

# The Spectral Relation between the Cube-Connected Cycles and the Shuffle-Exchange Network

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<sup>1</sup>Pattern Recognition Lab (CS 5)

<sup>2</sup>Hardware-Software Co-Design (CS 12),

<sup>3</sup>Artificial Intelligence (CS 8)

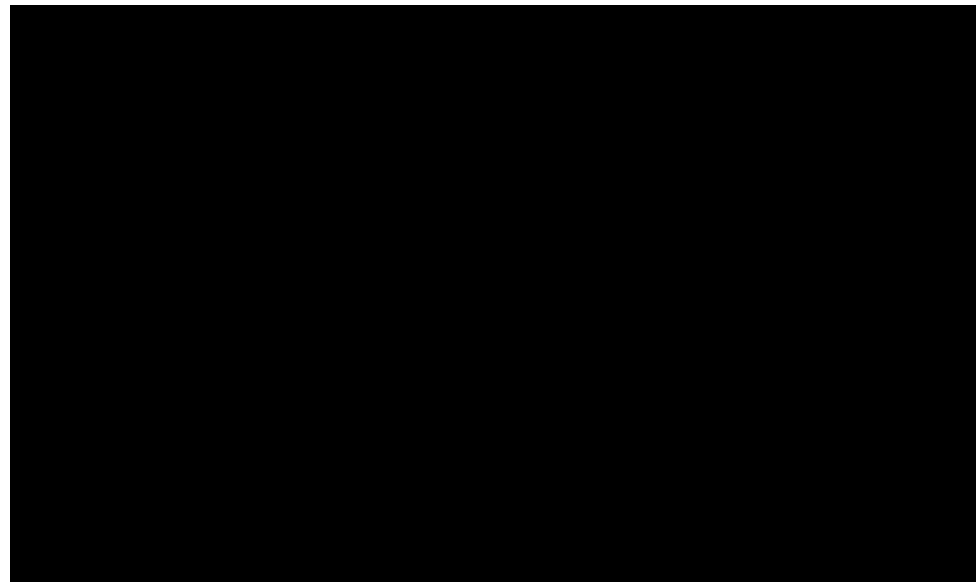


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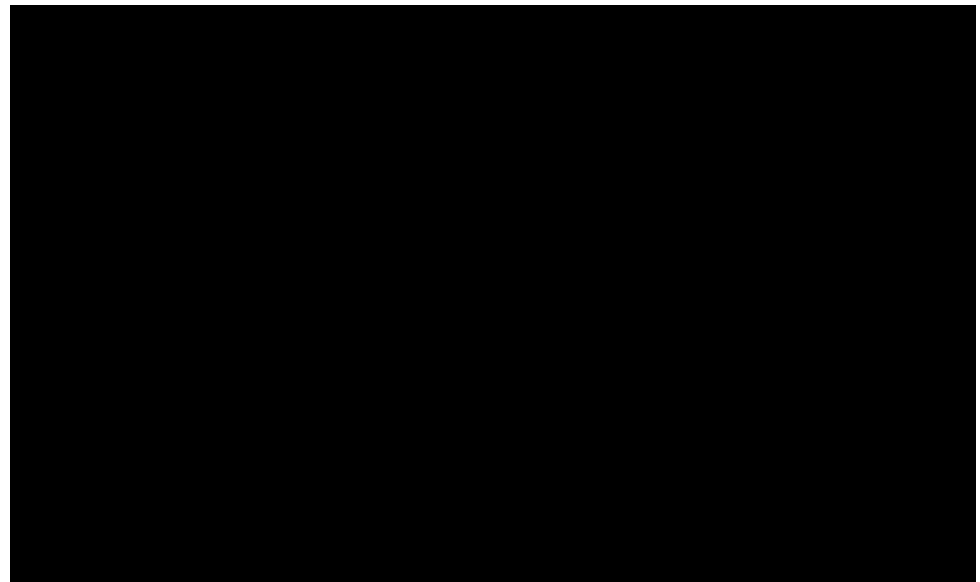
## Spectra of Networks

- Spectral set ( ) contains information on
  - Network throughput
  - Fault-tolerance
  - ...
- Known spectra:
  - Linear array
  - Cycle
  - Hypercube
  - Butterfly
  - De Bruijn



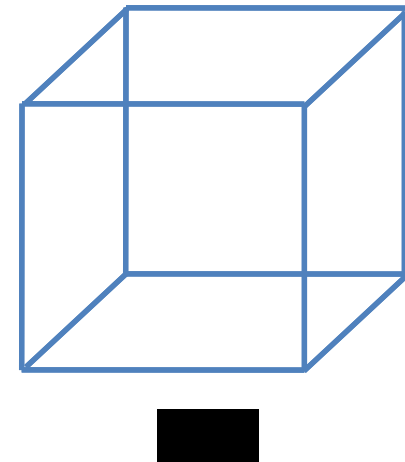
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- $\text{SpS}(\text{CCC}(d)) = ?$
- $\text{SpS}(\text{SE}(d)) = ?$



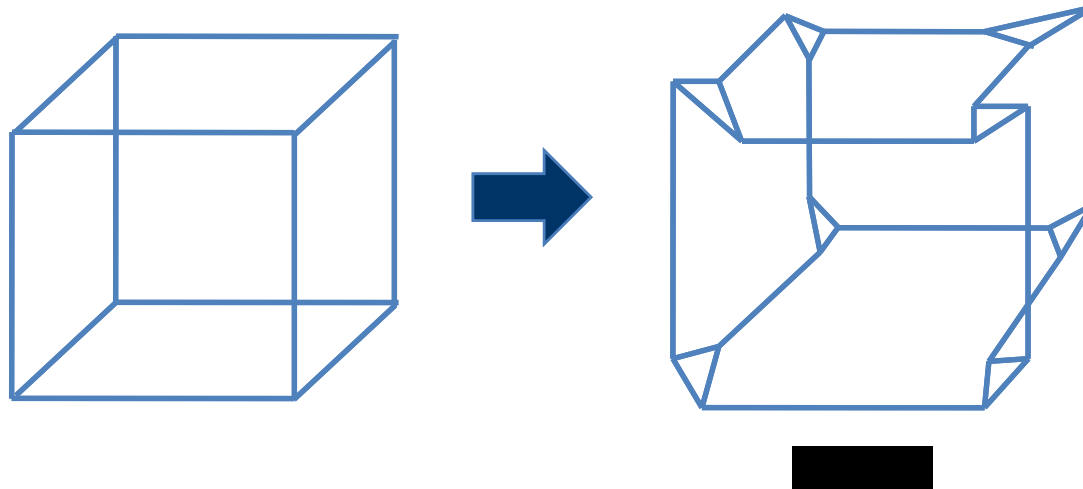
## Parallel Computations on the Hypercube

- $n$ -dimensional Hypercube: a popular architecture for parallel computations
- Versatile connection structure for many algorithms
  - Finite differences
  - Spanning trees
  - Connected components
  - Sorting networks
  - Nearest neighbor search
  - Chinese remaindering
  - ...
- Drawback: degree  $n$  in every node



## Hypercubic: Cube-Connected Cycles Network

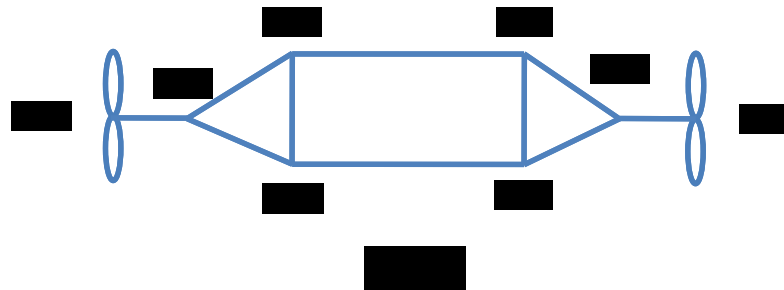
- $n$ -dimensional Cube-Connected Cycles  $CCC_n$ :  $2^n$  nodes
- Starting from a hypercube, every node is replaced by a cycle



- Runs hypercube algorithms with constant slowdown
- Constant degree  $3$

## Hypercubic: Shuffle-Exchange Network

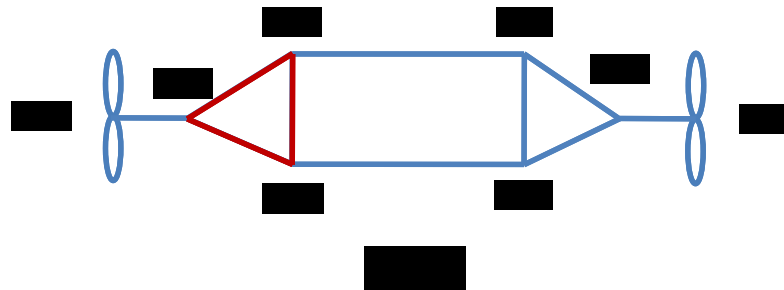
- $n$ -dimensional Shuffle-Exchange Network:  $2^n$  nodes
- Connections according to bit pattern:
  - Cyclic left- or right shifts
  - Flip (exchange) of the last bit



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## Hypercubic: Shuffle-Exchange Network

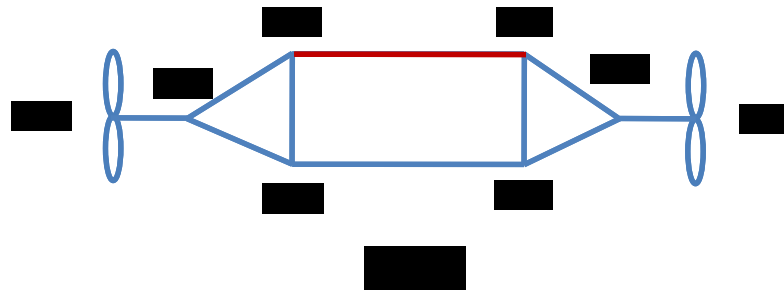
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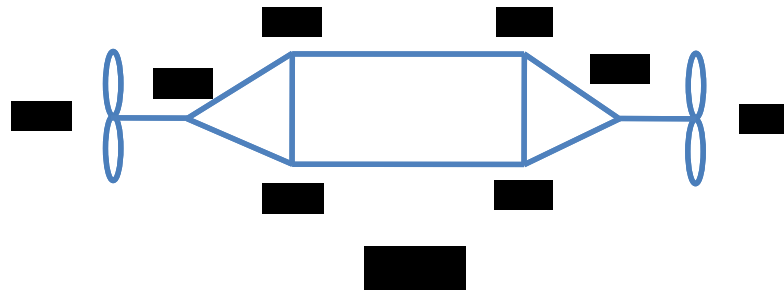


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# Mathematical Tools for Network Characterization

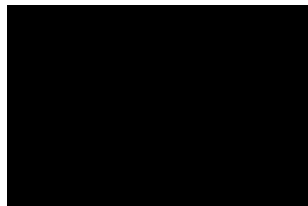
- Network = Graph (represented by its adjacency matrix)
- Spectral graph theory:

Examine relationship between the network and its Eigenvalues/Eigenvectors

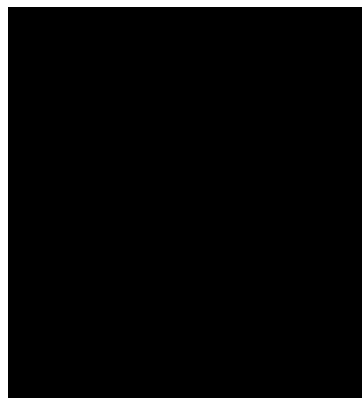
- A small example:



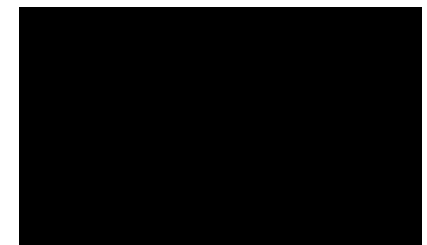
Adjacency:



Eigenvalues:



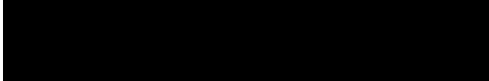
Average degree:



## Bounds from Graph Spectra

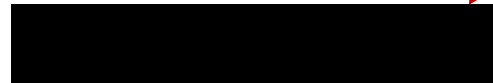
- Isoperimetric number
- Expansion property
- Routing number
- Chromatic number
- Independence number
- Bisection width

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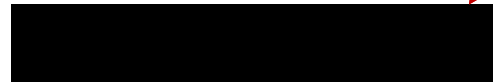


