

Rich Modelling

Capturing the context
of data as it exists in
the wild



FAB LABS

DATA SCIENCE
IN ENERGY

Production Tuner

for increased oil recovery

DATA SCIENCE
IN ENERGY

State of the art?

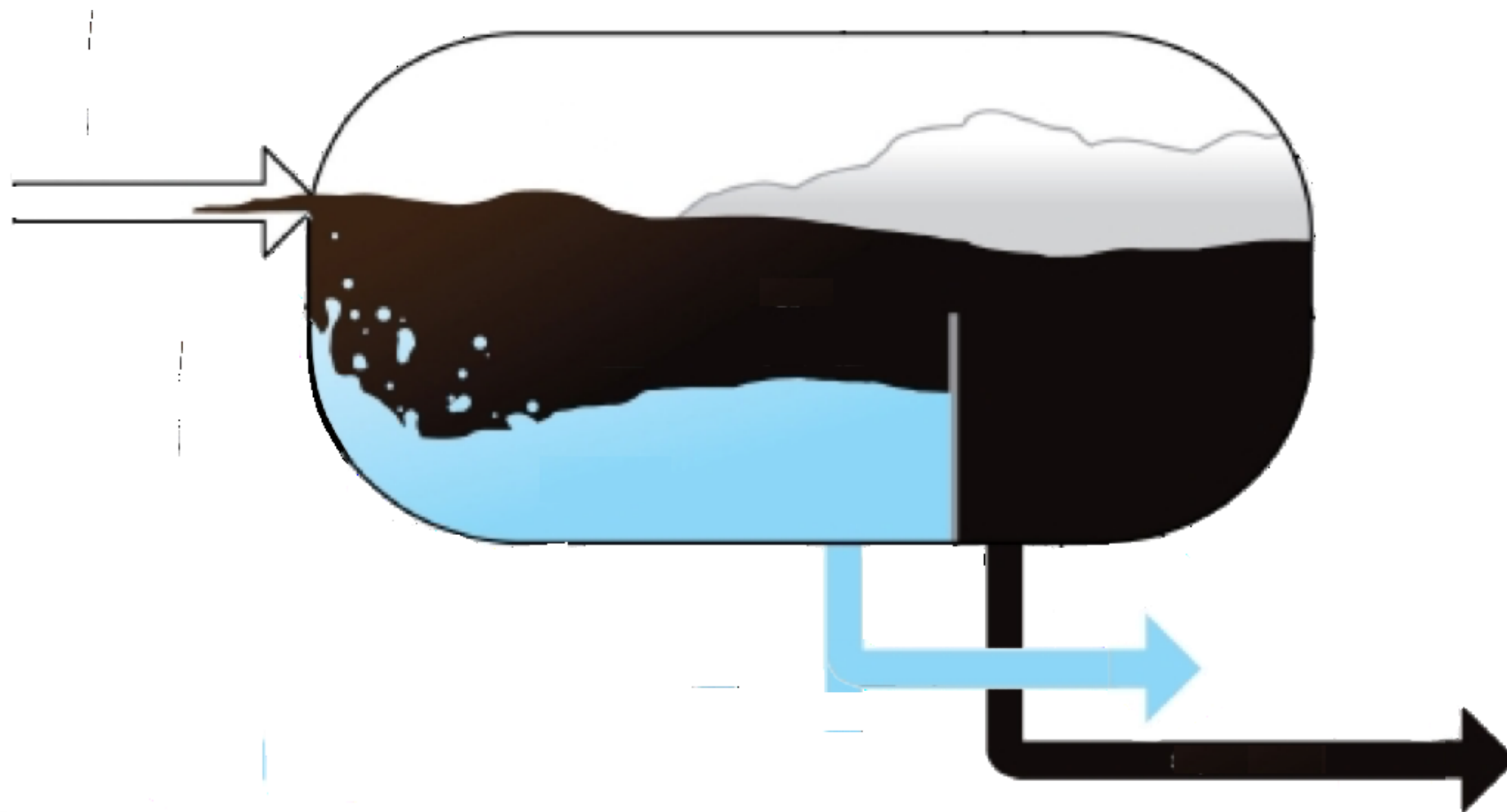


Model => Analysis => Decision
Short windows of opportunity



Topside Constraints

Multiple wells & separators all with different characteristics = tradeoffs.



