , discard the worst (least probable) paths, and keep only the  $L$  best paths.

At the end, select the single **most likely** path.

## List-decoding: complexity issues

The idea of branching while decoding is not new. In fact a very similar idea was applied for Reed-Muller codes.

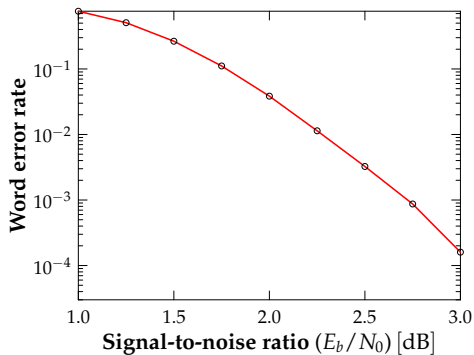
I. Dumer, K. Shabunov, Soft-decision decoding of Reed-Muller codes: recursive lists, *IEEE Trans. on Information Theory*, **52**, pp. 1260–1266, 2006.

### Our contribution

- We consider list decoding of polar codes.
- However, in a naive implementation, the time would be  $O(L \cdot n^2)$ .
- We show that this can be done in  $O(L \cdot n \log n)$  time and  $O(L \cdot n)$  space.

We will return to the complexity issue later. For now, let's see how decoding performance is affected.

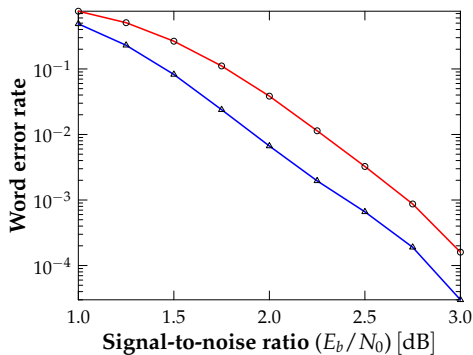
# Approaching ML performance



**Legend:**

—○—  $n = 2048, L = 1$

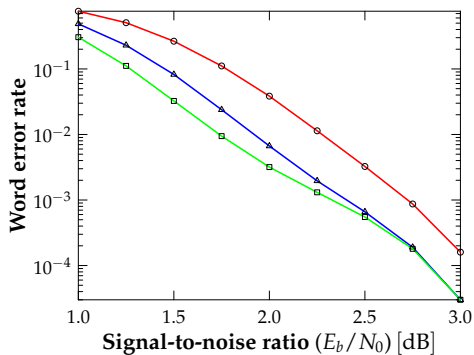
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## Legend:

- $n = 2048, L = 1$
- $n = 2048, L = 2$

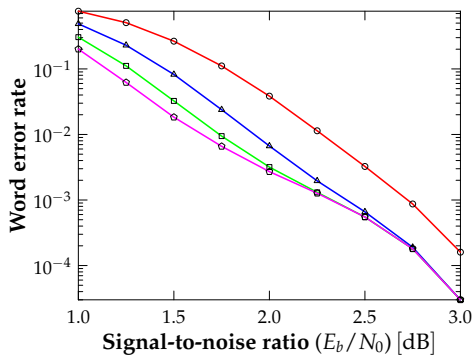
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## Legend:

- $n = 2048, L = 1$
- $n = 2048, L = 2$
- $n = 2048, L = 4$

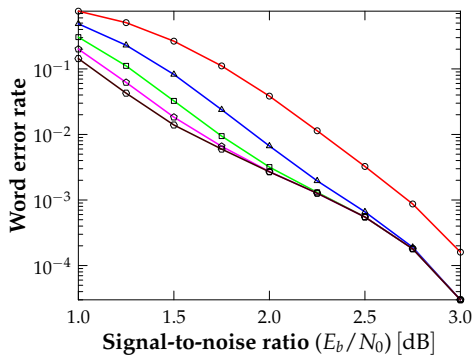
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## Legend:

- $n = 2048, L = 1$
- $n = 2048, L = 2$
- $n = 2048, L = 4$
- $n = 2048, L = 8$

# Approaching ML performance

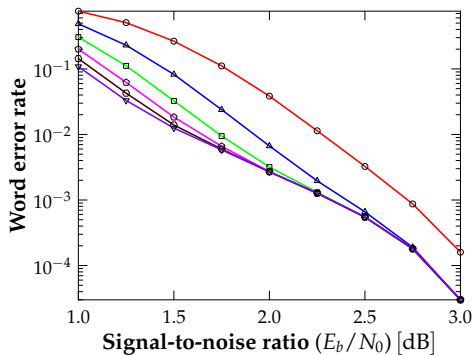


## Legend:

- $n = 2048, L = 1$
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- $n = 2048, L = 8$
- $n = 2048, L = 16$



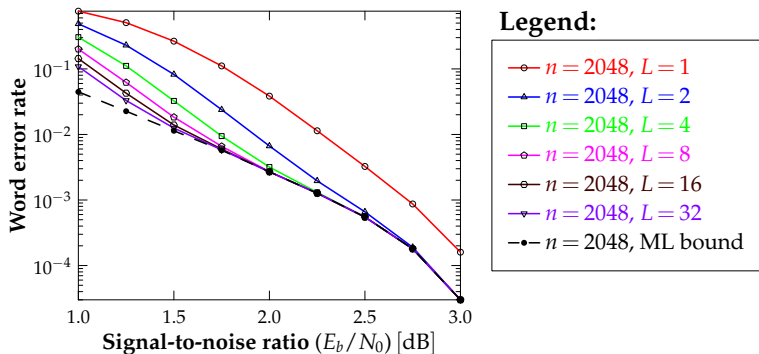
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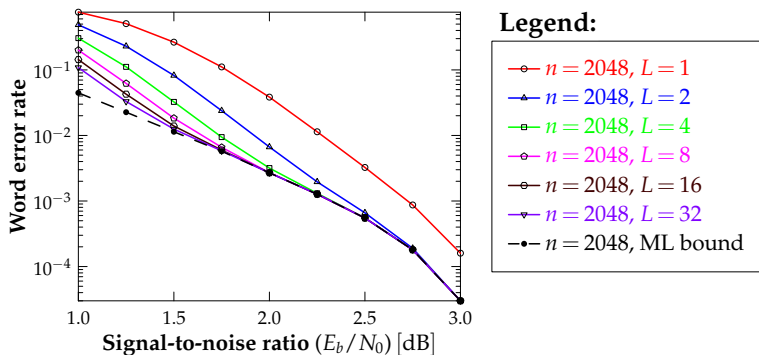
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- $n = 2048, L = 16$
- $n = 2048, L = 32$

# Approaching ML performance



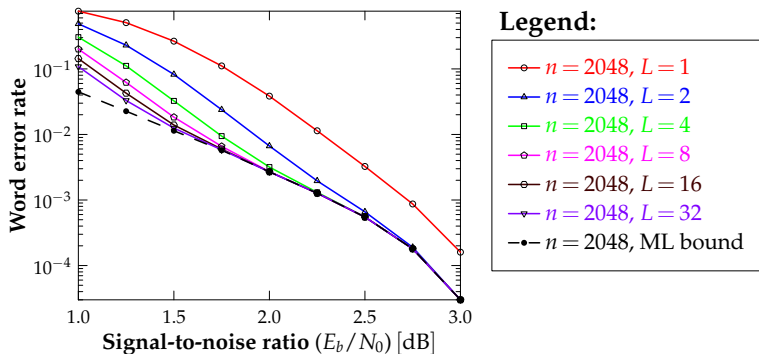
- List-decoding performance quickly approaches that of maximum-likelihood decoding as a function of list-size.

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- Good: our decoder is essentially optimal.
- Bad: Still not competitive with LDPC...

# Approaching ML performance



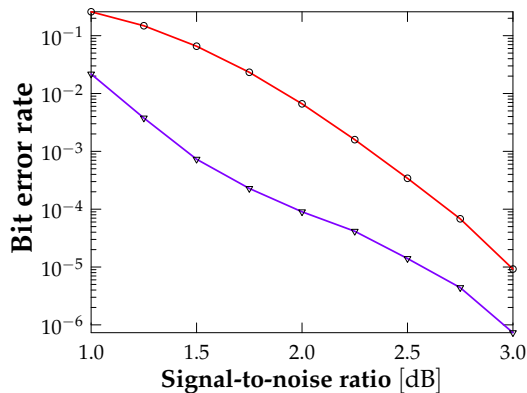
- List-decoding performance quickly approaches that of maximum-likelihood decoding as a function of list-size.
- Good: our decoder is essentially optimal.
- Bad: Still not competitive with LDPC...
- Conclusions: Must somehow “fix” the polar code.

## A simple concatenation scheme

- Recall that the last step of decoding was “pick the most likely codeword from the list”.
- An error: the transmitted codeword is not the **most likely** codeword in the list.
- However, very often, the transmitted codeword is still a **member** of the list.
- We need a “genie” to single-out the transmitted codeword.
- Idea: Let there be  $k + r$  unfrozen bits. Of these,
  - Use the first  $k$  bits to encode **information**.
  - Use the last  $r$  unfrozen bits to encode the **CRC value** of the first  $k$  bits.
  - Pick the most probable codeword on the list **with correct CRC**.

# Approaching LDPC performance

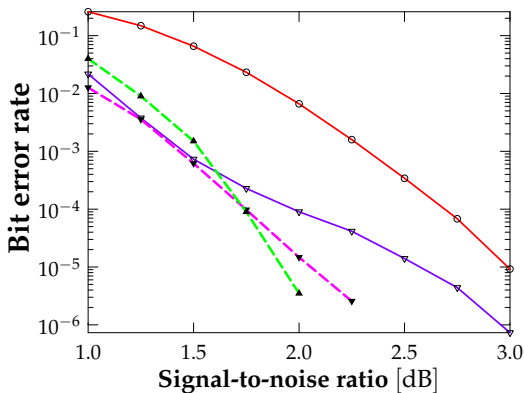
Simulation results for a polar code of length  $n = 2048$  and rate  $R = 0.5$ , optimized for a BPSK-AWGN channel with  $E_b/N_0 = 2.0$  dB.