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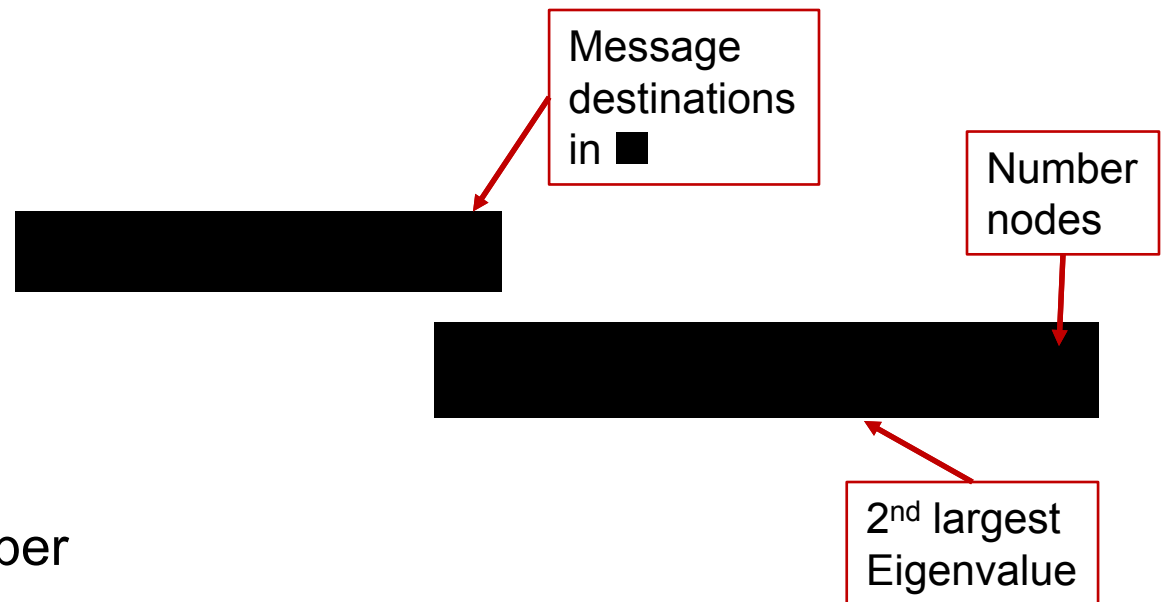


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## Spectral Relation of CCC and SE

### Decomposing CCC



- Graph transform
- Spectrum is identical

### Decomposing SE



- Similarity transform
- Spectrum is identical

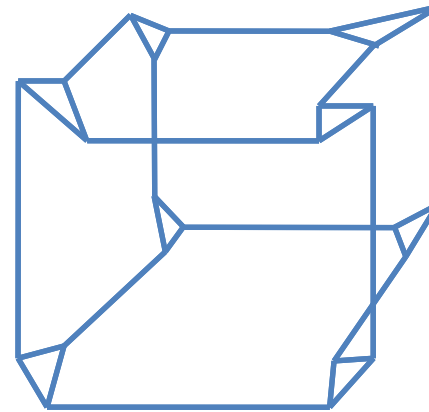
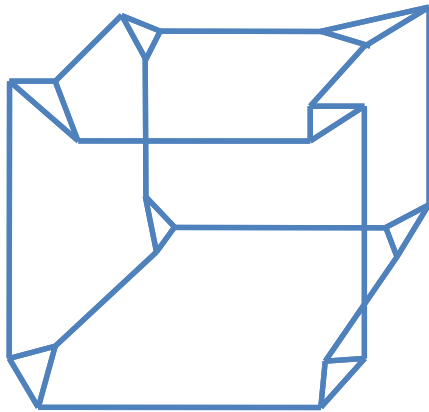
### Relation of CCC and SE

- Matching of subgraphs
- If  $d$  odd:  $\text{SpS}(\text{CCC}(d)) = \text{SpS}(\text{SE}(d))$
- If  $d$  even:  $\text{SpS}(\text{CCC}(d)) \supset \text{SpS}(\text{SE}(d))$



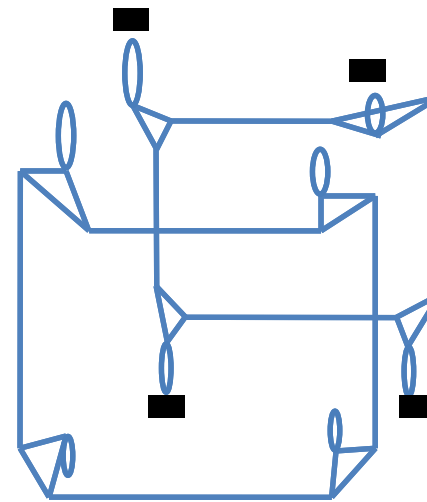
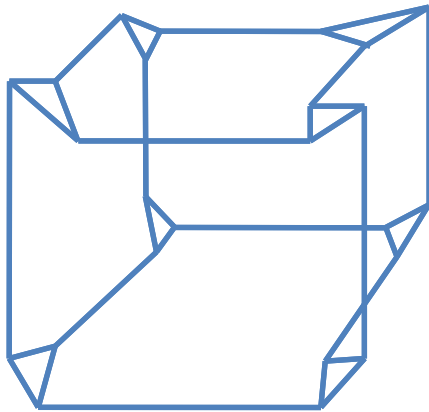
## Decomposing CCC