Production Tuner

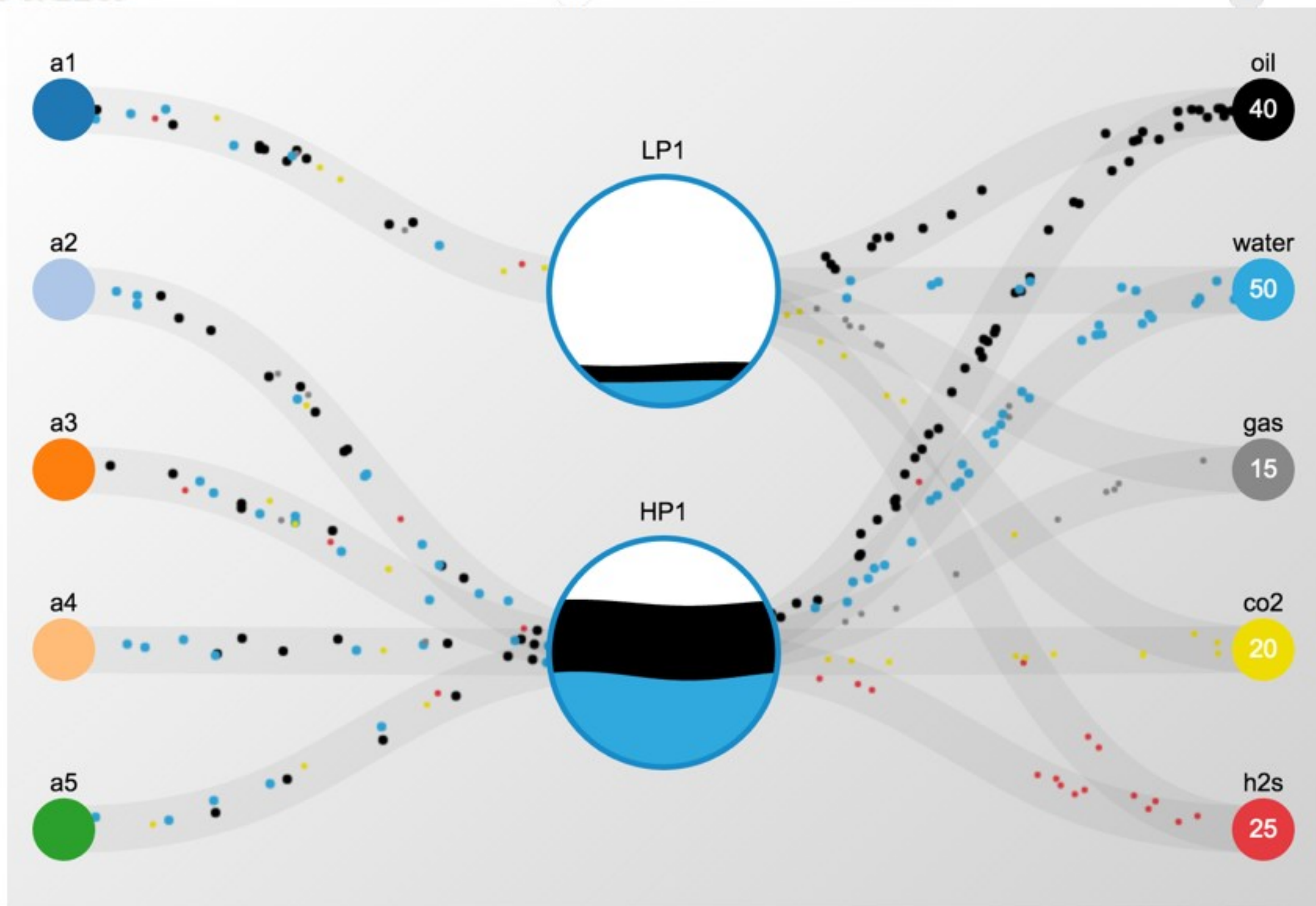
- near-realtime monitoring
- runs millions of scenarios per day
- not more alarms, best opportunities
- cuts execution lead-time to zero*
- sees hidden edge cases