—○— Successive cancellation  
—▽— List-decoding ( $L = 32$ )

# Approaching LDPC performance

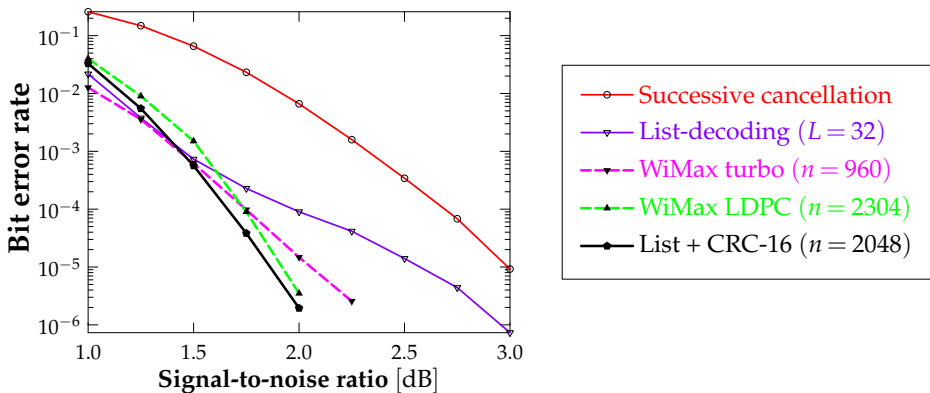
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- Successive cancellation
- List-decoding ( $L = 32$ )
- WiMax turbo ( $n = 960$ )
- WiMax LDPC ( $n = 2304$ )

# Approaching LDPC performance

Simulation results for a polar code of length  $n = 2048$  and rate  $R = 0.5$ , optimized for a BPSK-AWGN channel with  $E_b/N_0 = 2.0$  dB.



*Polar codes (+CRC) under list decoding are competitive with the best LDPC codes at lengths as short as  $n = 2048$ .*

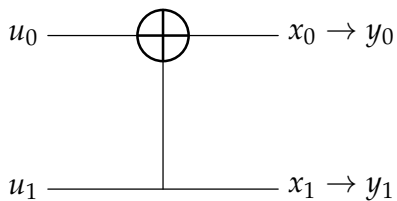


# Quadratic complexity of list decoding

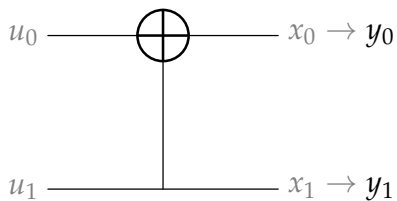
## Naive implementation recap

- In a naive implementation, the decoding paths are **independent**. They don't share information.
- Each decoding path has a set of variables associated with it. For example, at stage  $i$ , each decoding path must remember the values of the bits  $\hat{u}_0, \hat{u}_1, \dots, \hat{u}_{i-1}$ .
- It turns out (as we shall see) that each decoding path has  $\Theta(n)$  memory associated with it.
- When a path is split in two, one decoding path is left with the **original** variables while the other must be handed a **copy** of them.
- Each copy operation takes  $O(n)$  time.
- Thus, the overall time complexity is  $O(L \cdot n^2)$ .

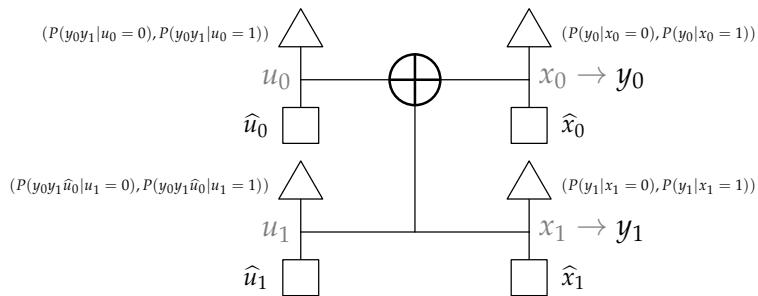
## A closer look at successive cancellation





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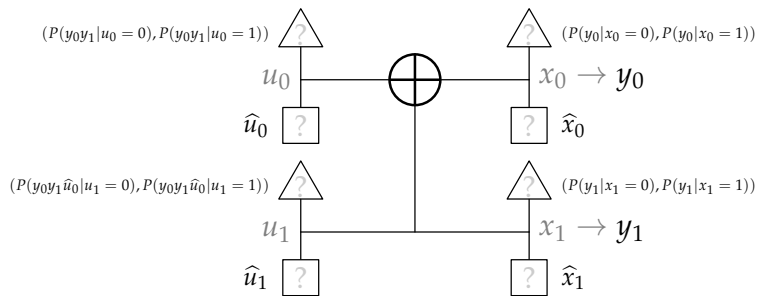
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 probability pair variable

 boolean variable (bit)

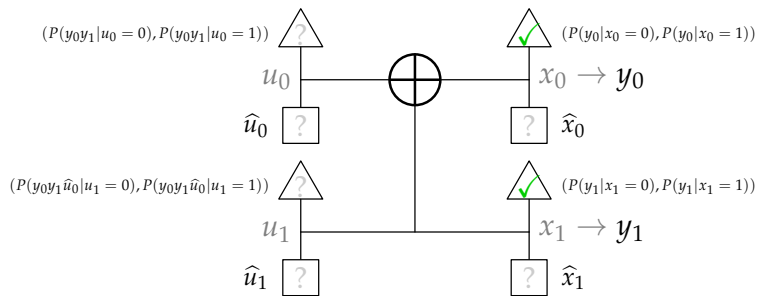
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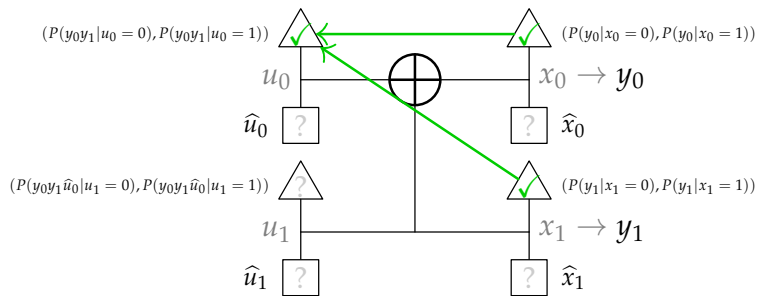
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
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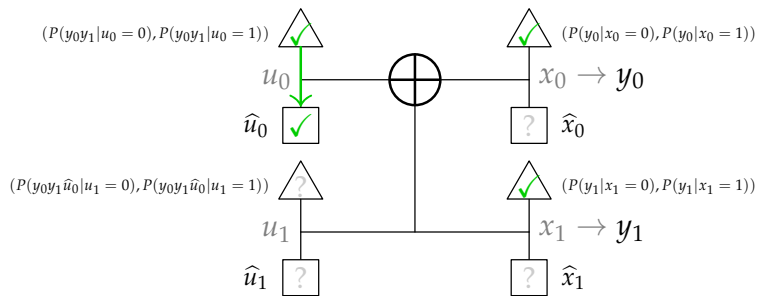
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
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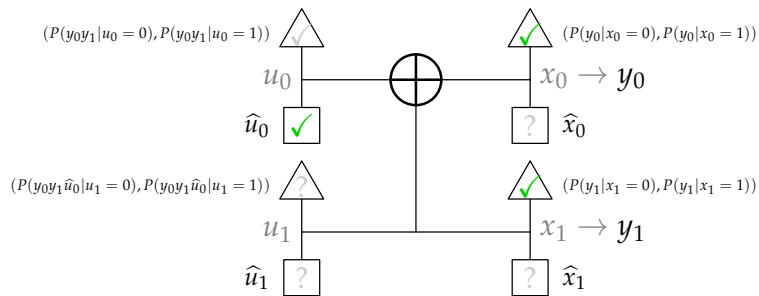


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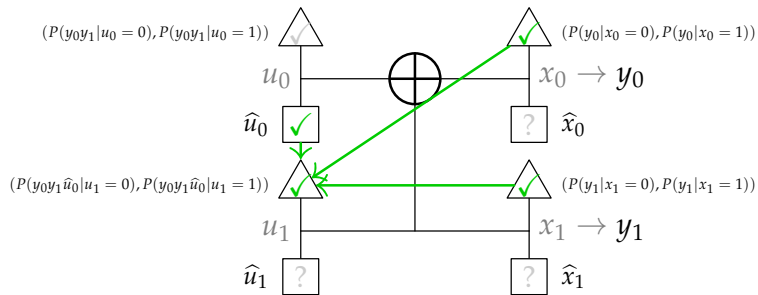
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
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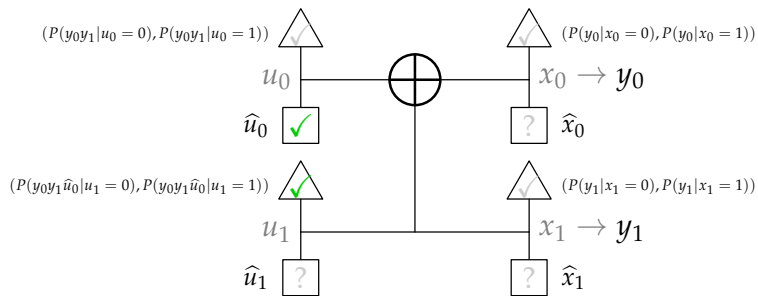
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
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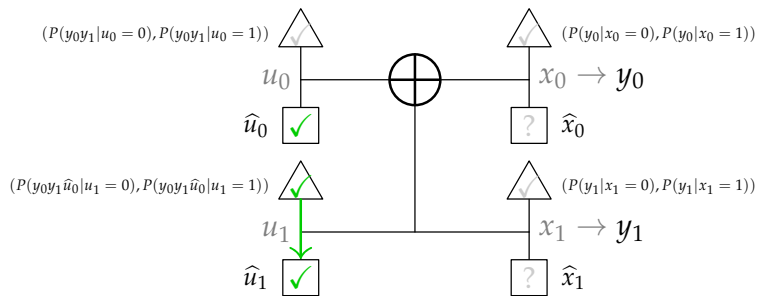
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
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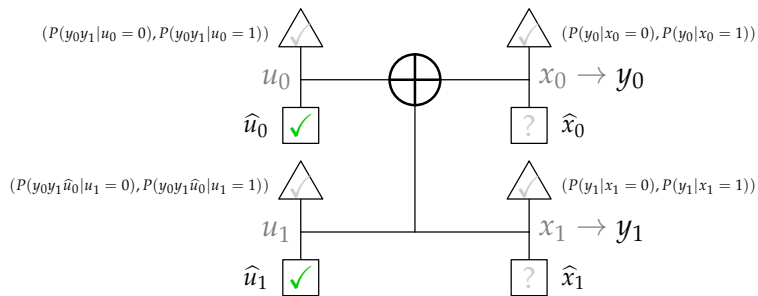
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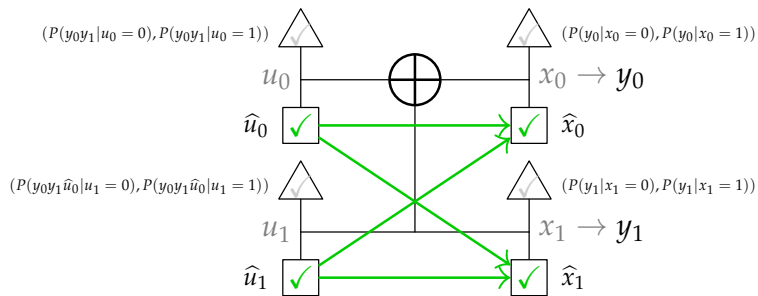
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
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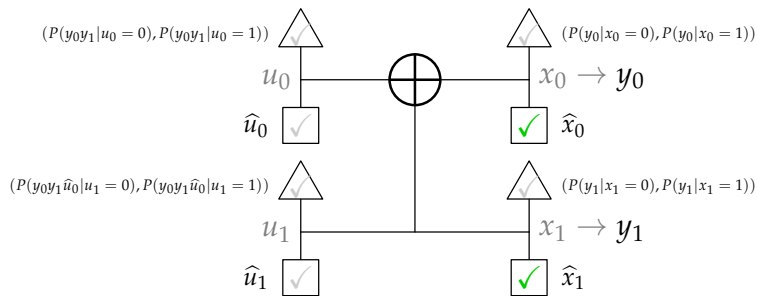
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
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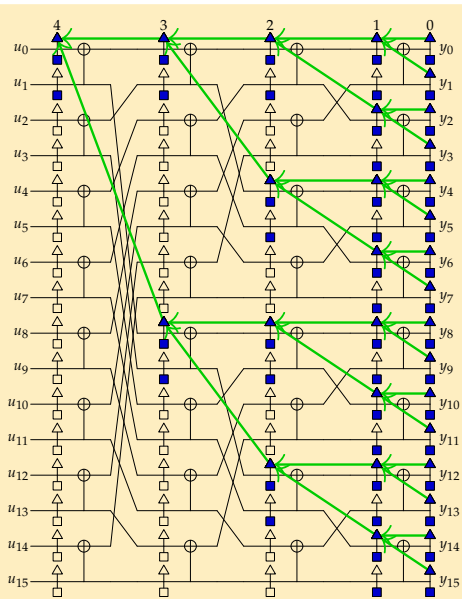
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# A larger example