- Spectrum-preserving graph editing



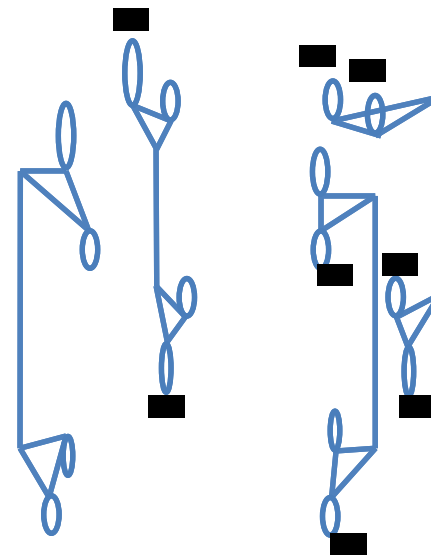
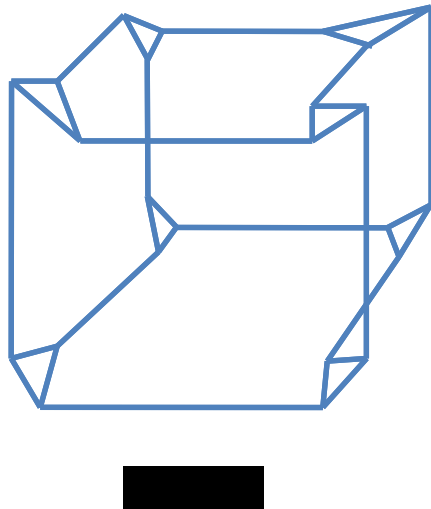
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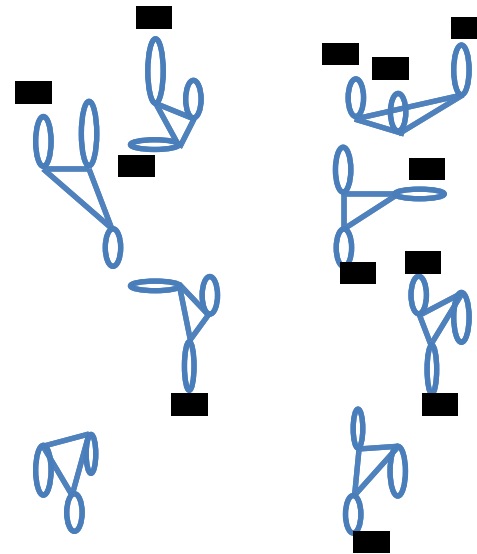
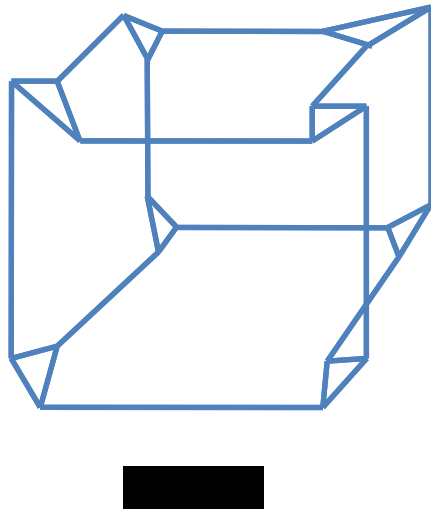
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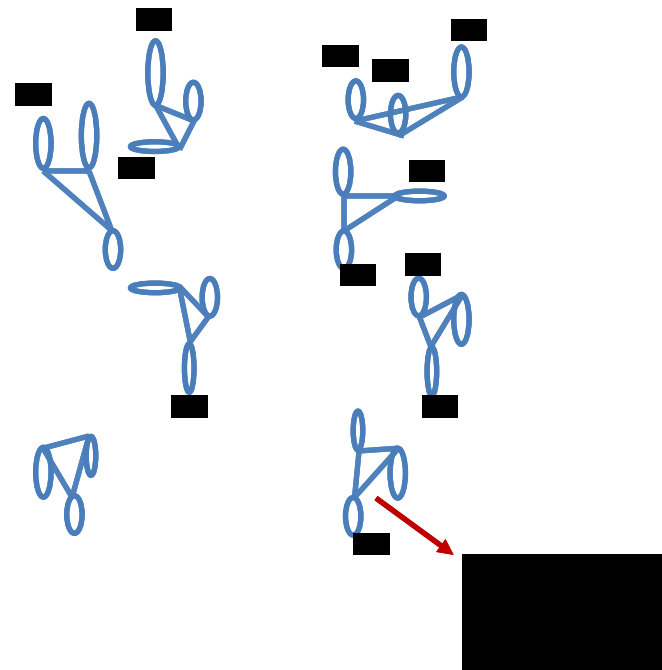
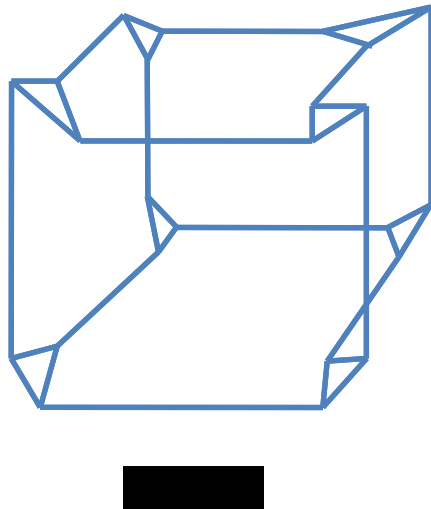
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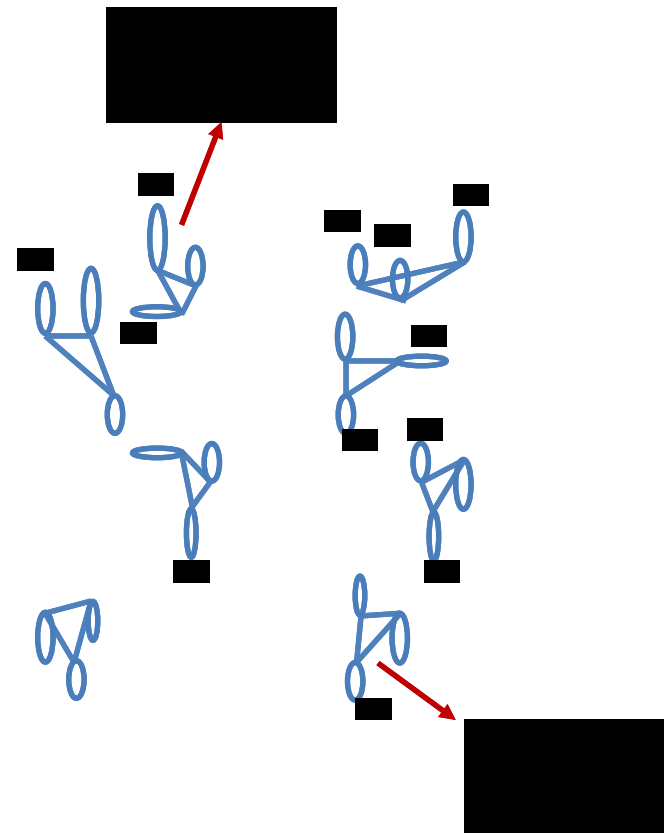
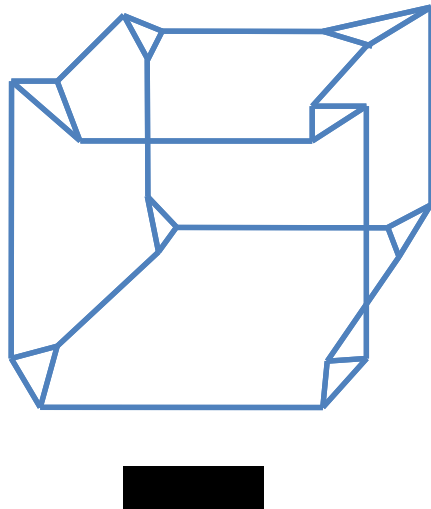
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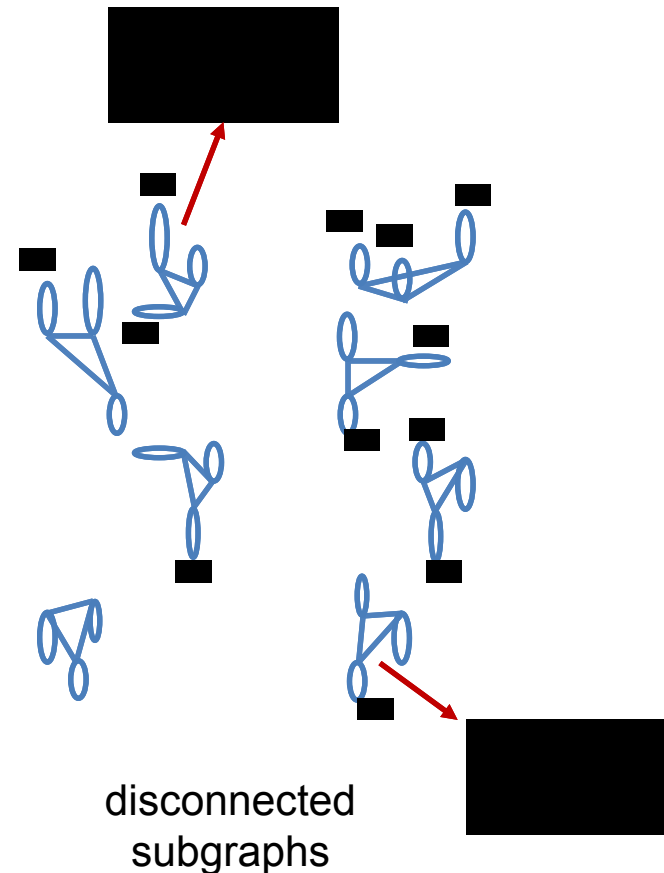
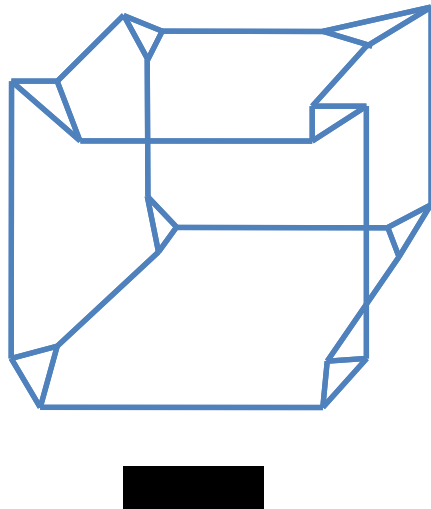
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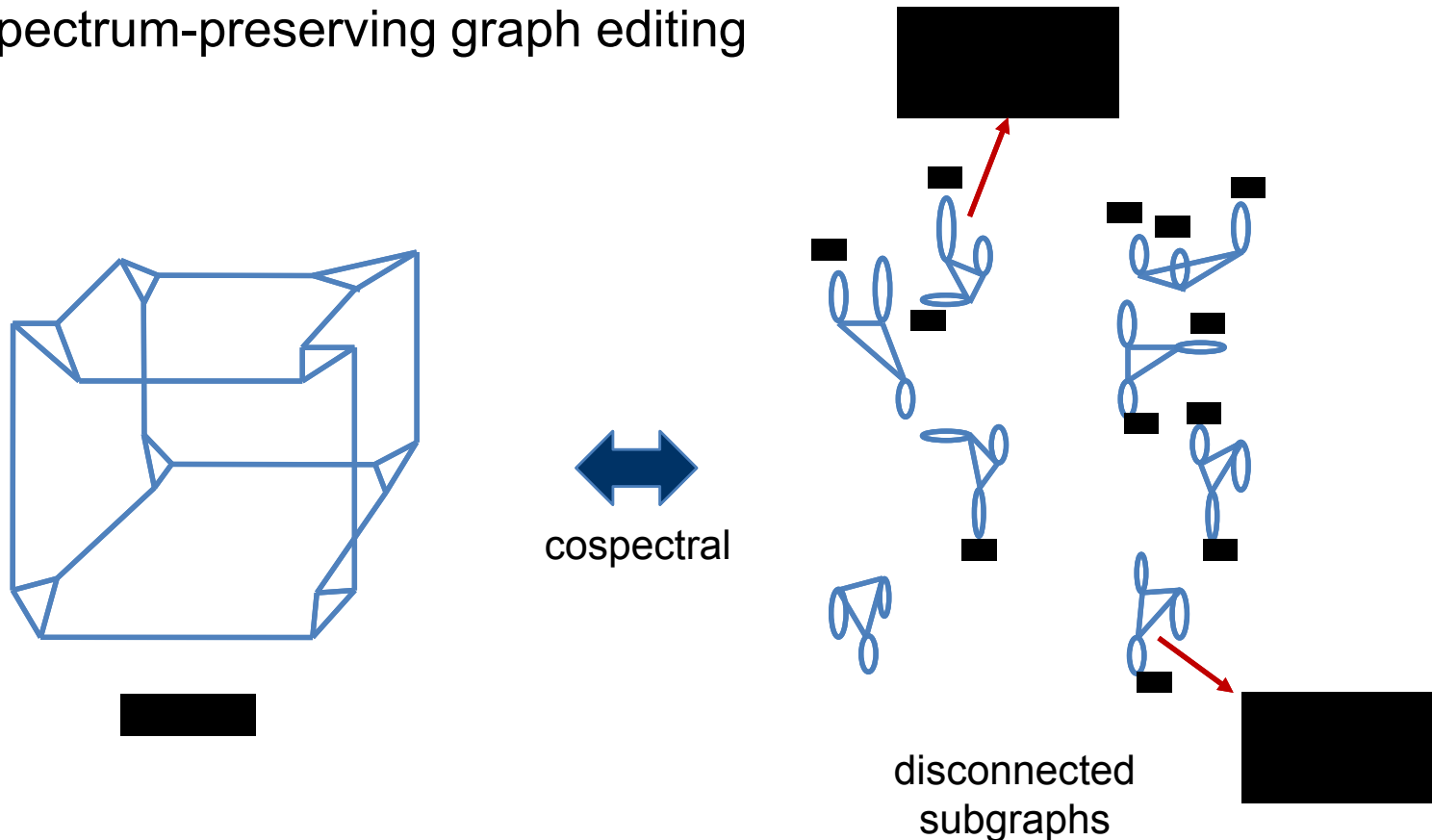
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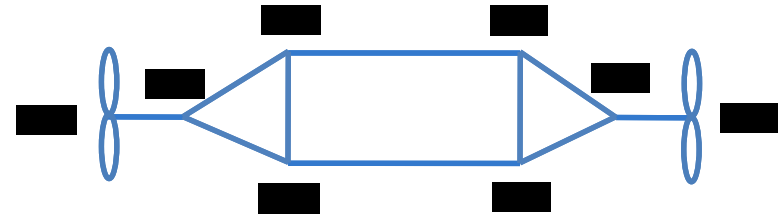
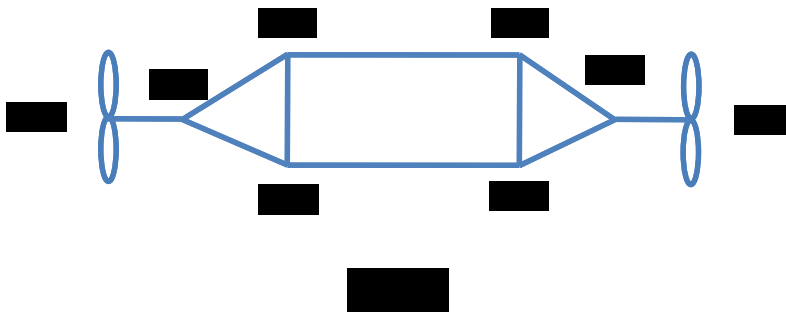
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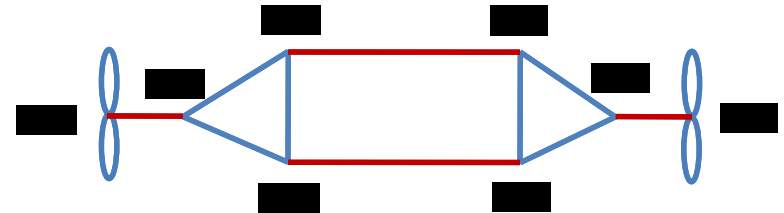
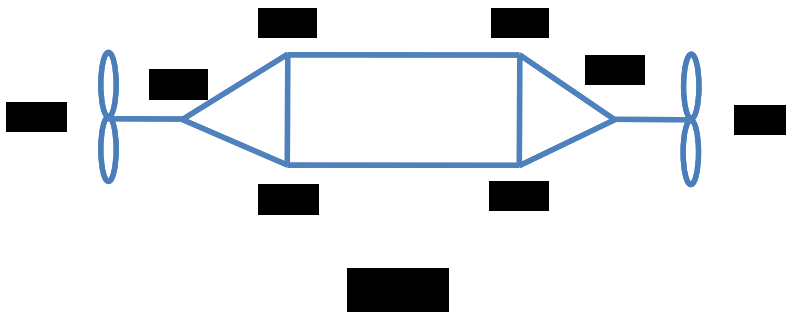
## Decomposing SE

- Removal of „exchange edges“



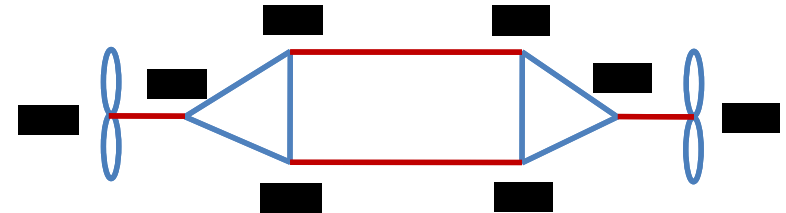
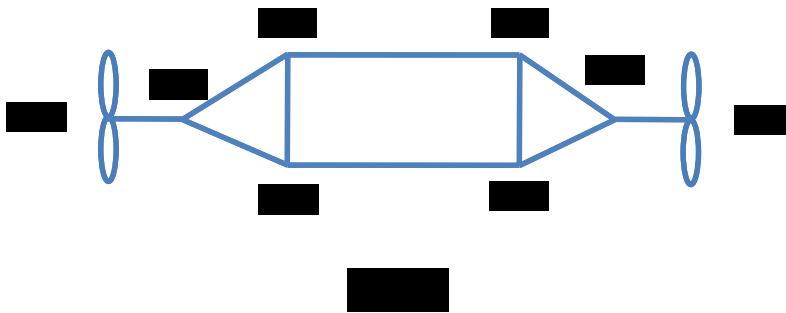
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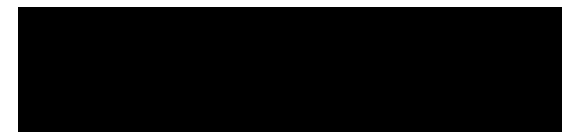


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Similarity transform with

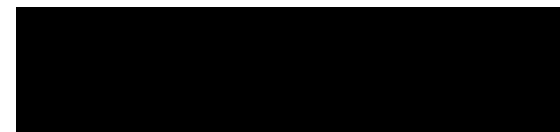


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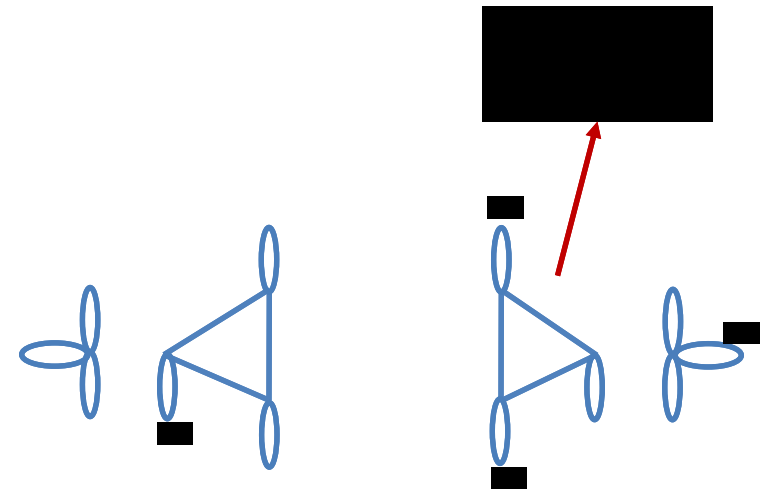
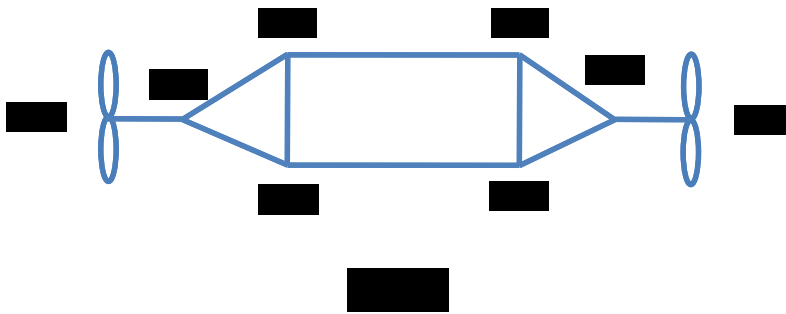


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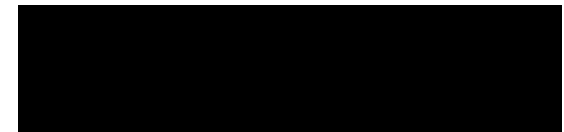