wells

separators

export

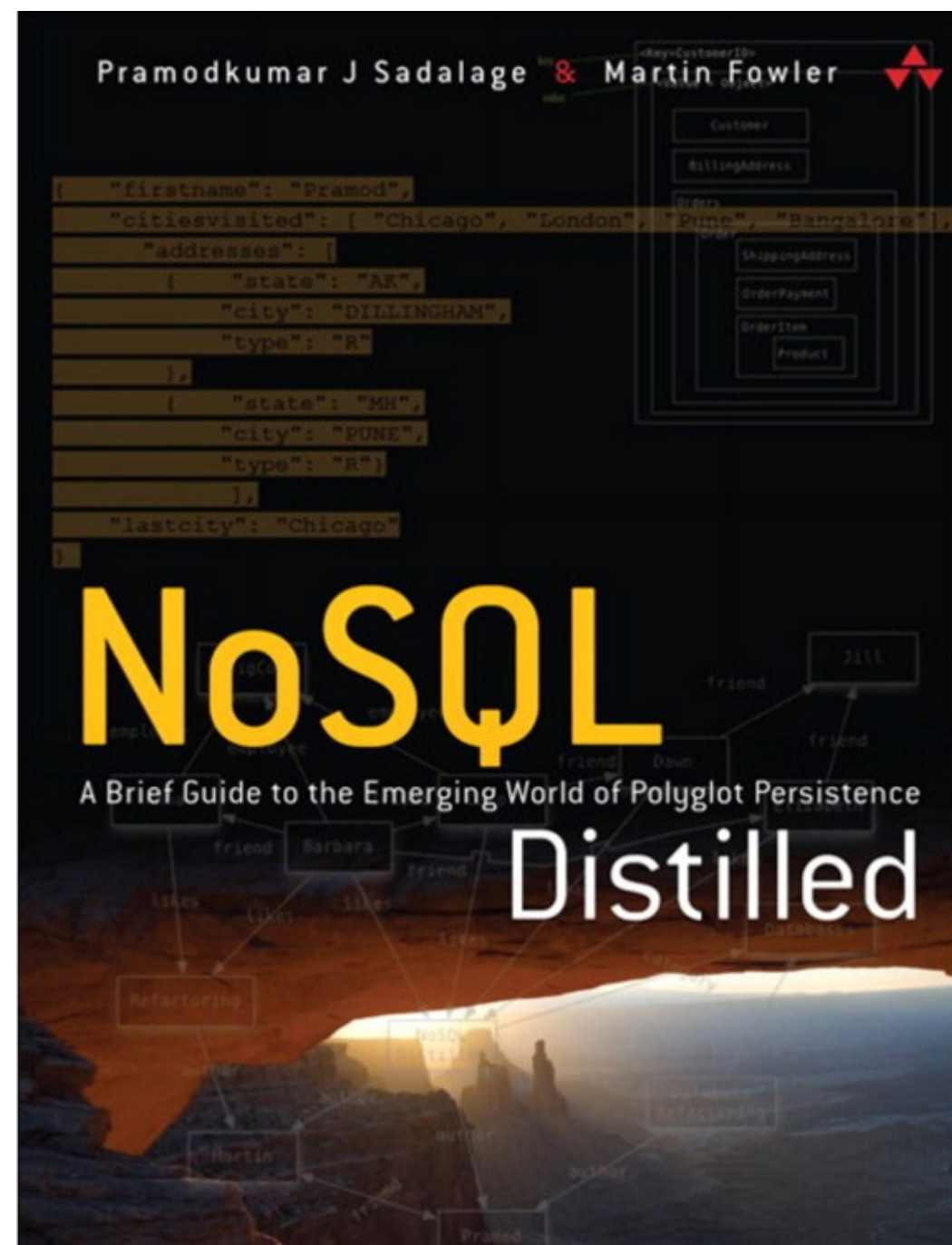