## Key point

The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .

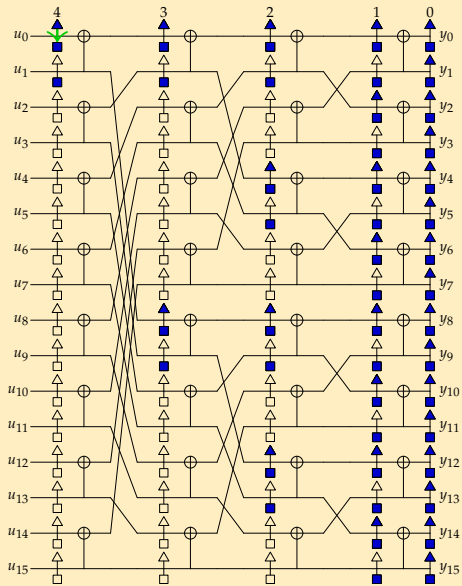




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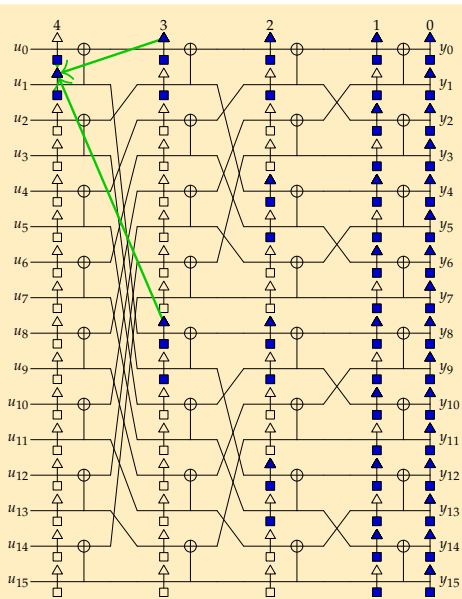
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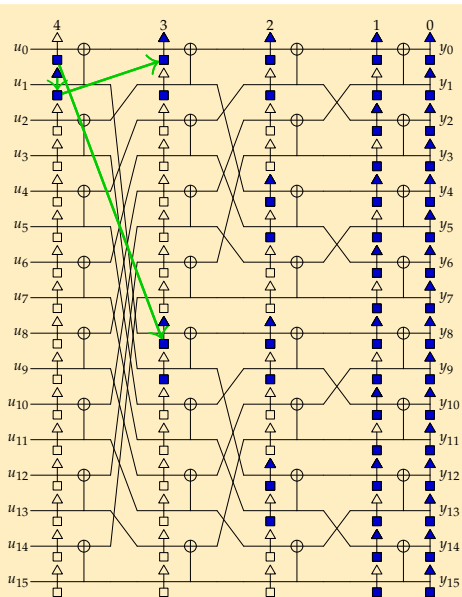
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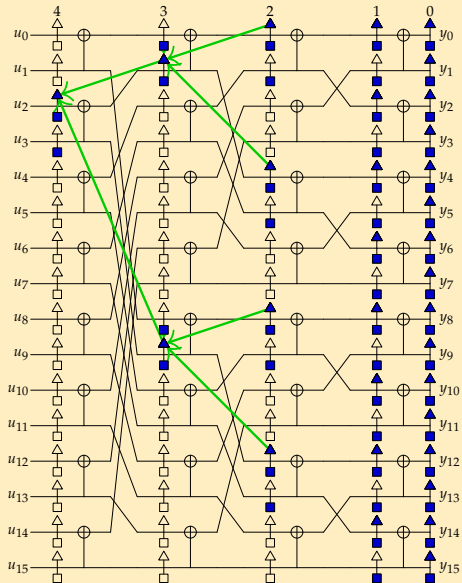
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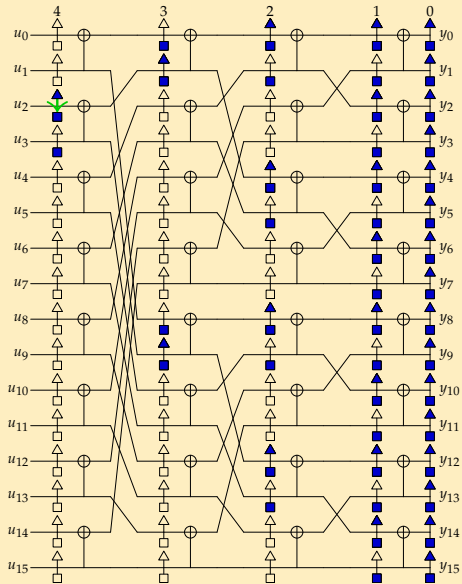
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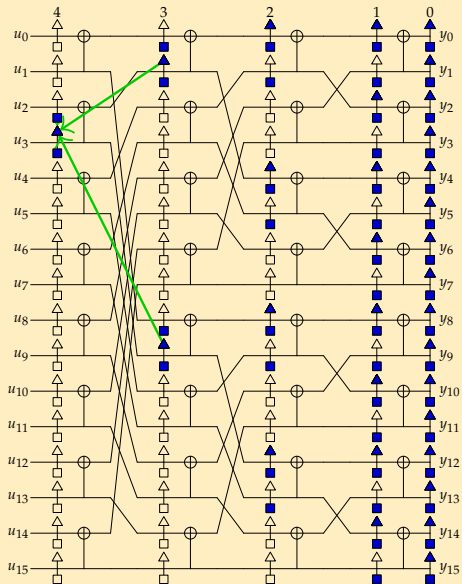
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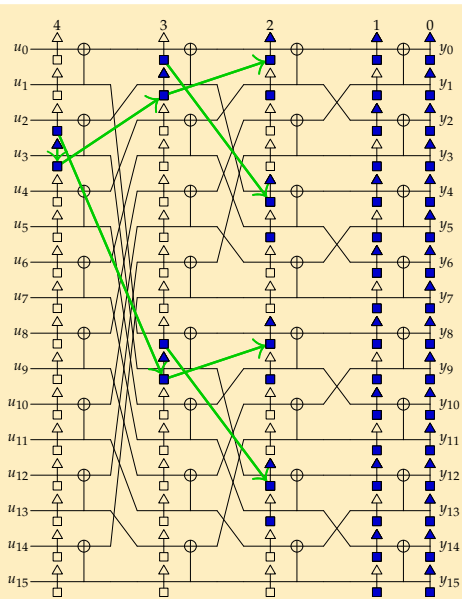
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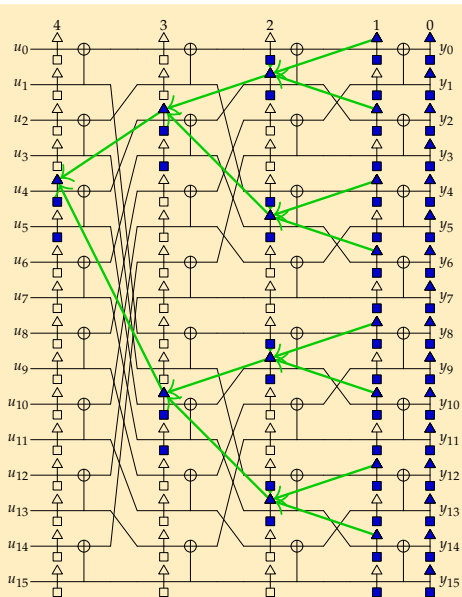
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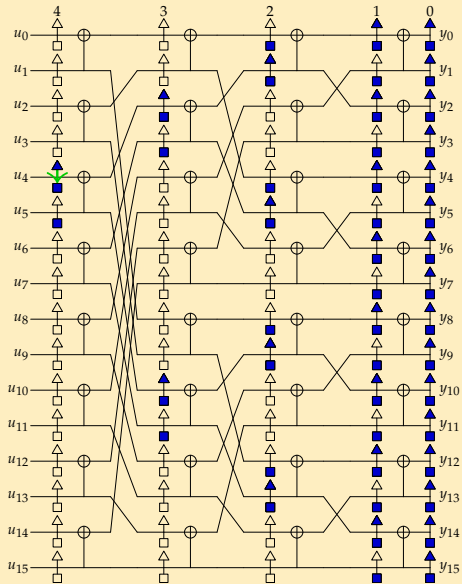