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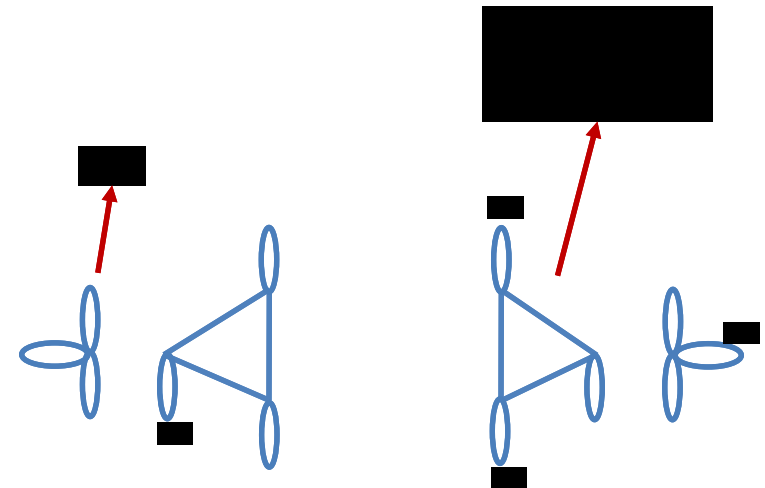
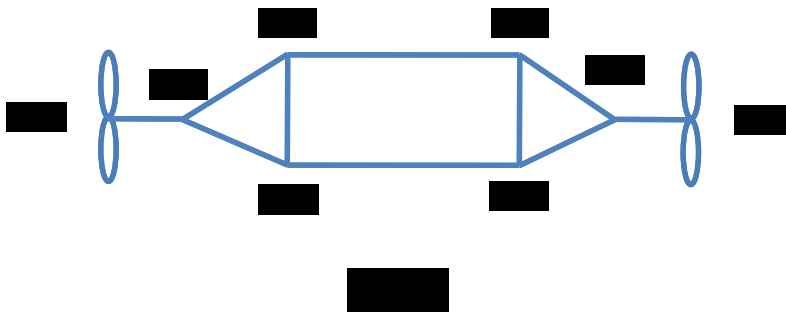


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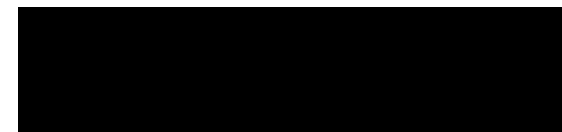


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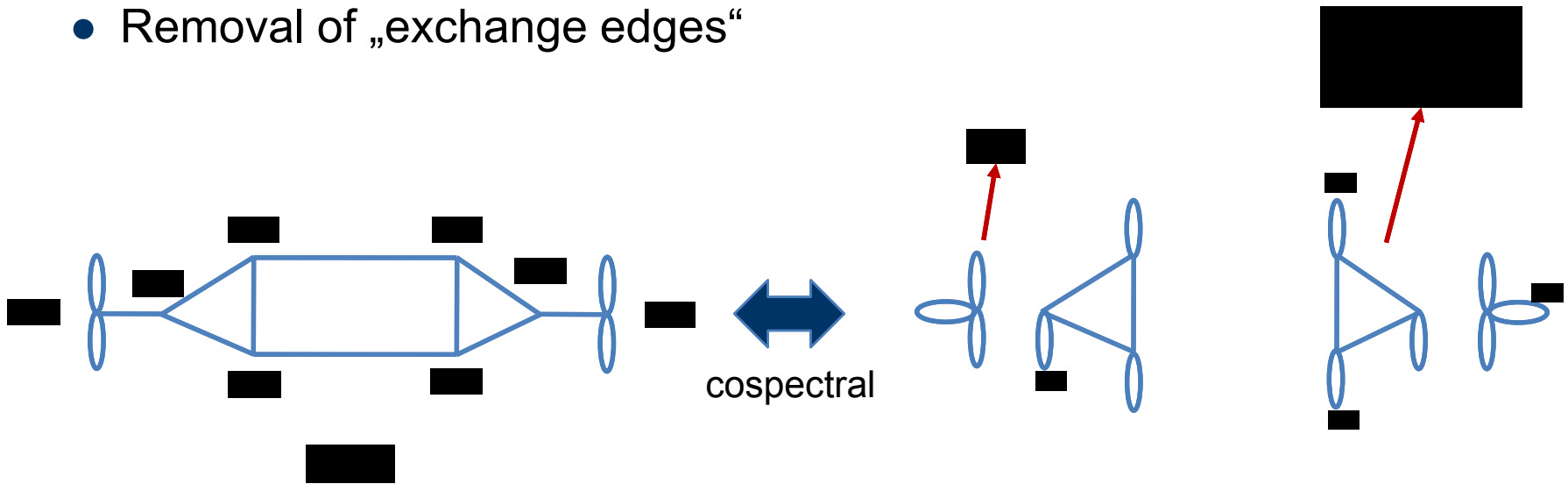


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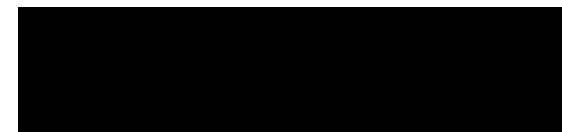


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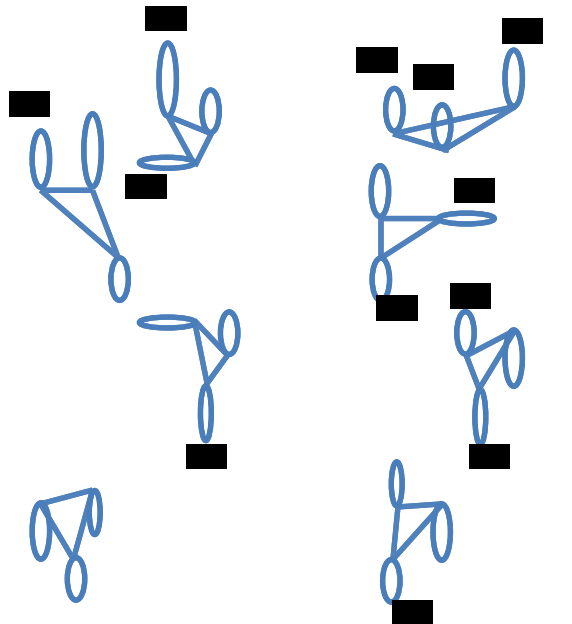
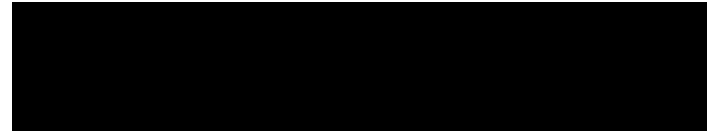


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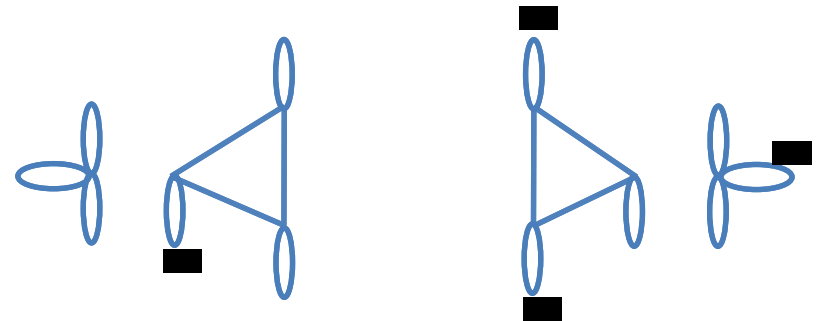


## Relating the Circles

- We characterize the mappings  
i.e.



Cospectral graph to 

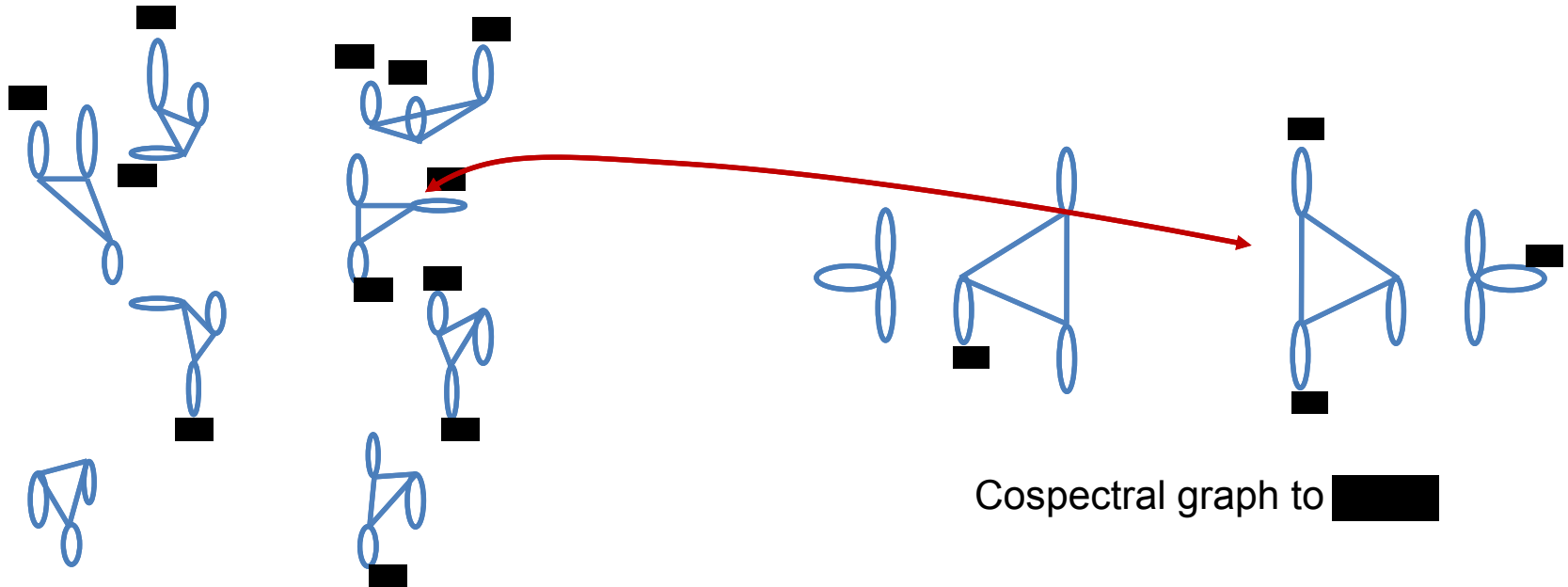
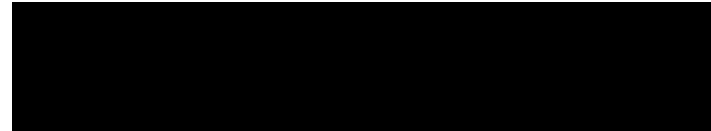


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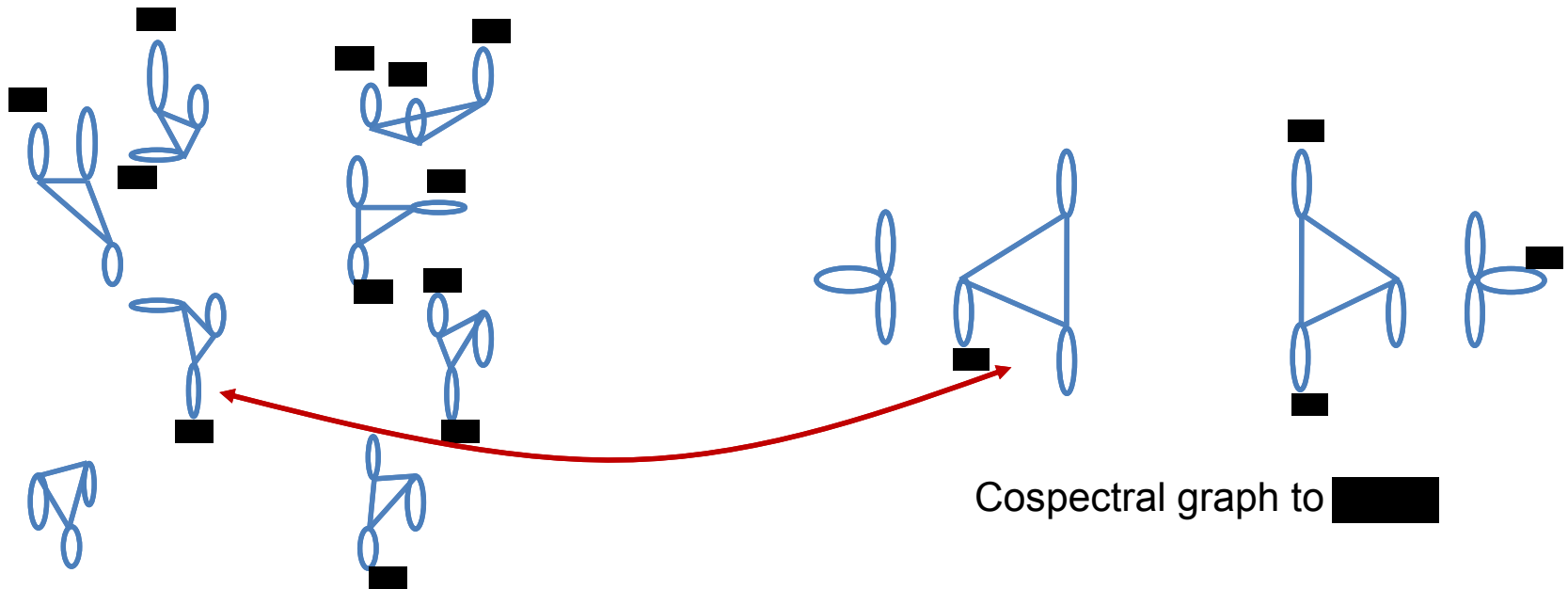
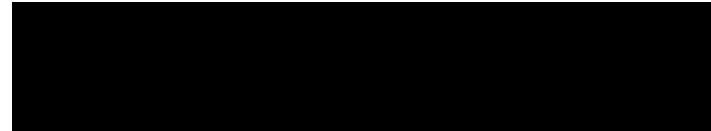