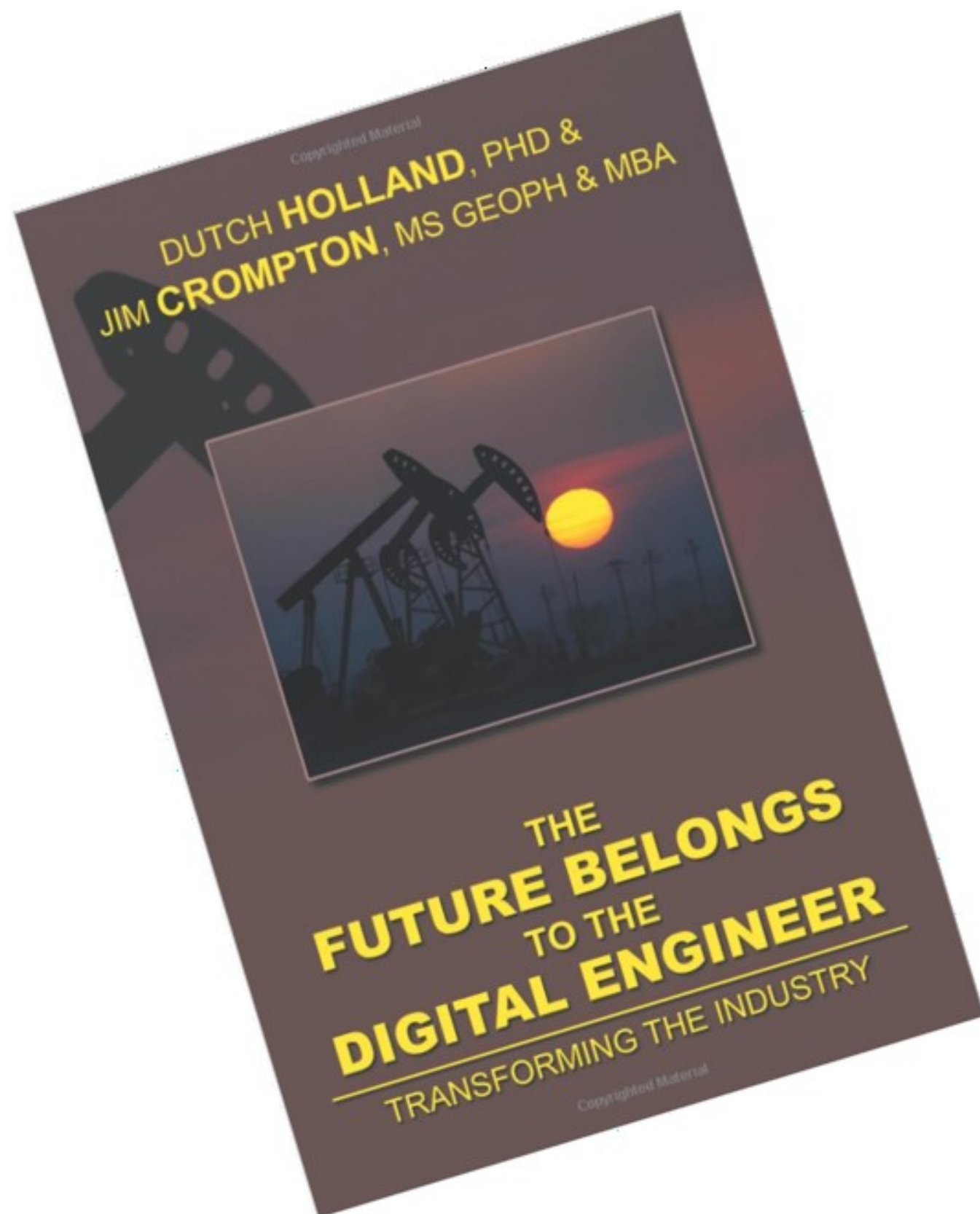