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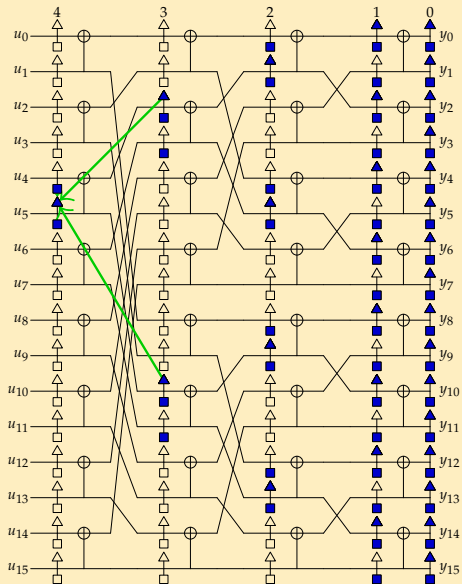
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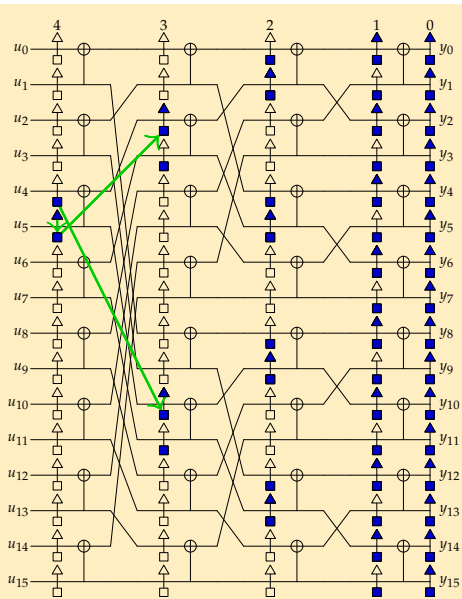
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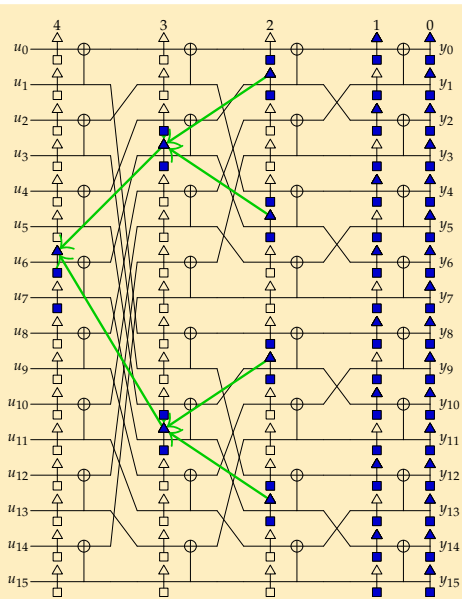
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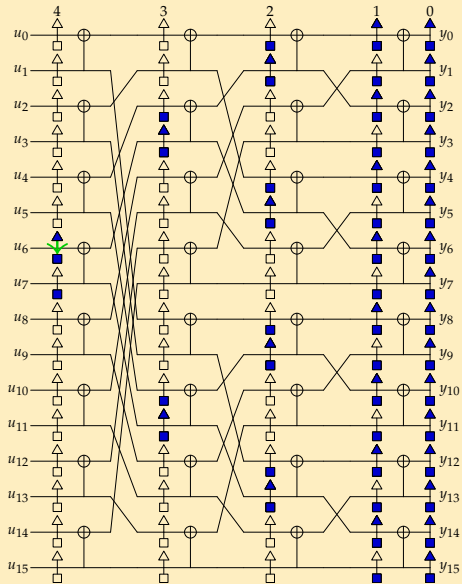
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## Key point

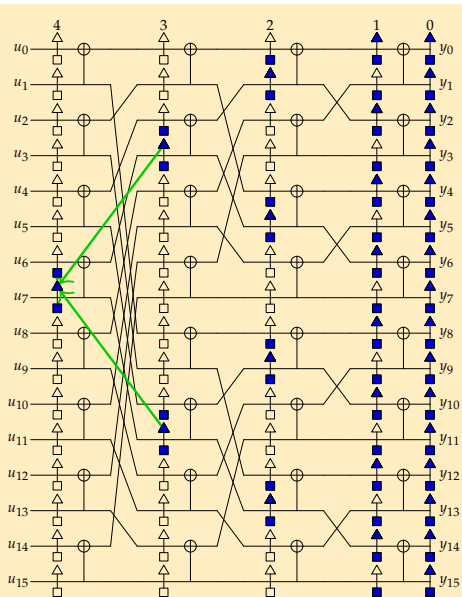
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

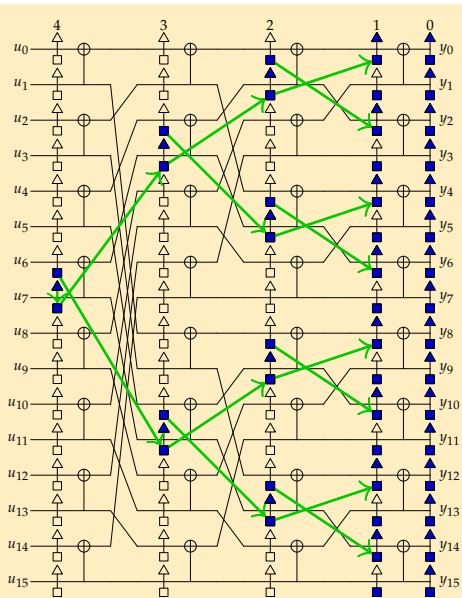
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

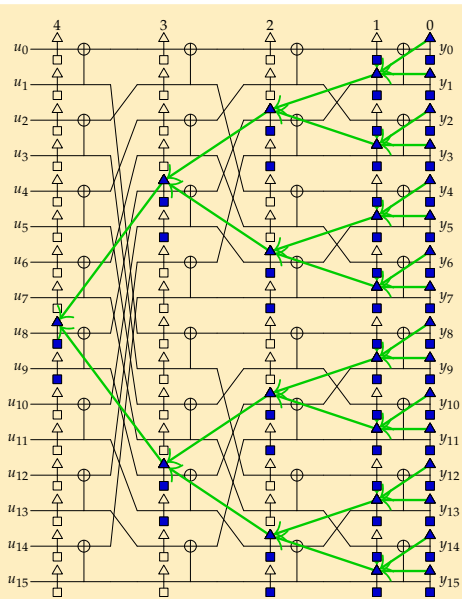
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .

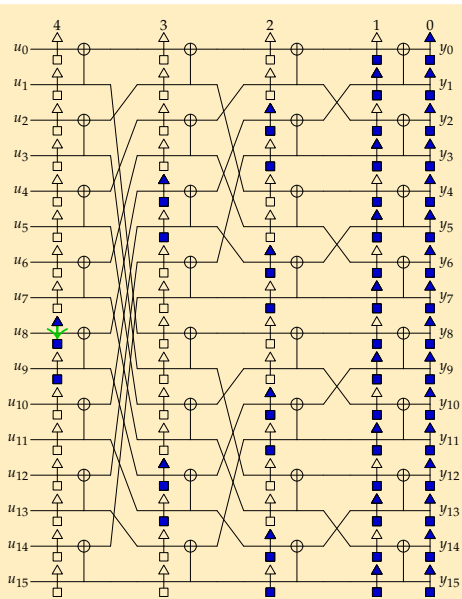




# A larger example

## Key point

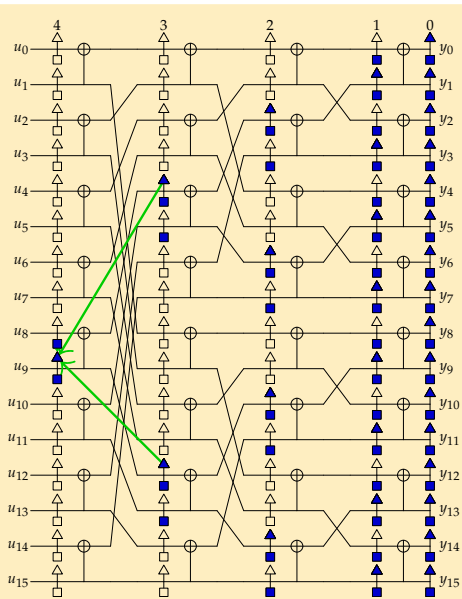
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

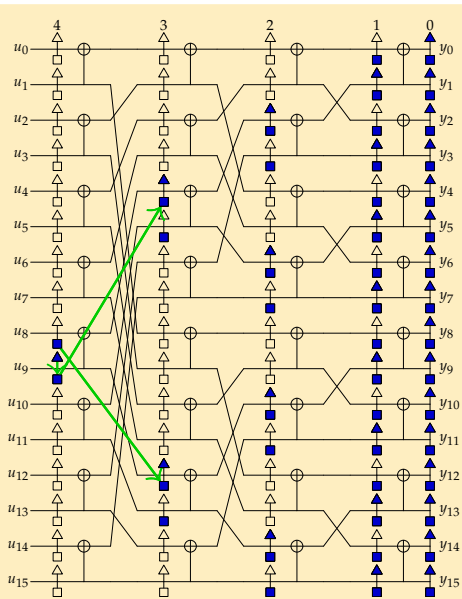
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

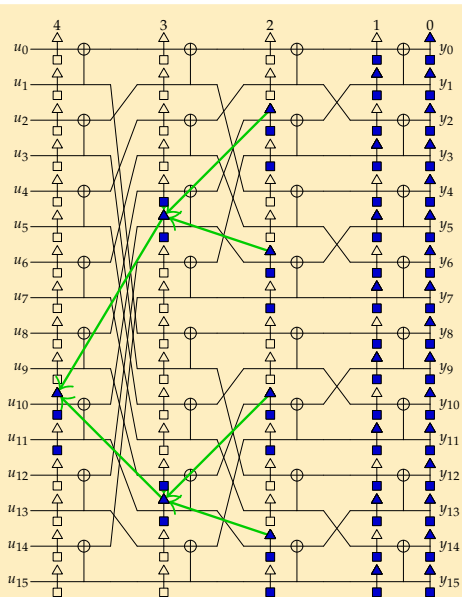
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

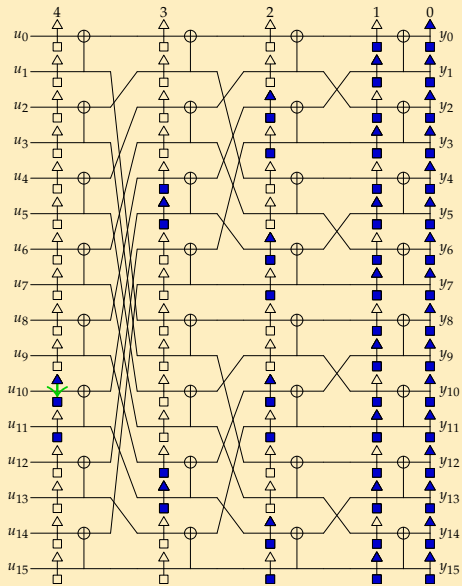
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