Cospectral graph to [redacted]

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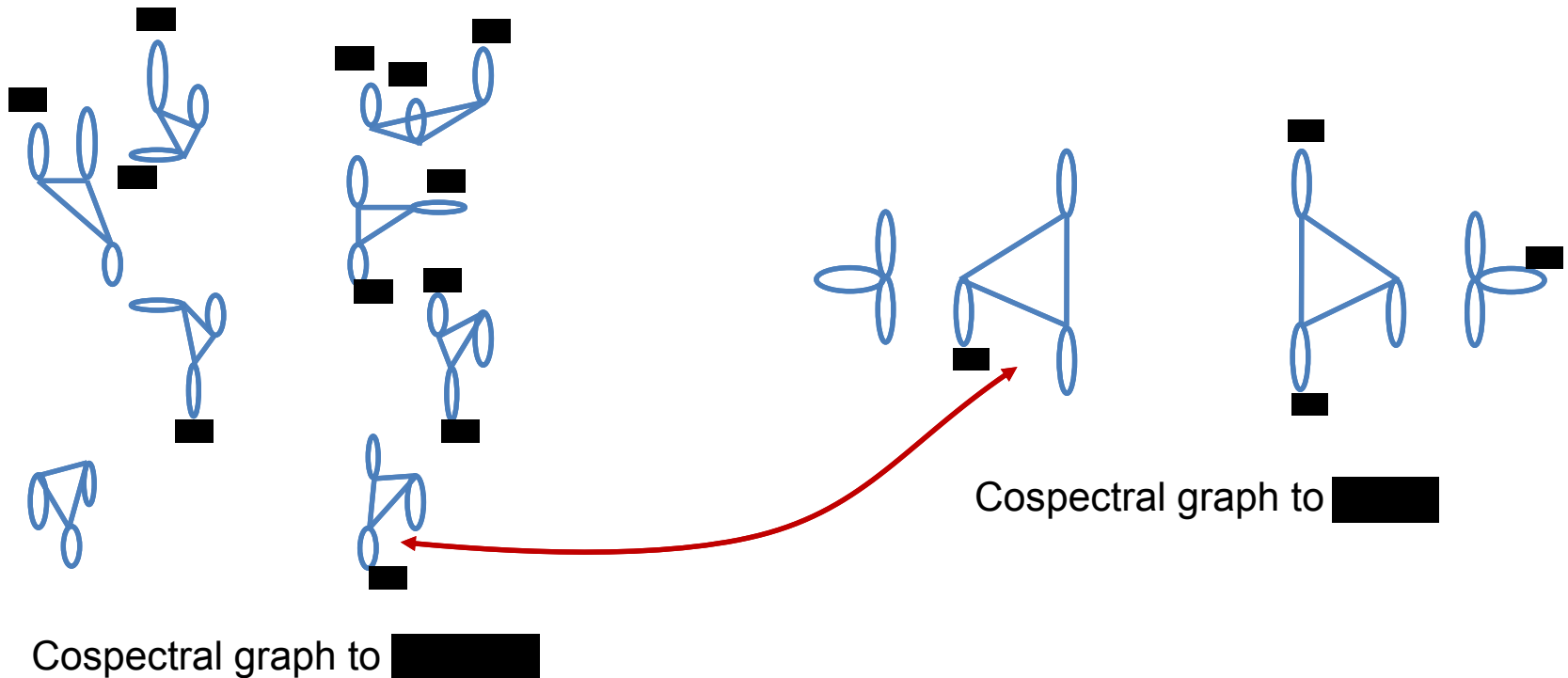
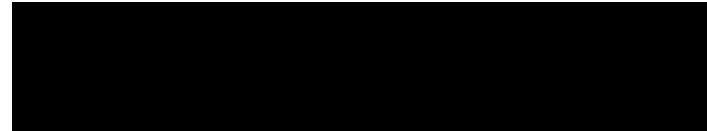


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## Result

- If  $n$  odd: [REDACTED]
- If  $n$  even: [REDACTED]
  - [REDACTED]  
[REDACTED]
  - Note that  $n$  is not the only difference.

We illustrate that in a second.

## Result

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There is always a complete eigenvalue mapping 😊

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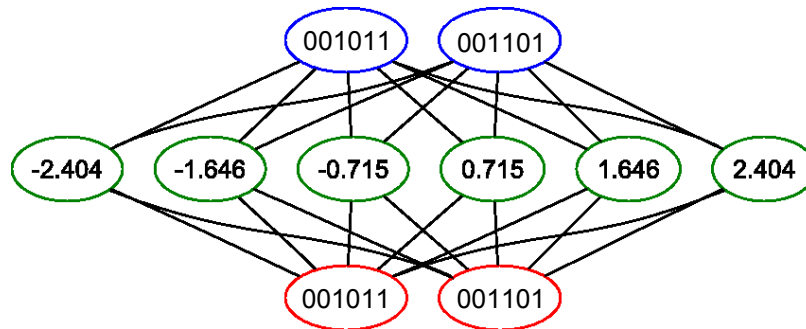
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## Relating the Circles

- Let  $\mathbf{A}$
- Denote e.g.  $\begin{bmatrix} -1 & & \\ & -1 & \\ & & 1 \end{bmatrix}$  as  $\mathbf{110}$  (by the diagonal entries)
- Aperiodic case: a cycle appears in both graphs

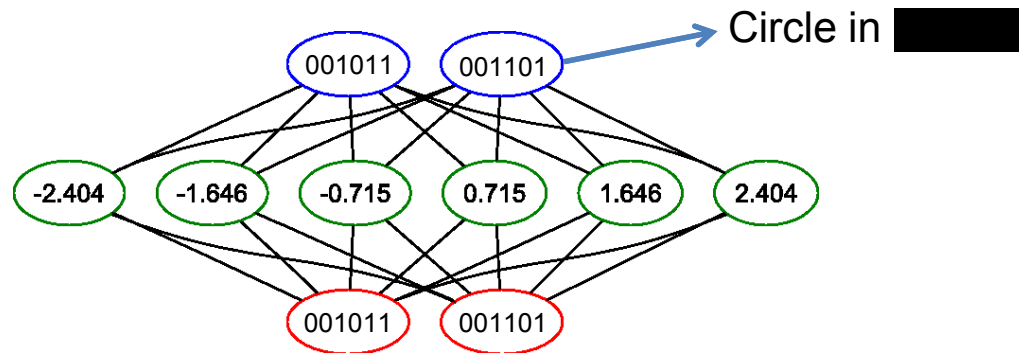


- Sometimes, this does not work!  
 $\mathbf{A}$  compresses periodic cycles

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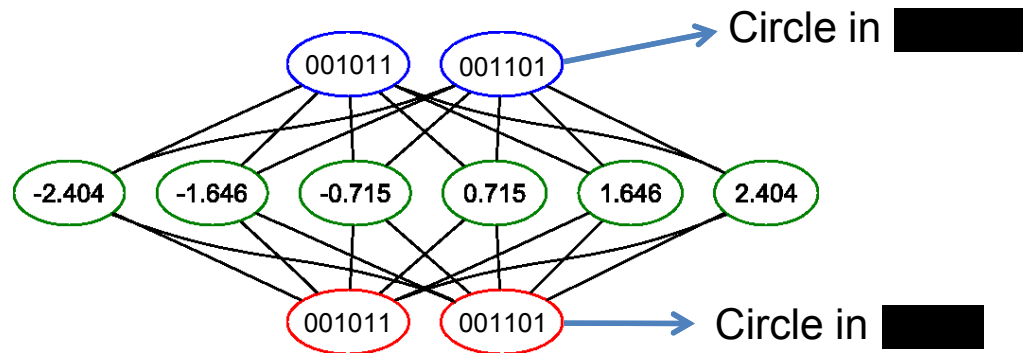
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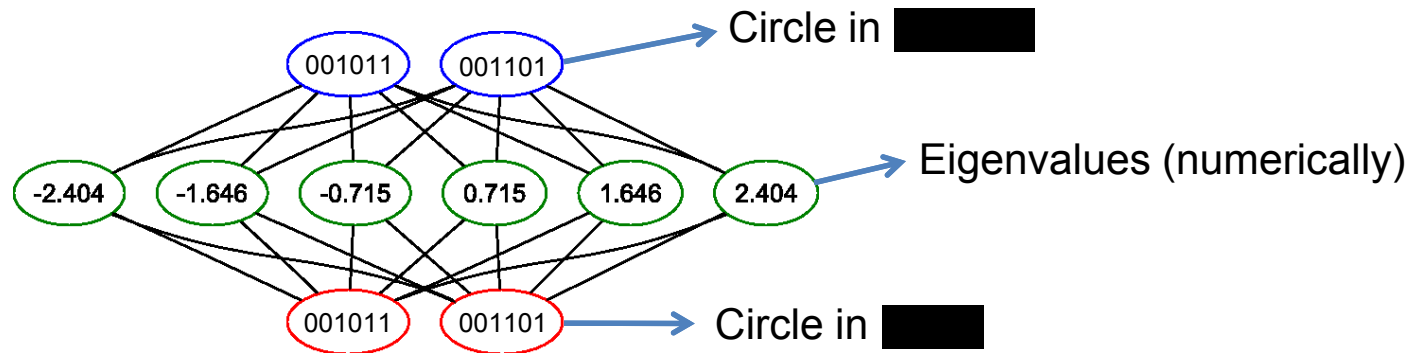


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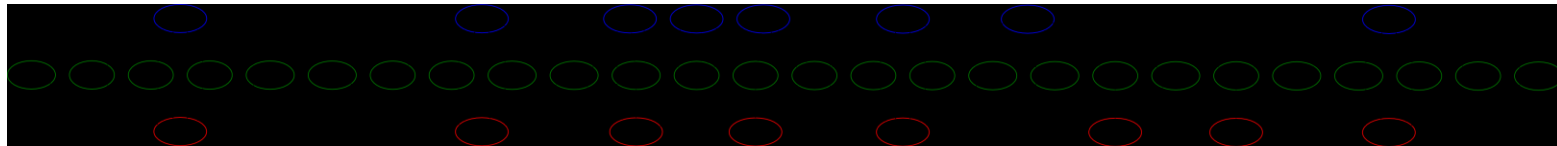
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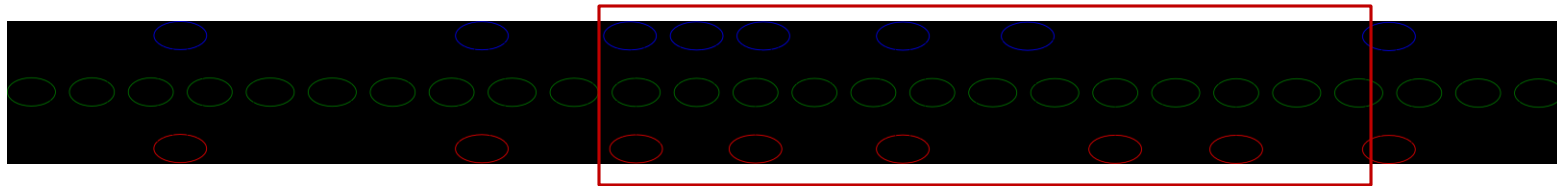
# Non-trivial correspondences between EVs and Circles



011011    111111    000000                    000001                    001001

011                    0                    000001                    001                    111111