exhaust

Wells

Separators

Export



Key -> Value

Key -> Doc

Column Family

Graph

~Real Time

Riak
Redis
Memcached DB
Berkeley DB
Hamster DB
Amazon Dynamo
Voldemort
FoundationDB
LevelDB
Tokyo Cabinet


MongoDB
CouchDB
Terrastore
OrientDB
RavenDB
Elasticsearch

Cassandra
HBase
Hypertable
Amazon SimpleDB
Accumulo
HPCC
Cloudata



Neo4J
Infinite Graph
OrientDB
FlockDB
Gremlin
Titan

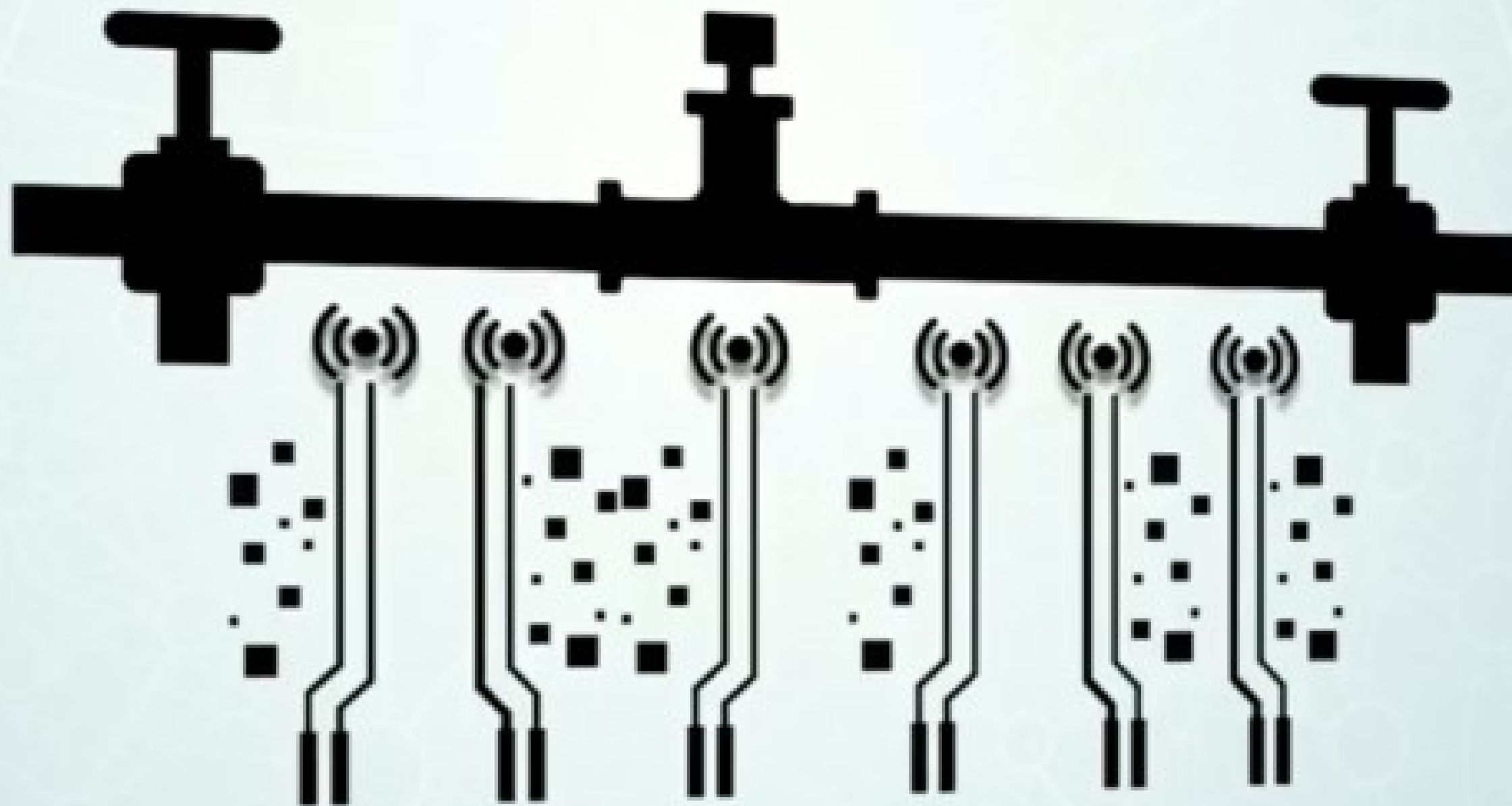
Storm
Impala
Stinger/Tez
Drill
Solr/Lucene