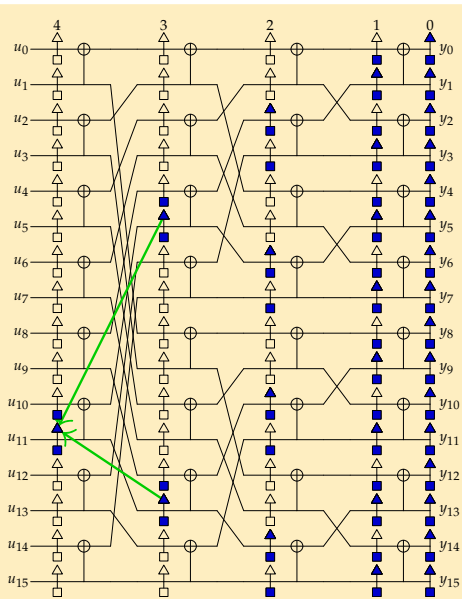
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

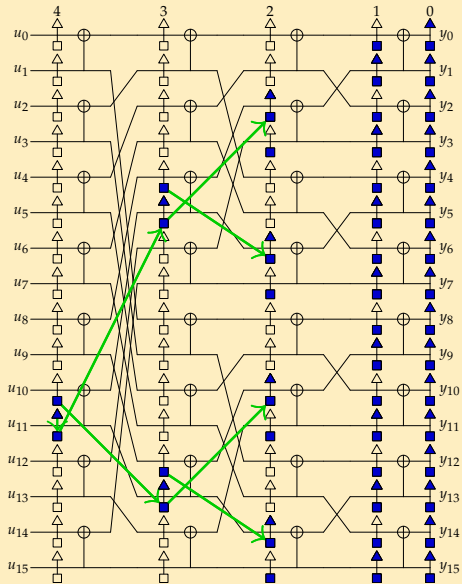
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

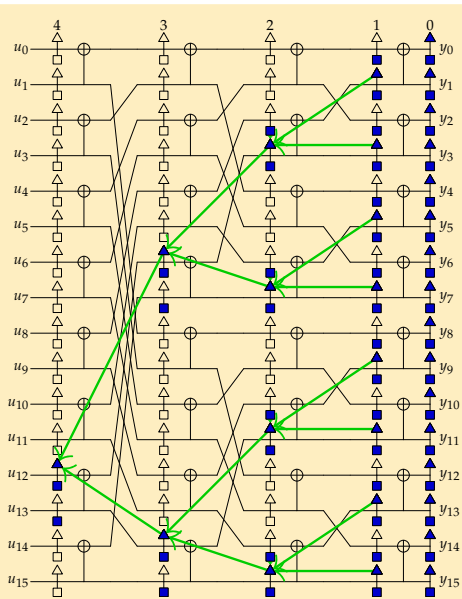
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .

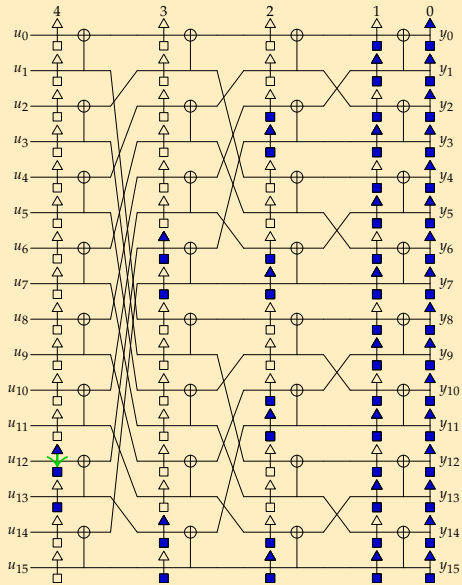




# A larger example

## Key point

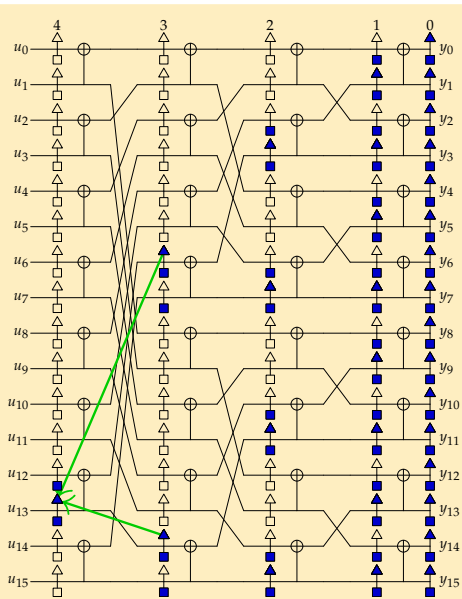
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

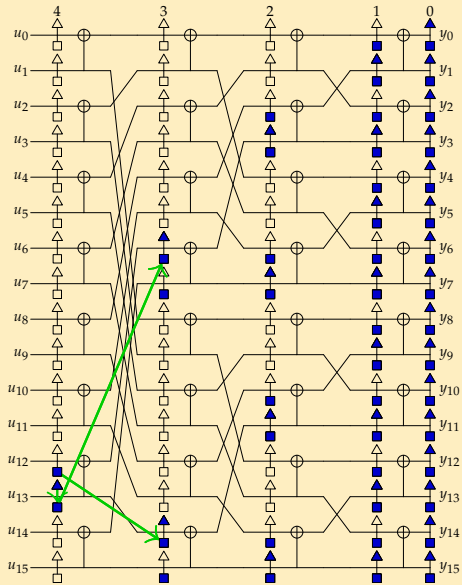
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

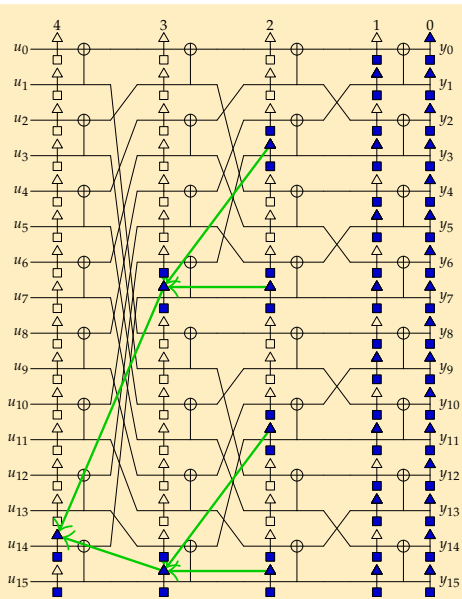
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

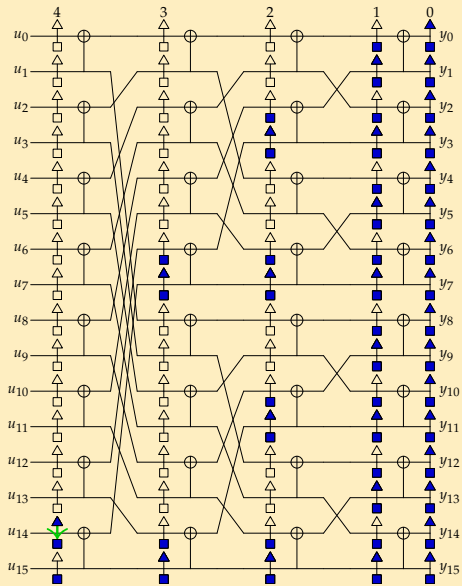
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

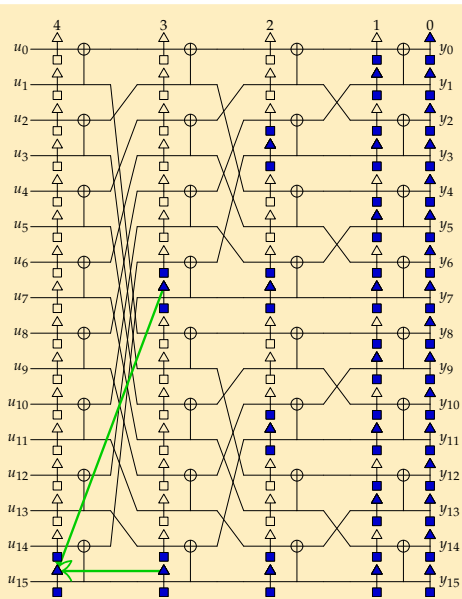
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

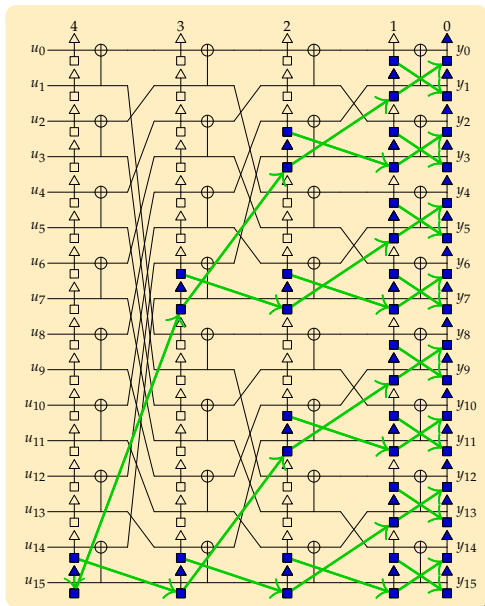
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

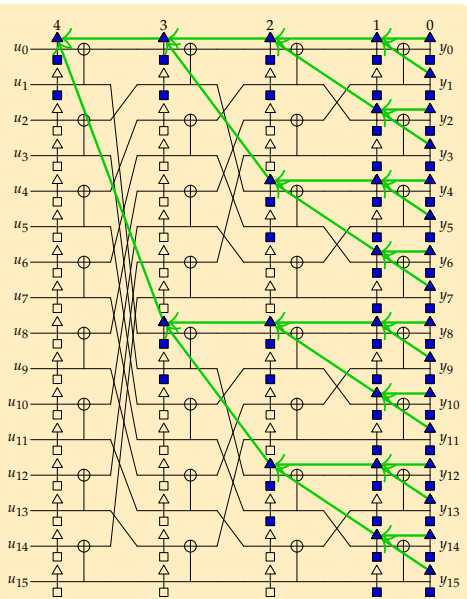
The memory needed to hold the variables at level  $t$  is  $O(n/2^t)$ .