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011011    111111    000000

000001

001001

011

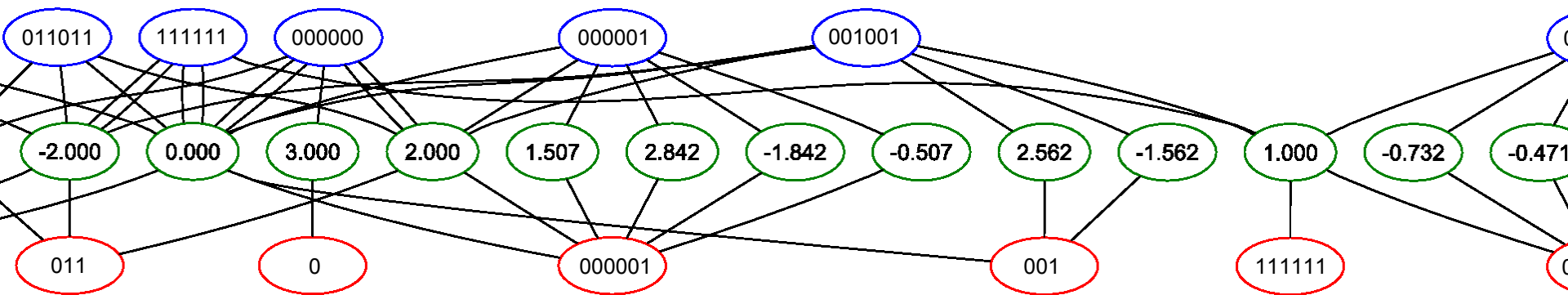
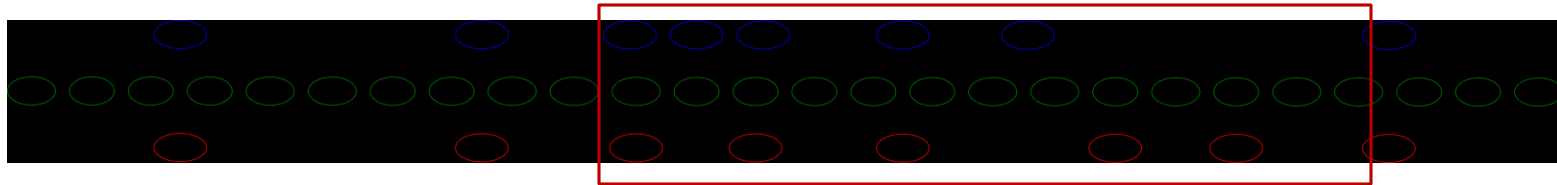
0

000001

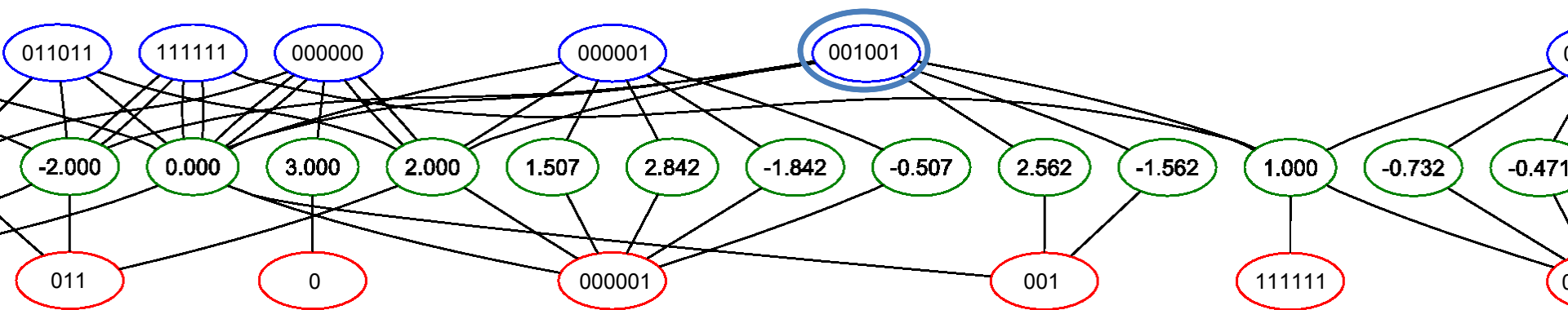
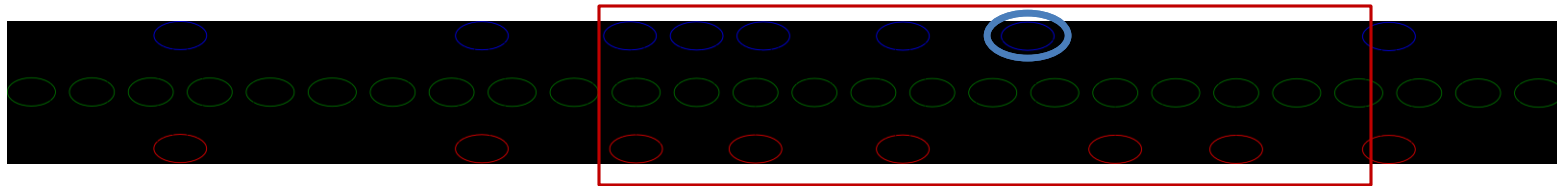
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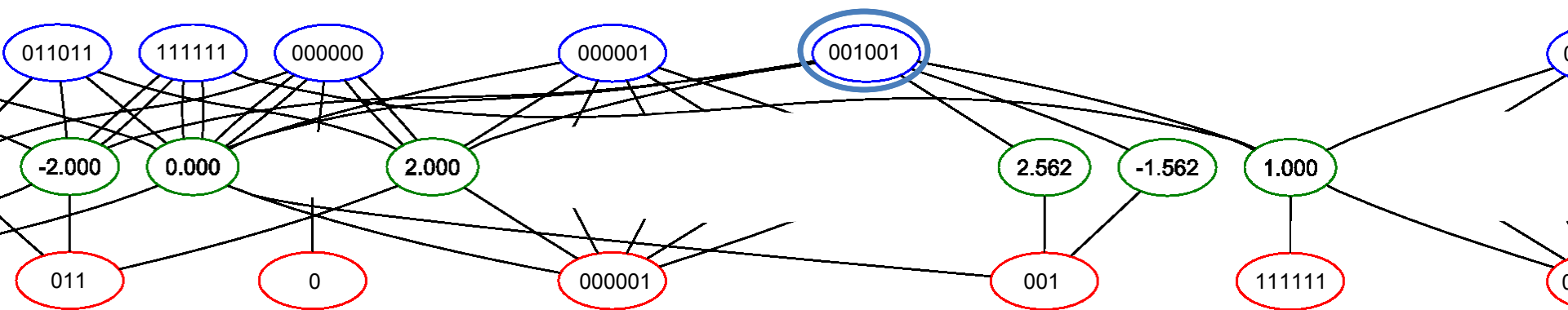
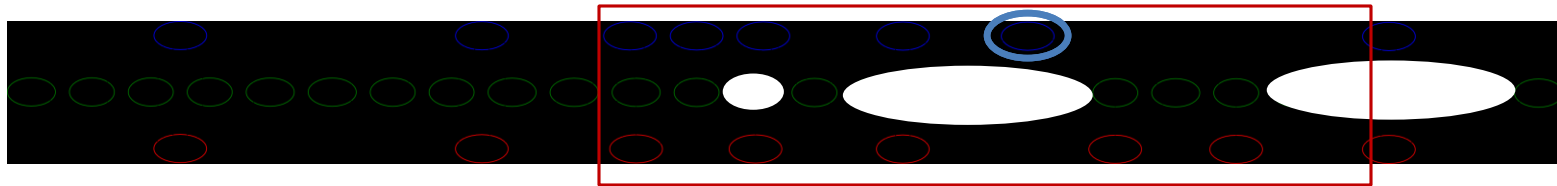
# Non-trivial correspondences between EVs and Circles



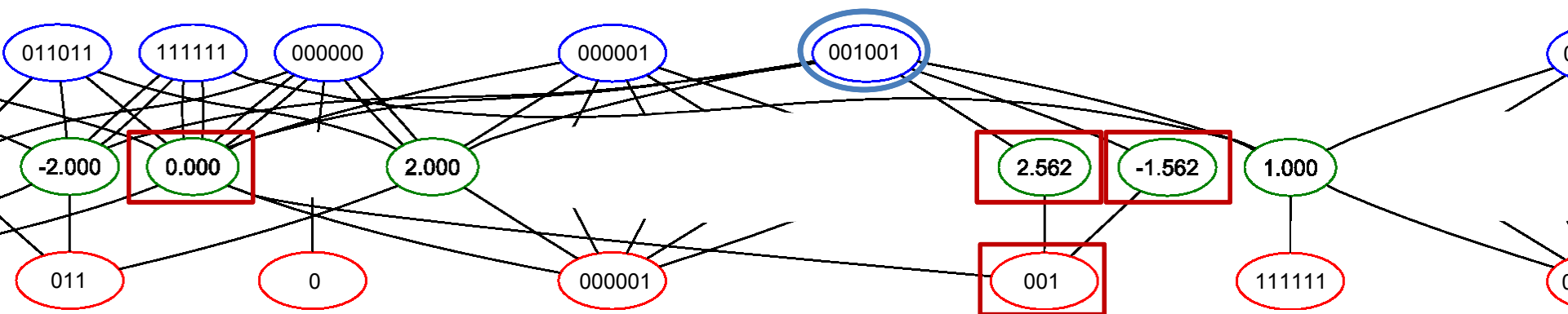
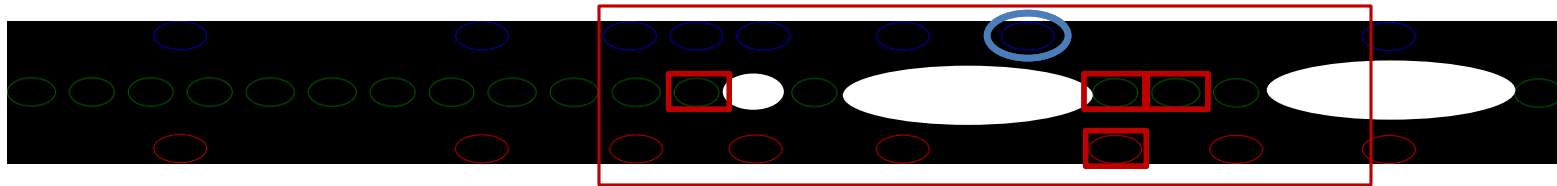
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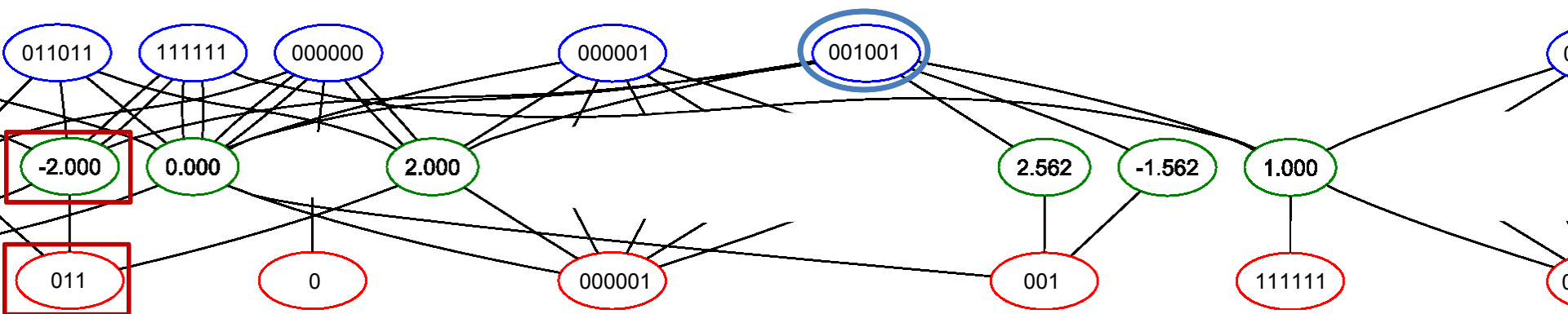
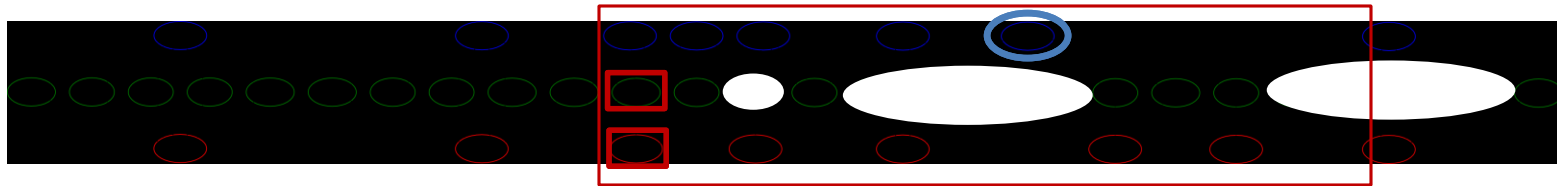