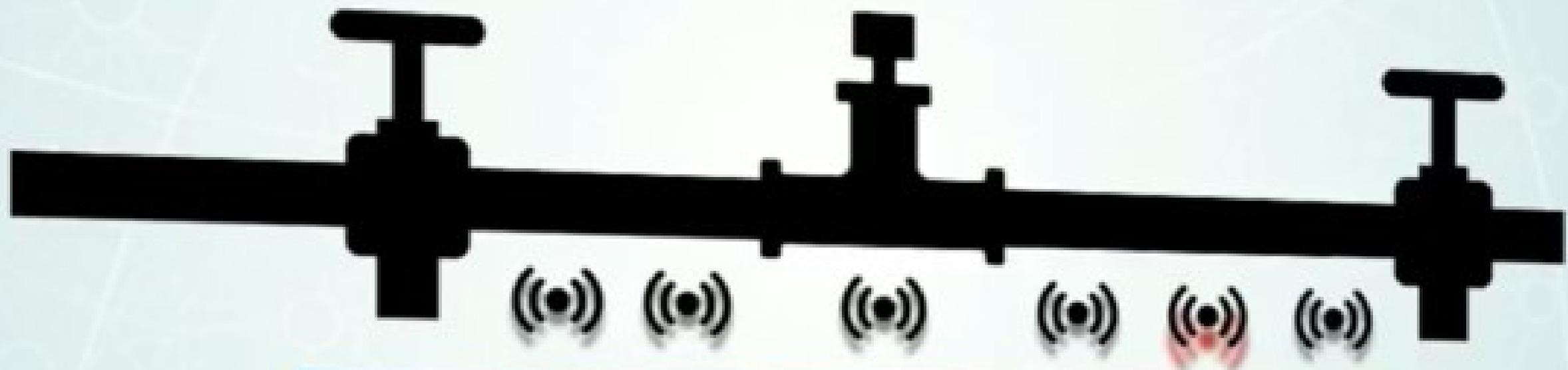


...there's an 87% chance
Linus Torvalds hates your code.

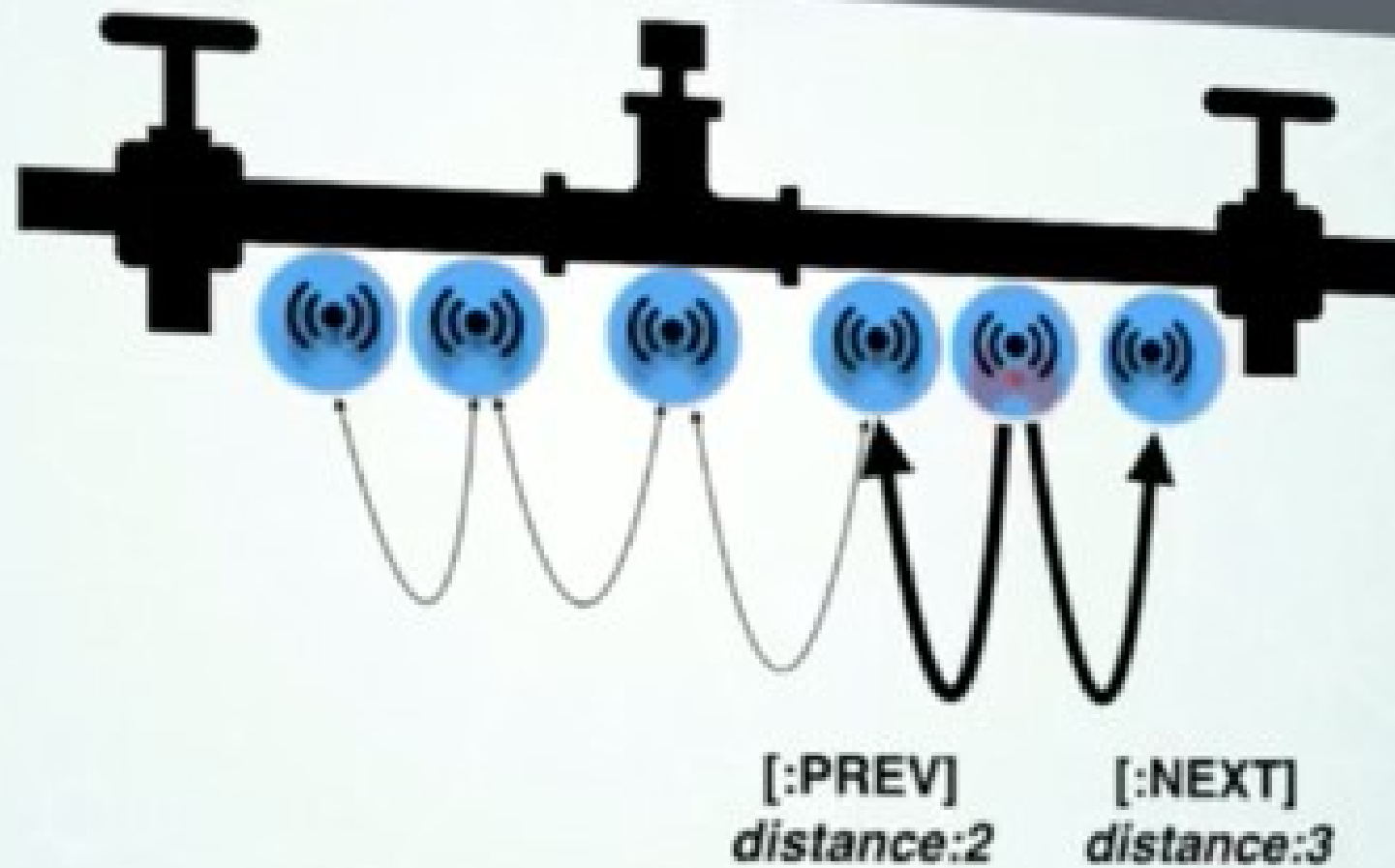
"Bad programmers **worry about the code**. Good programmers **worry about data structures** and their relationships."