# A larger example

## Key point

Level  $t$  is written to once every  $O(2^{m-t})$  stages.

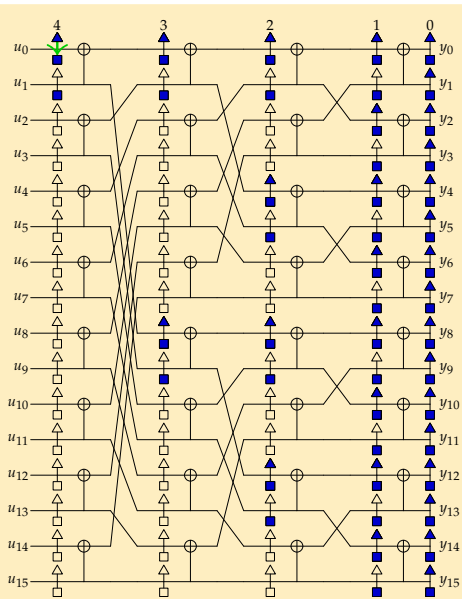




# A larger example

## Key point

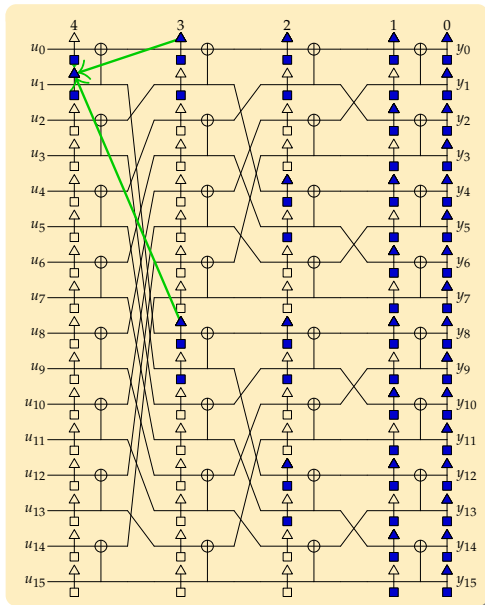
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

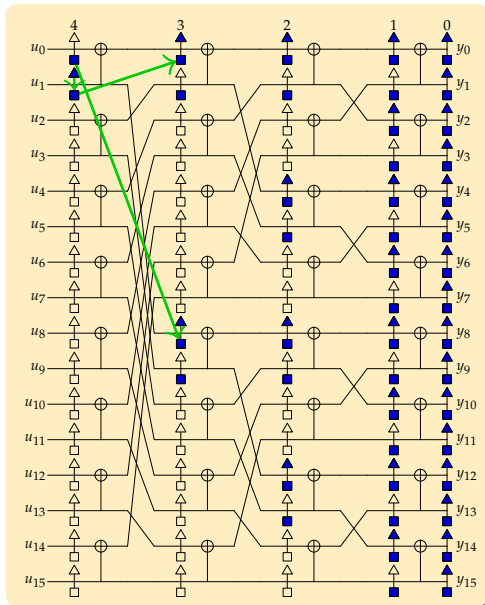
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

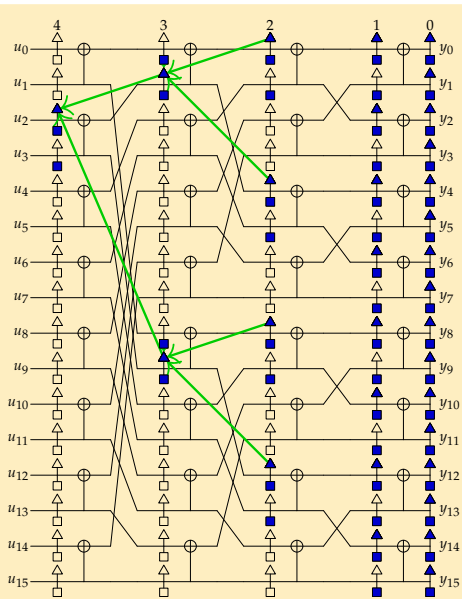
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

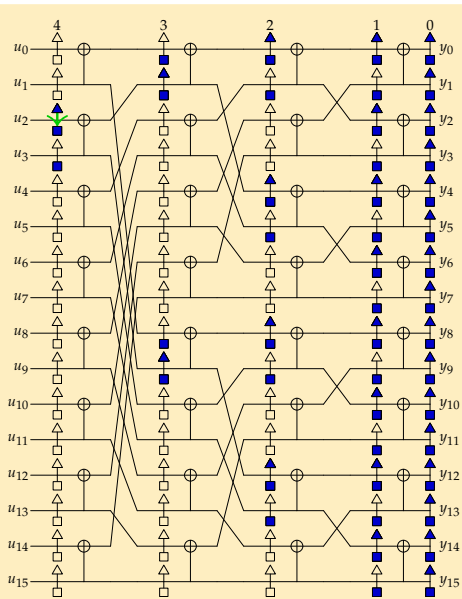
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

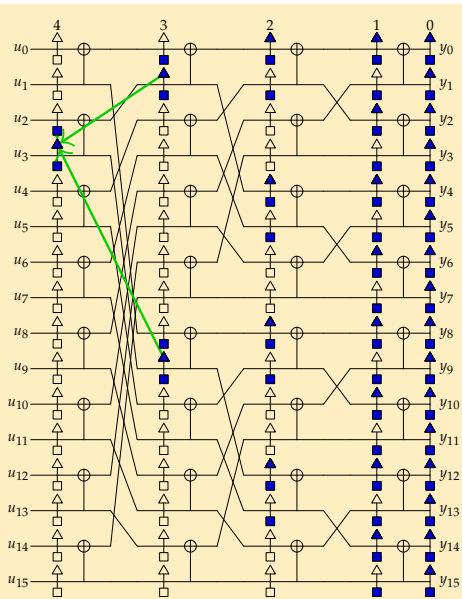
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

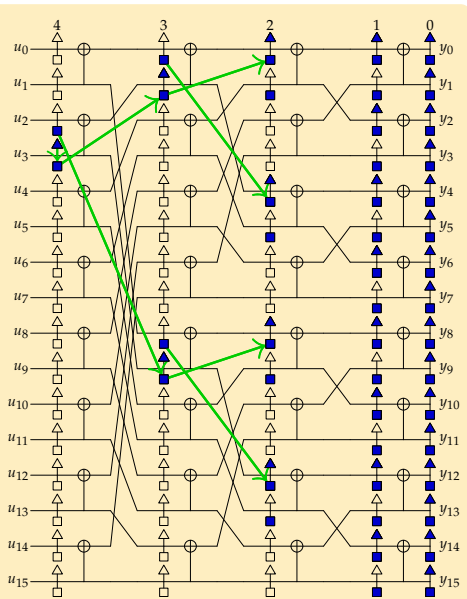
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

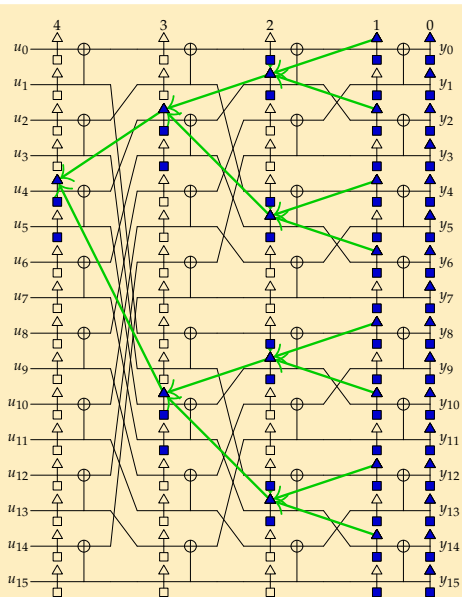
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

Level  $t$  is written to once every  $O(2^{m-t})$  stages.

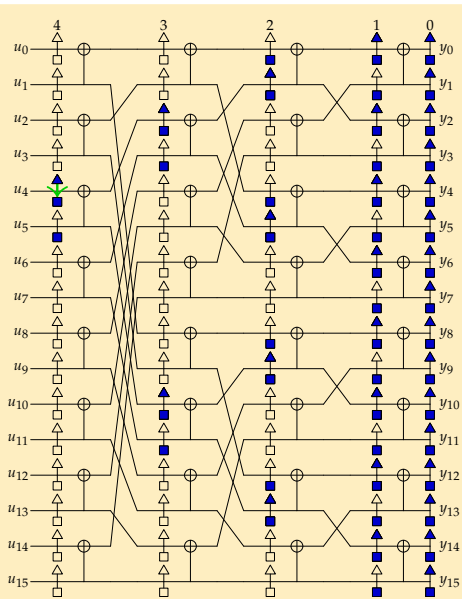




# A larger example

## Key point

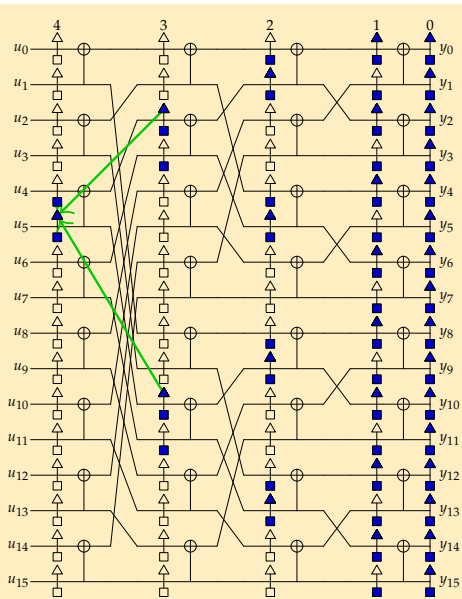
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

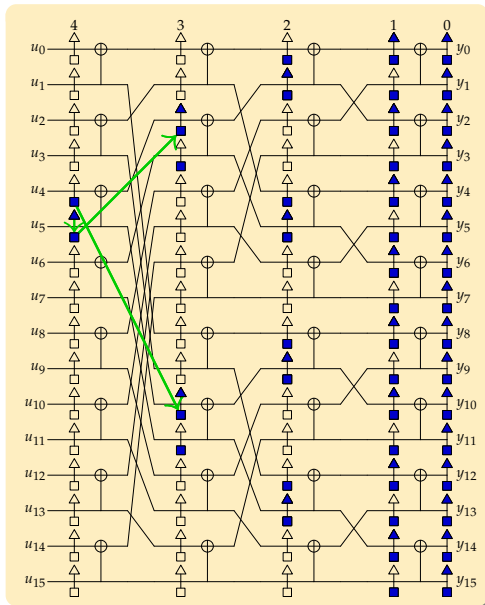
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