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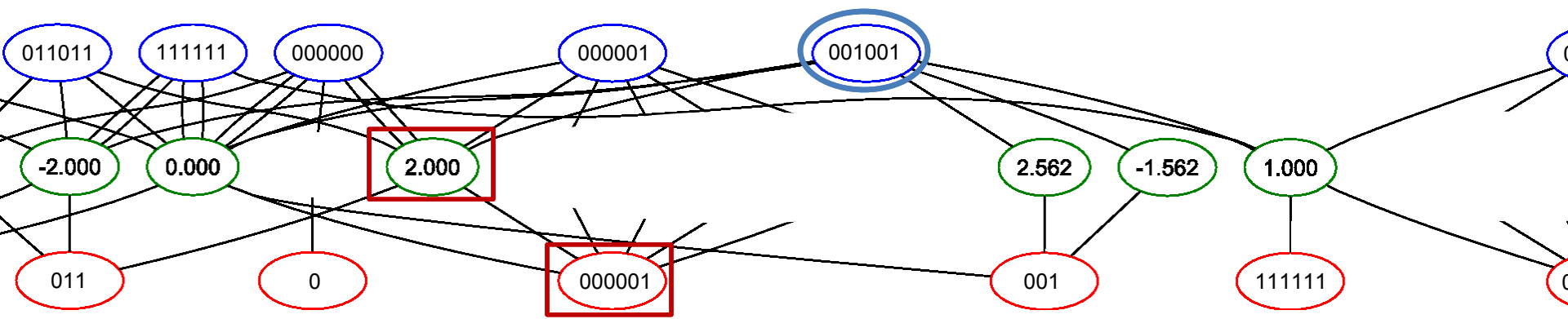
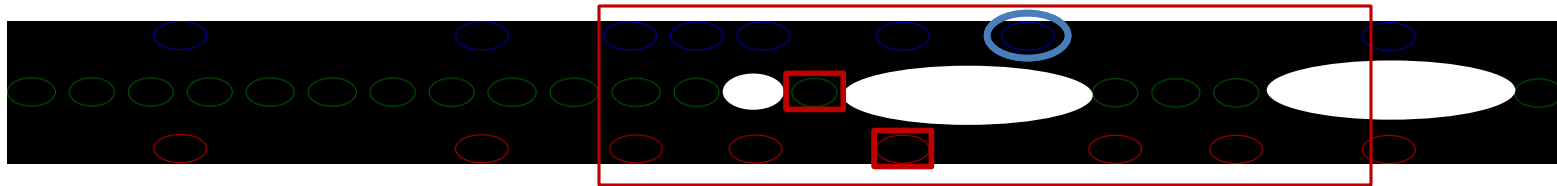




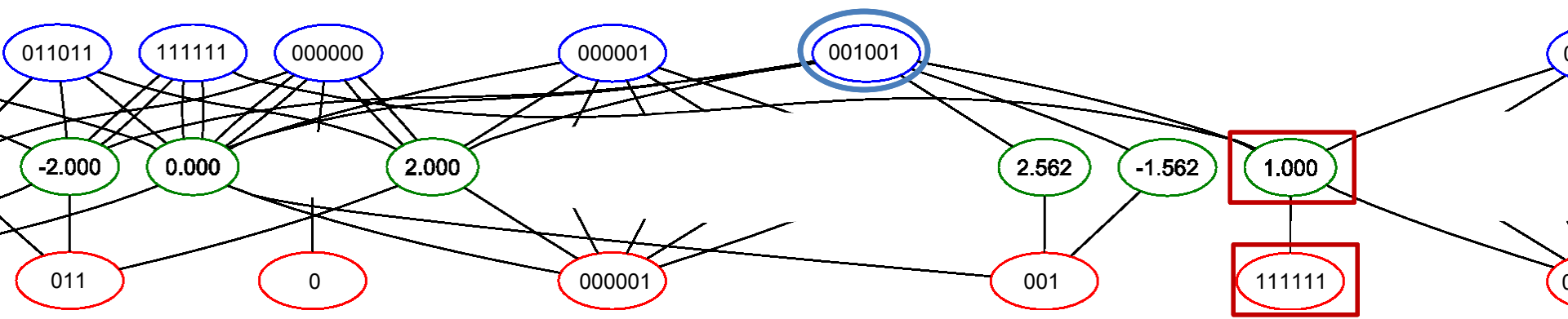
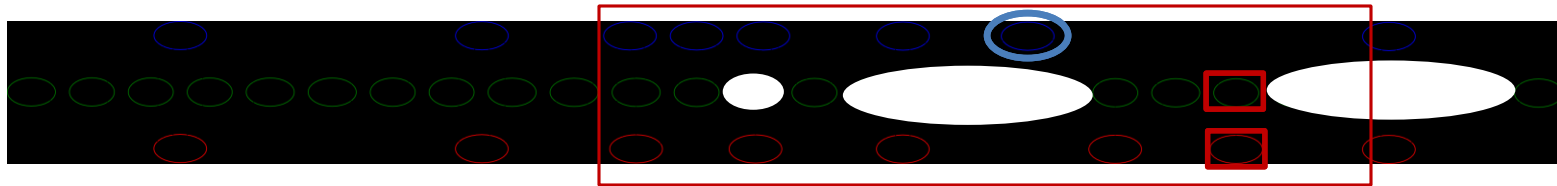
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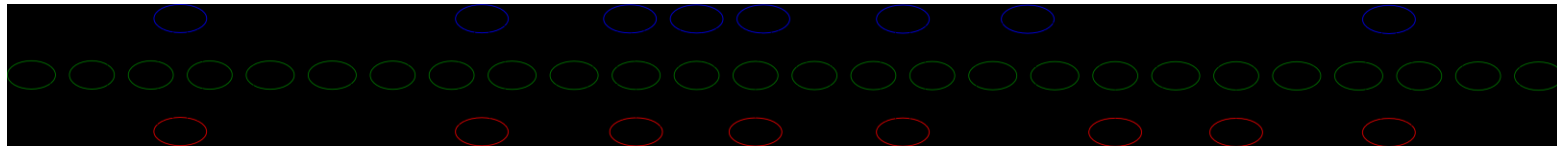
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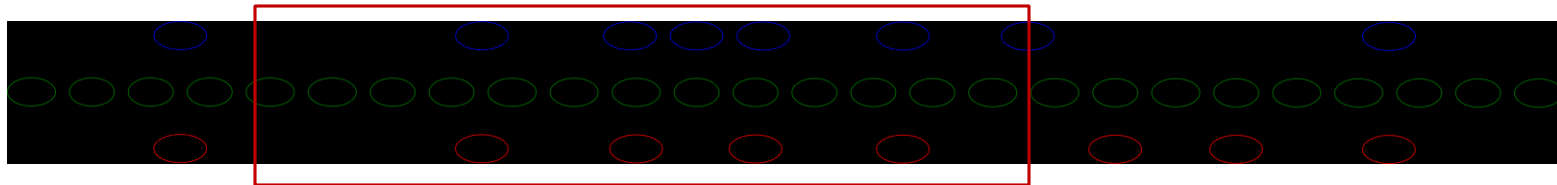
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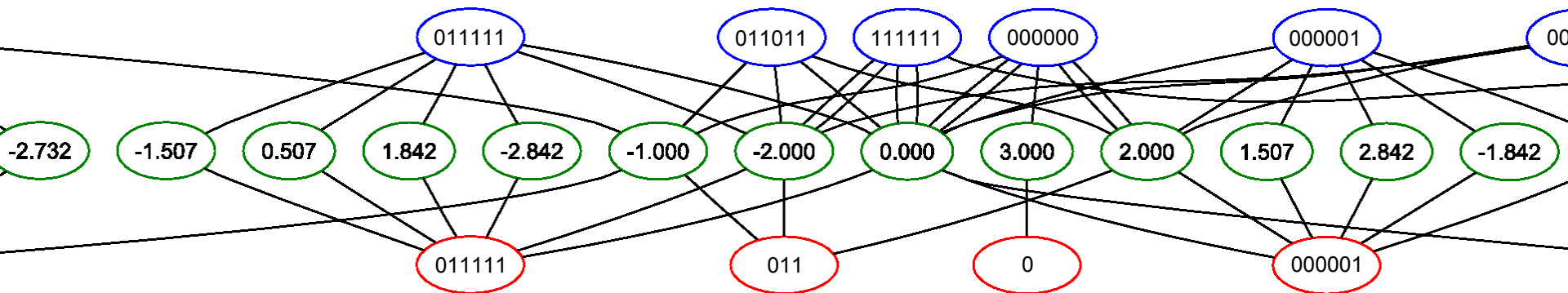
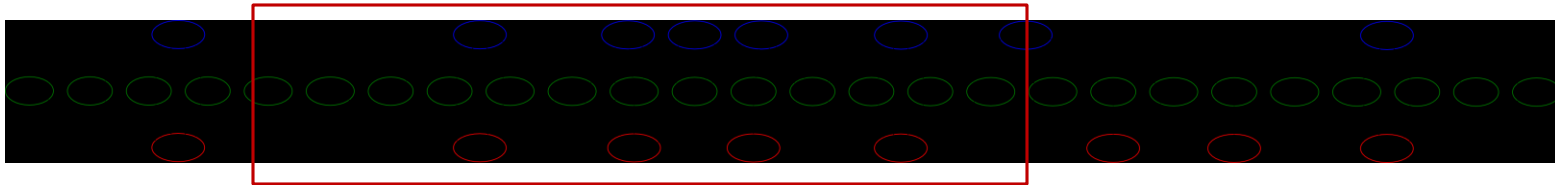
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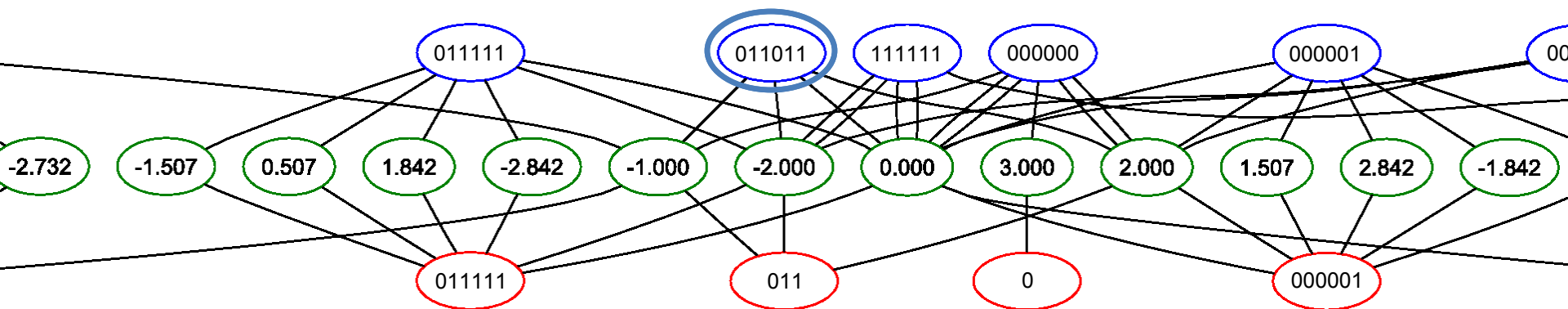
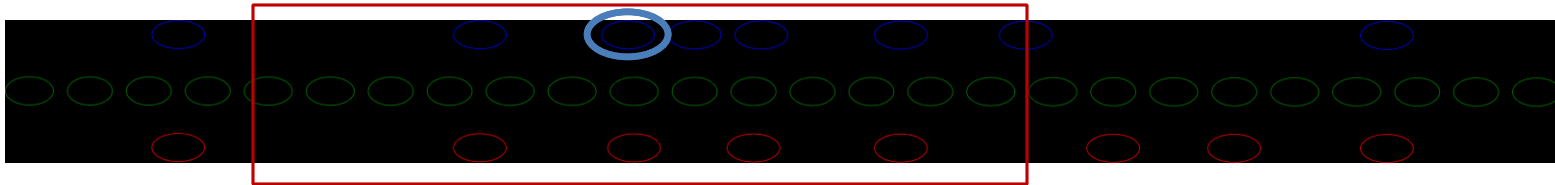
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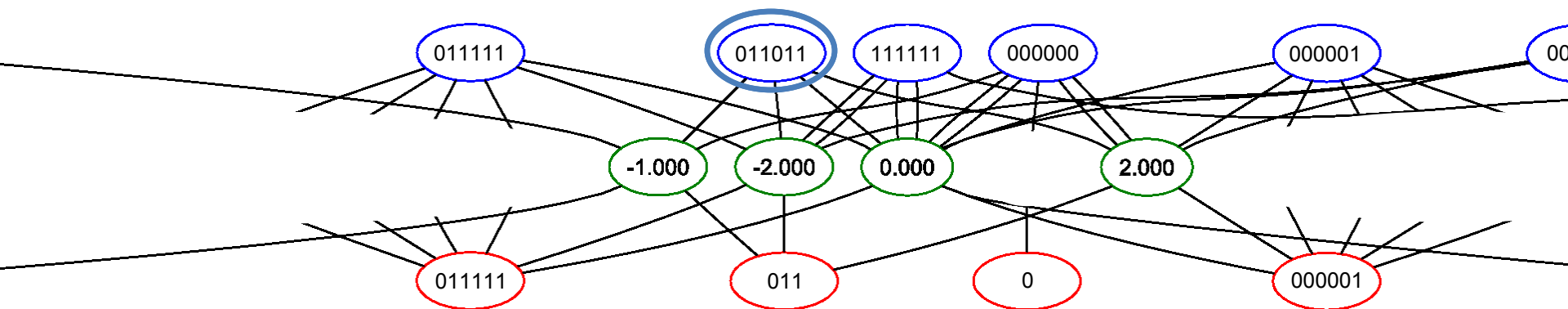
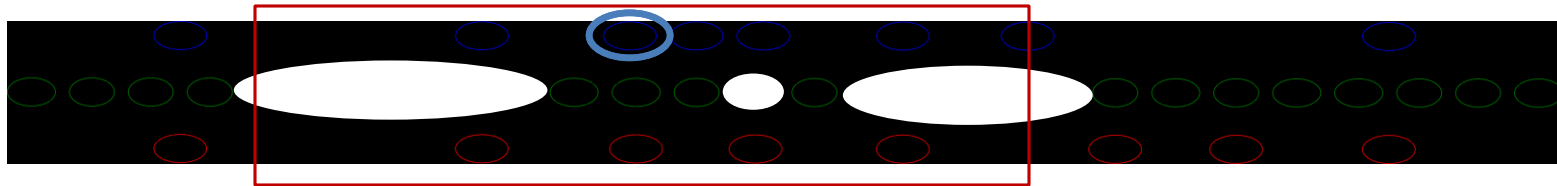
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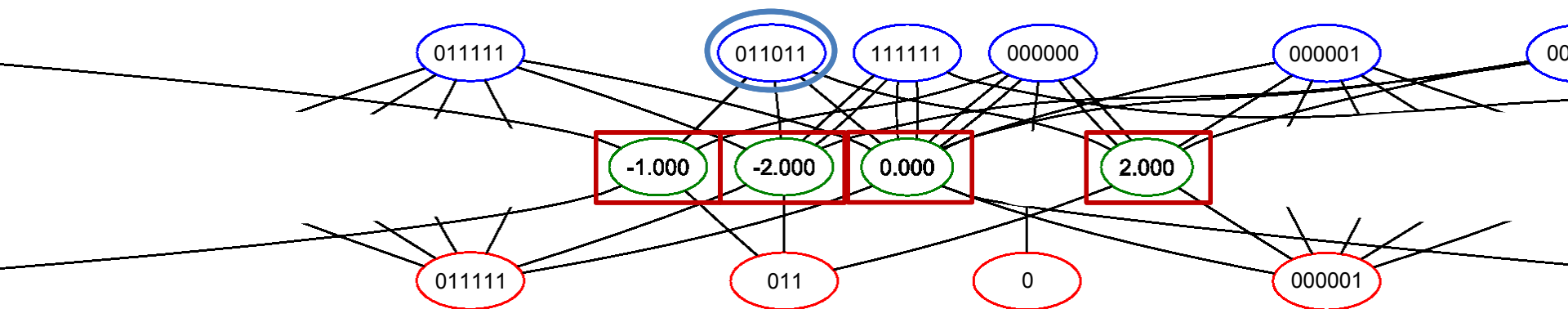
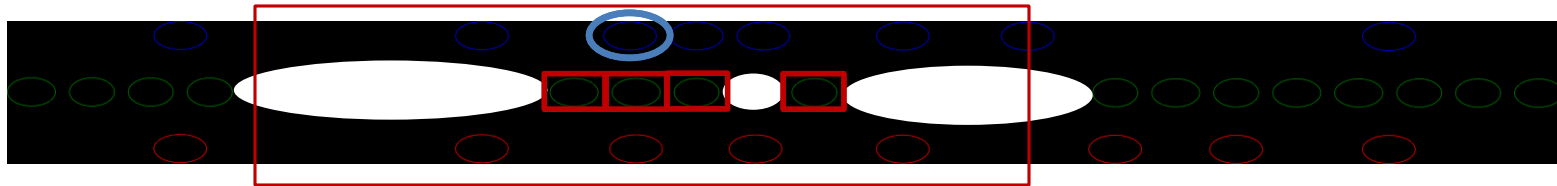


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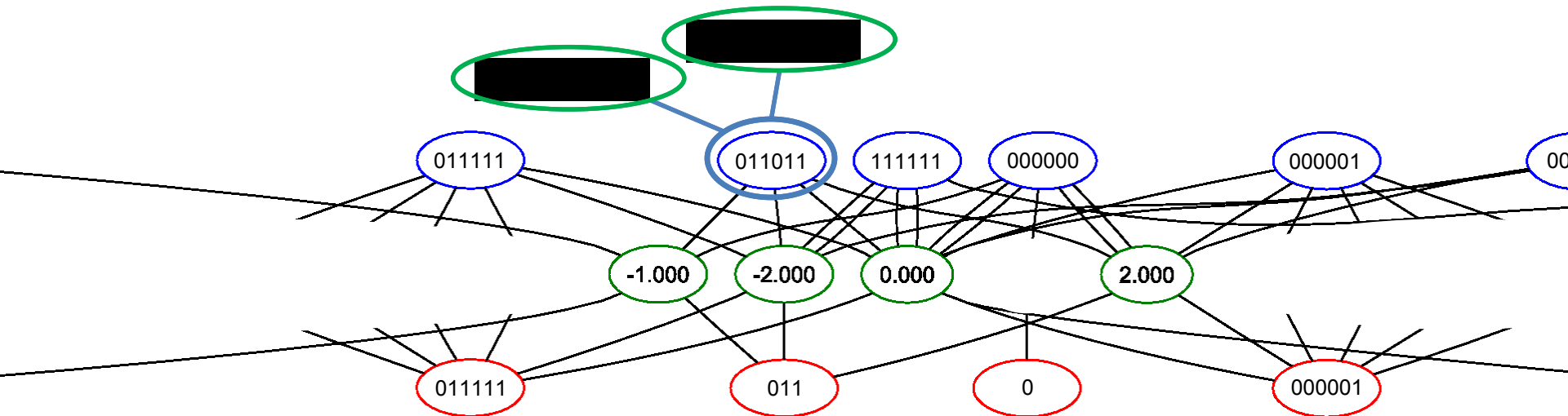
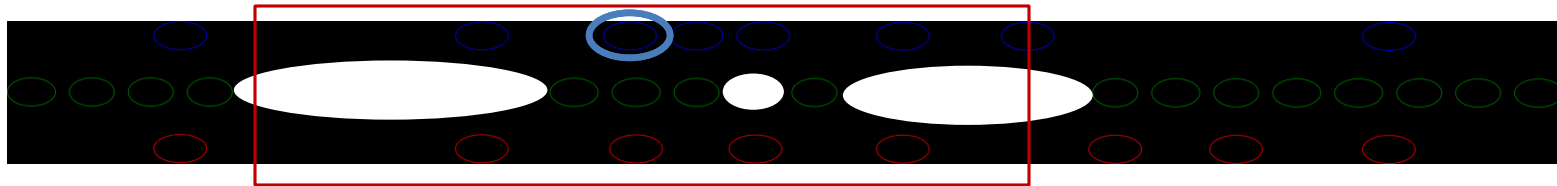




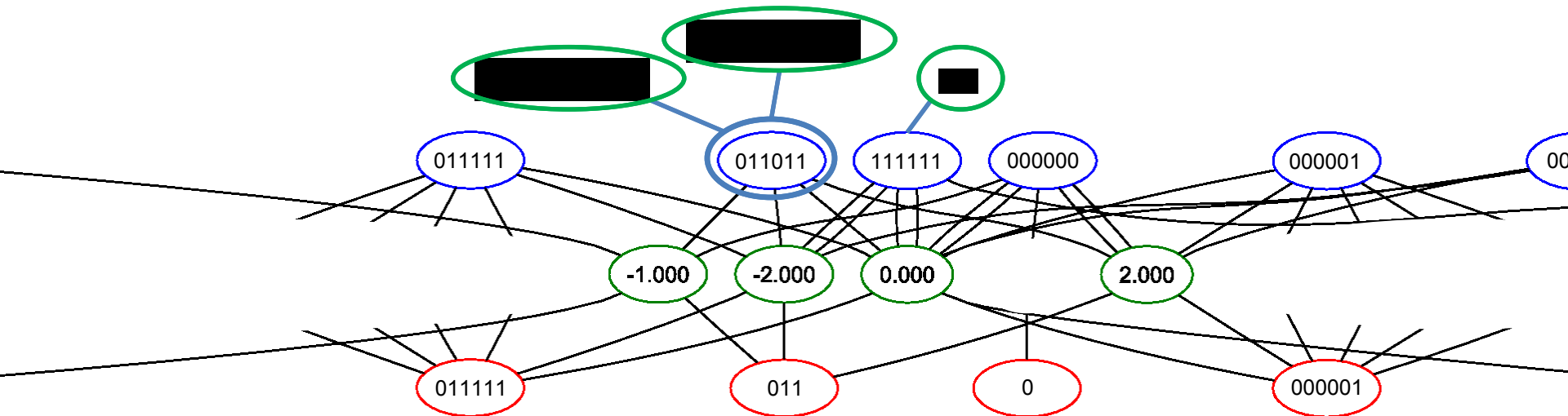
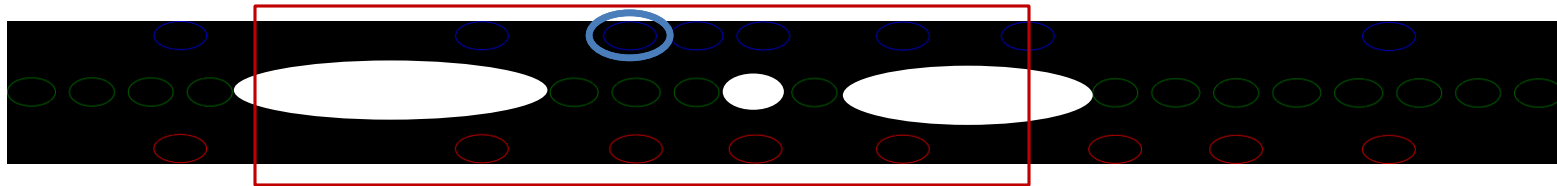
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████████████████████
  - Spectral sets **differ** when  $d$  **even**  
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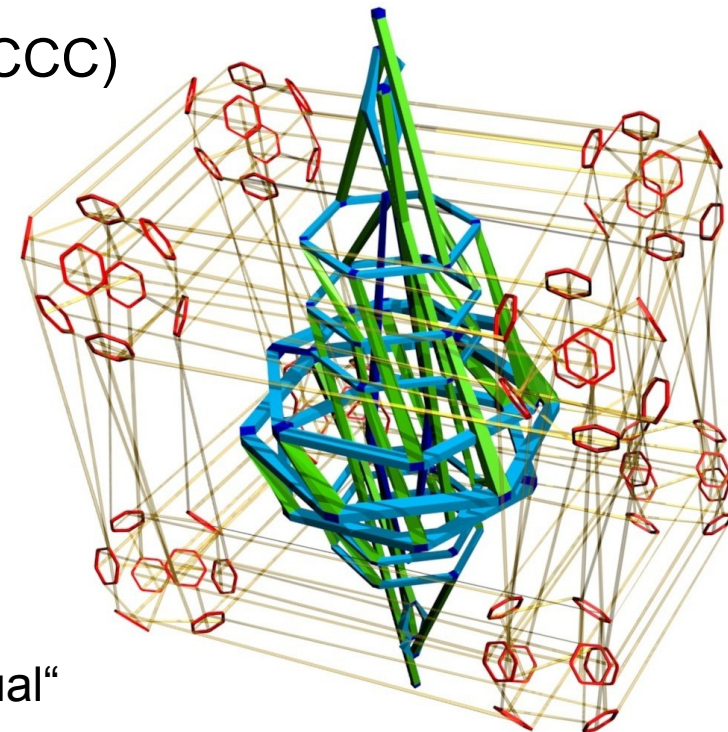
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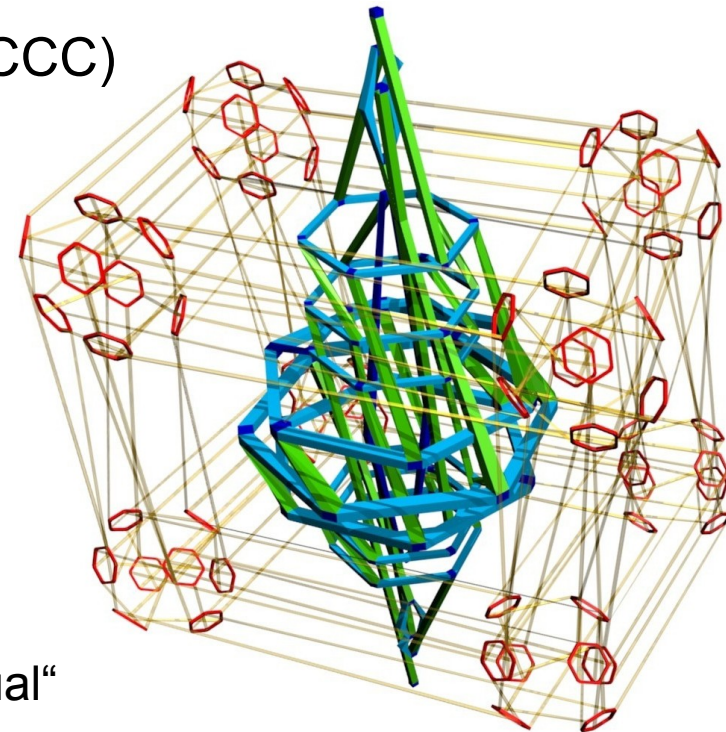


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- Future work: pin down the “accidental”



**Thank you for your attention.**