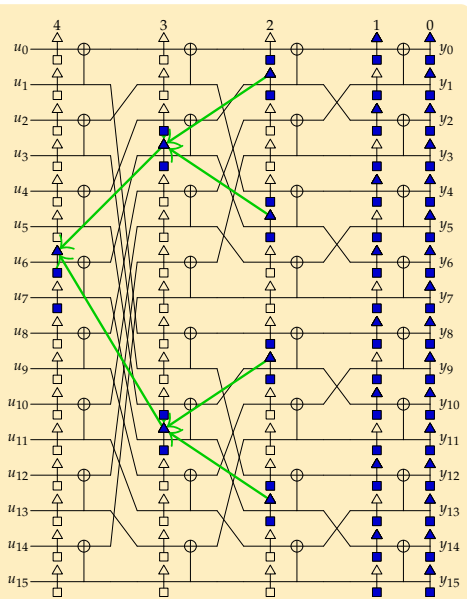
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

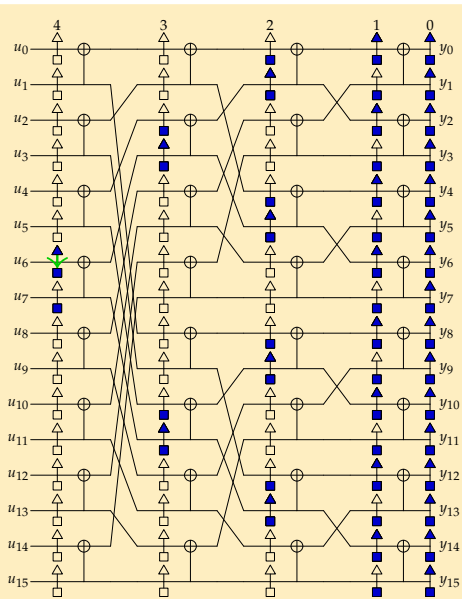
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

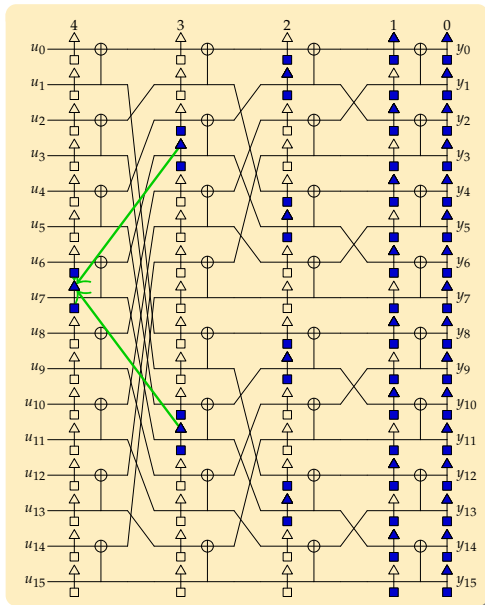
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

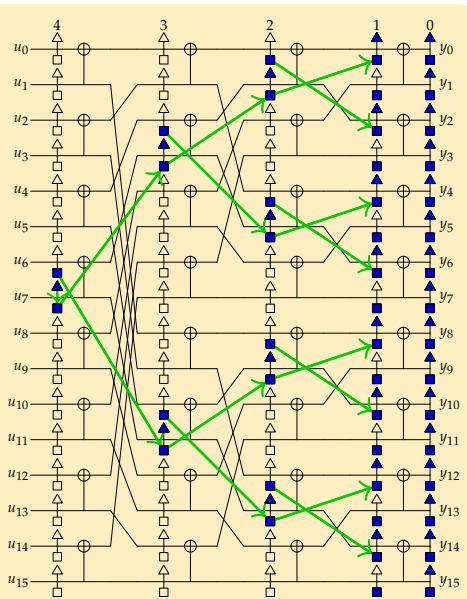
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

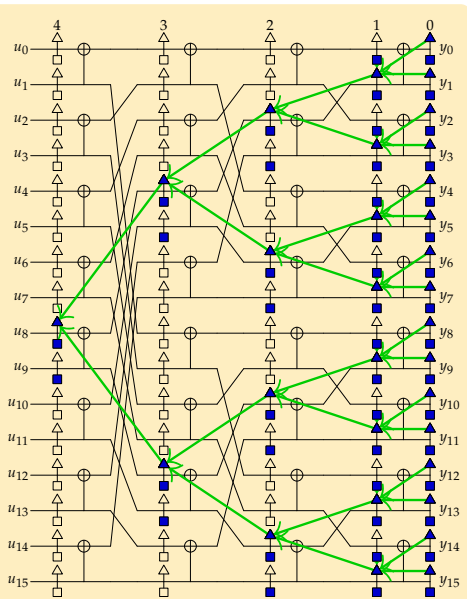
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

Level  $t$  is written to once every  $O(2^{m-t})$  stages.

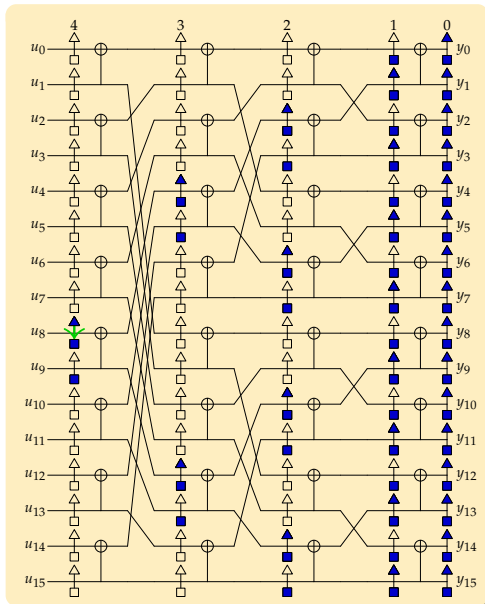




# A larger example

## Key point

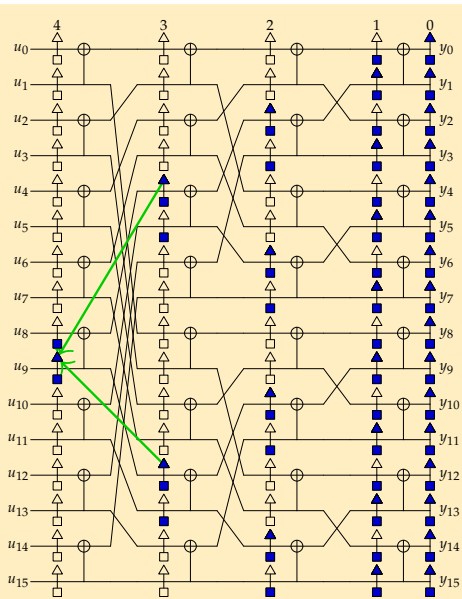
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

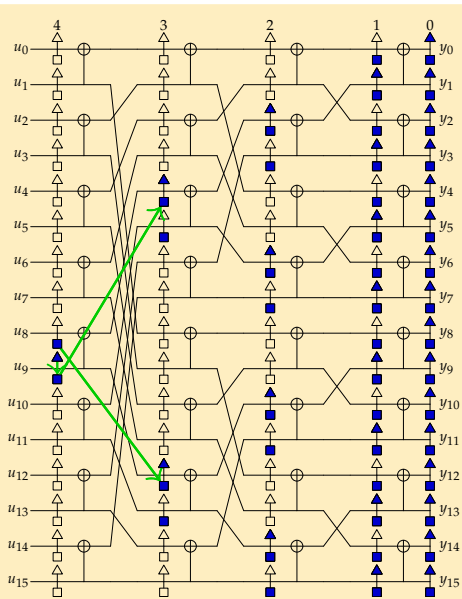
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

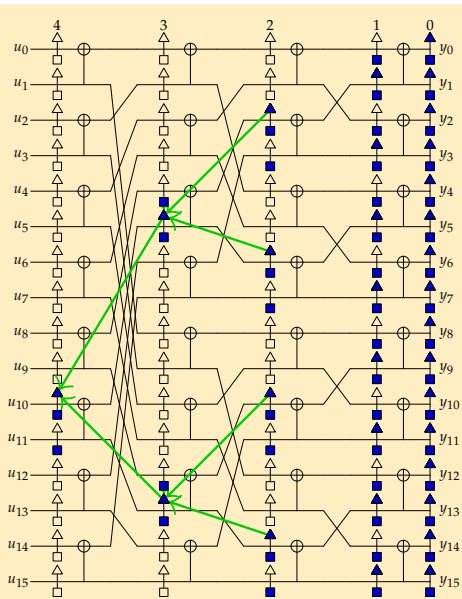
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

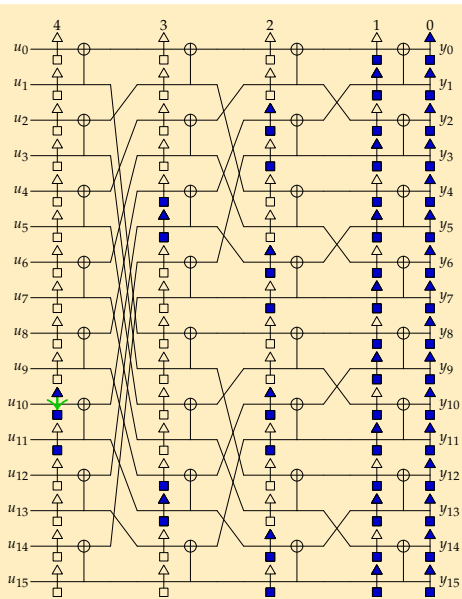
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

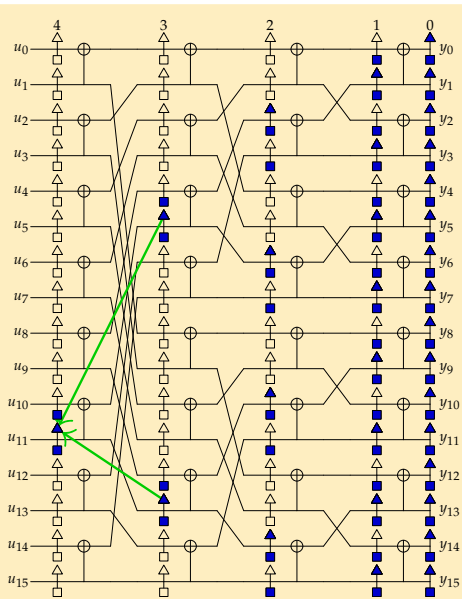
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

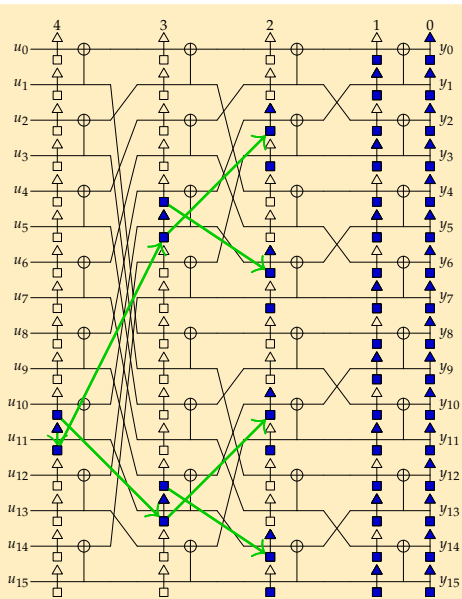
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

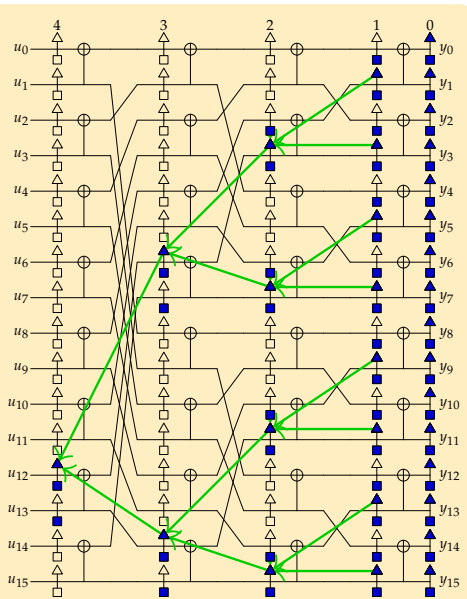
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

Level  $t$  is written to once every  $O(2^{m-t})$  stages.

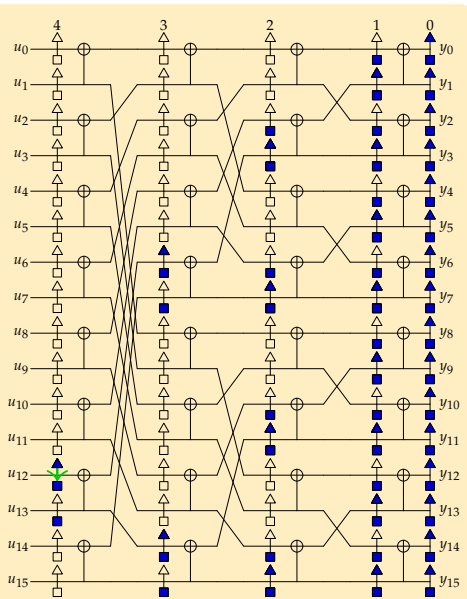




# A larger example

## Key point

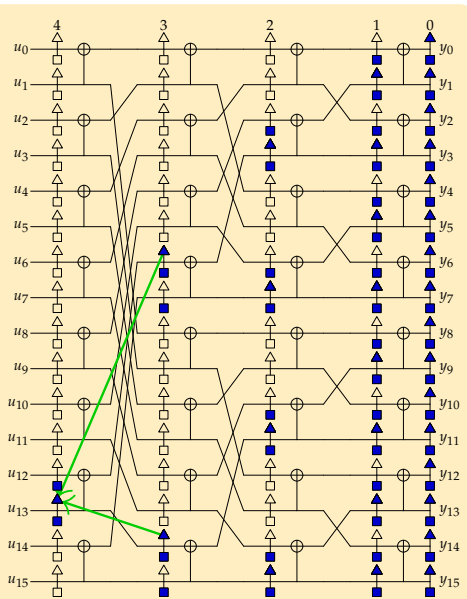
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

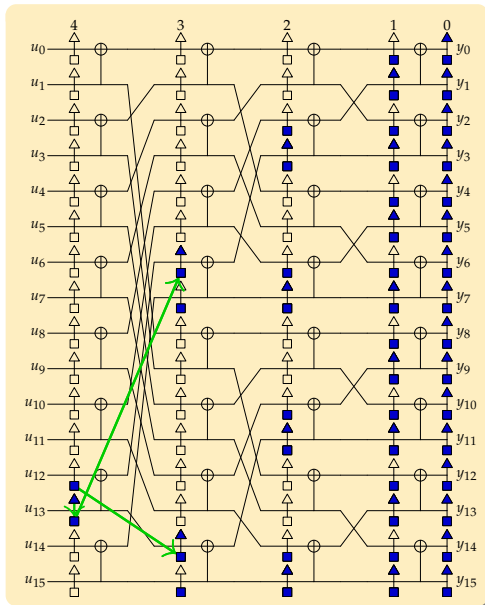
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

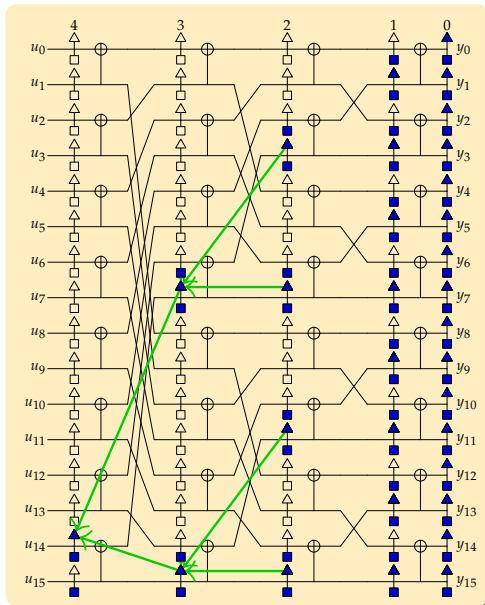
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

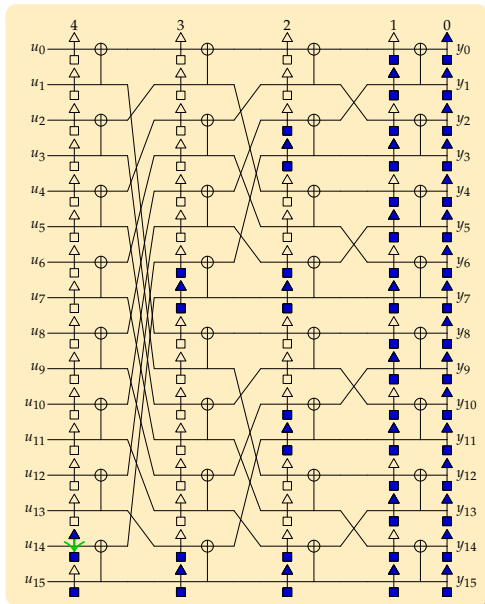
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

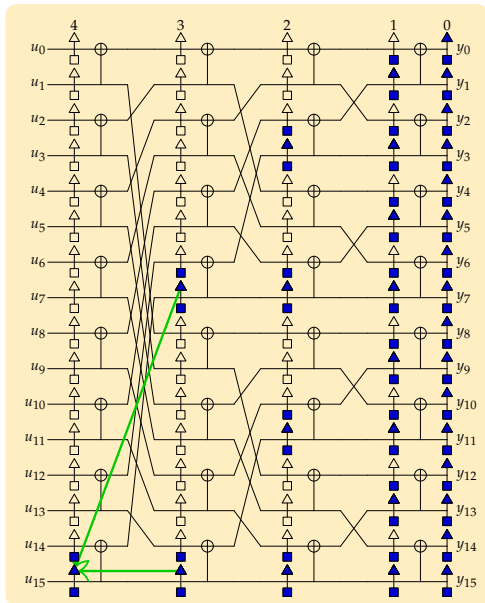
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

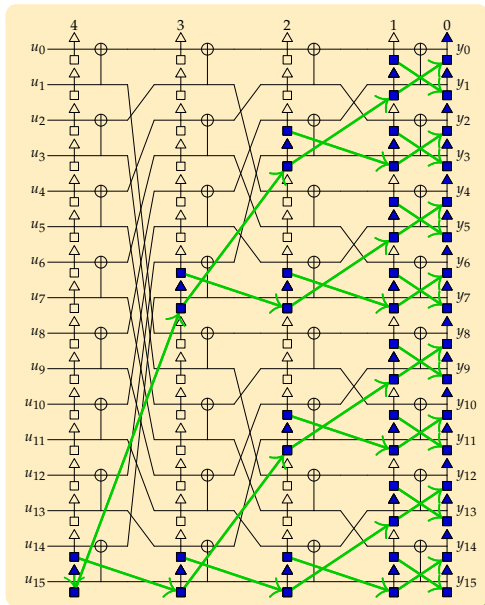
Level  $t$  is written to once every  $O(2^{m-t})$  stages.



# A larger example

## Key point

Level  $t$  is written to once every  $O(2^{m-t})$  stages.



## Application to list decoding

- In a naive implementation, at each split we make a copy of the variables.
- We can do better:
  - At each split, **flag** the corresponding variables as belonging to **both** paths.
  - Give each path a **unique** variable (make a **copy**) only before that variable will be **written** to.
  - If a path is killed, **deflag** its corresponding variables.
- Thus, instead of wasting a lot of time on copy operations at each stage, we **typically** perform only a **small** number of copy operations.

This was a mile high view, there are many details to be filled (book-keeping, data structures), but the end result is a running time of  $O(L \cdot n \log n)$  with  $O(L \cdot n)$  memory requirements.

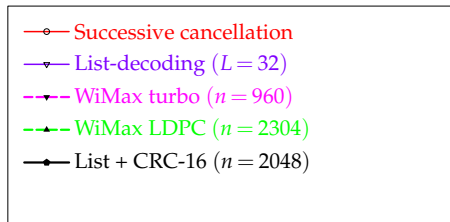
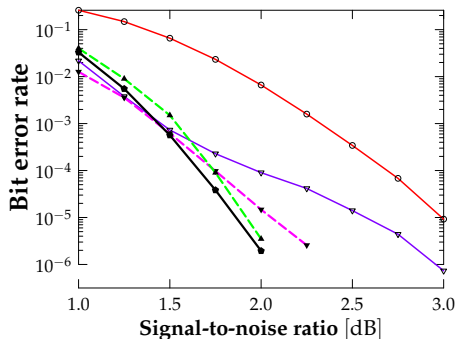


# Very recent results

Gross and Sarkis (MacGill University) have recently attained the following results.

- Full **independent** verification of our simulation data.
- Further improvement of performance using **systematic** polar codes.

E. Arıkan, Systematic polar codes, *IEEE Comm. Letters*, accepted for publication.



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