What about the
Relationship Between
Sensors?





	T1	T2	T3	T4	T5	T6
09:00:00	40	41	40	40	67	40
09:00:01	37	41	40	40	67	40
09:00:02	40	41	40	40	67	40
09:00:03	30	41	40	40	67	40



MATCH

(sp:Sensor) <- [rp:PREV] - (s:Sensor {id:1501}) - [rn:NEXT] -> (sn:Sensor)

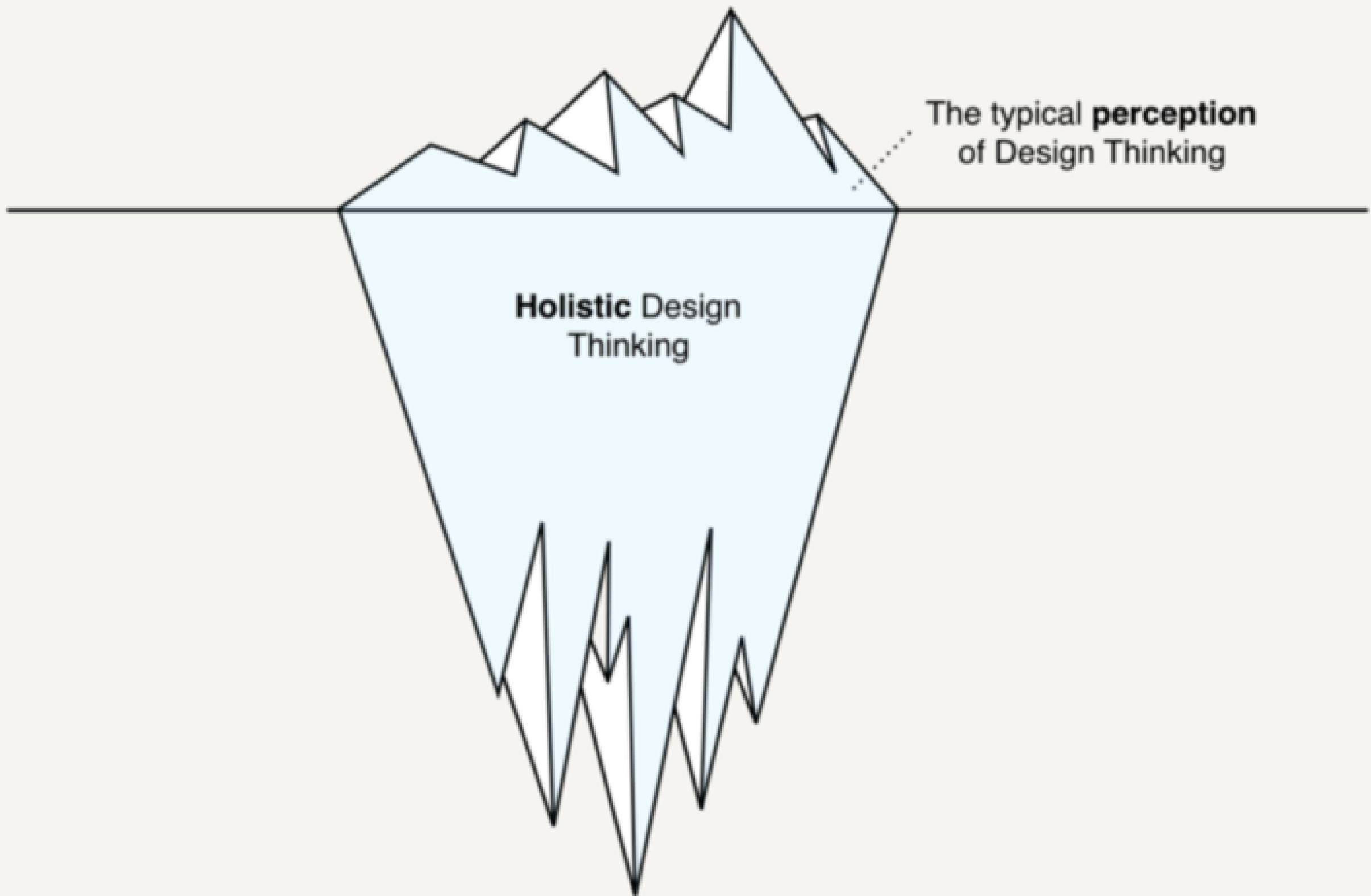
WHERE

rp.distance < 5 AND rn.distance < 5

RETURN

sp.temp,

sn.temp

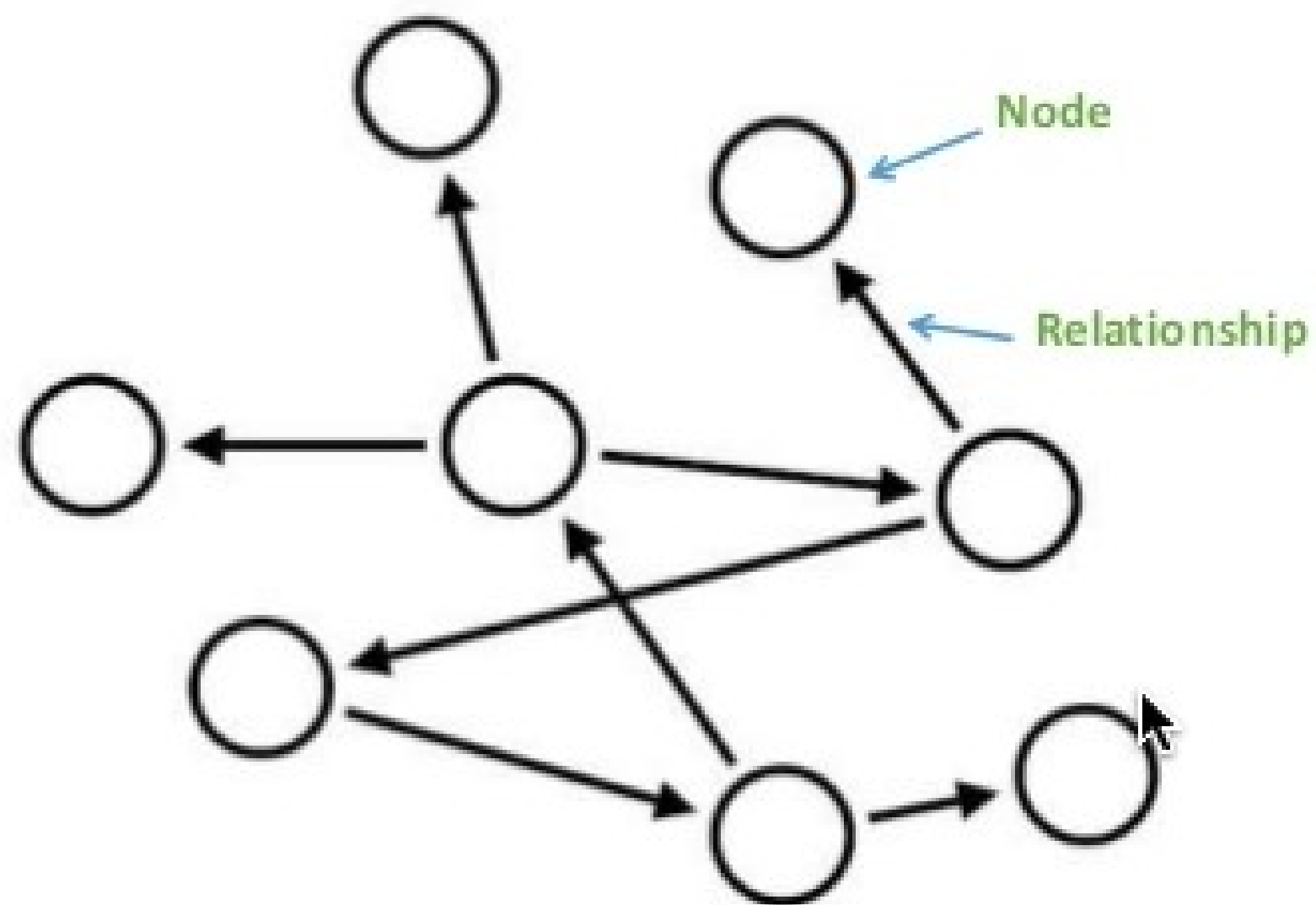


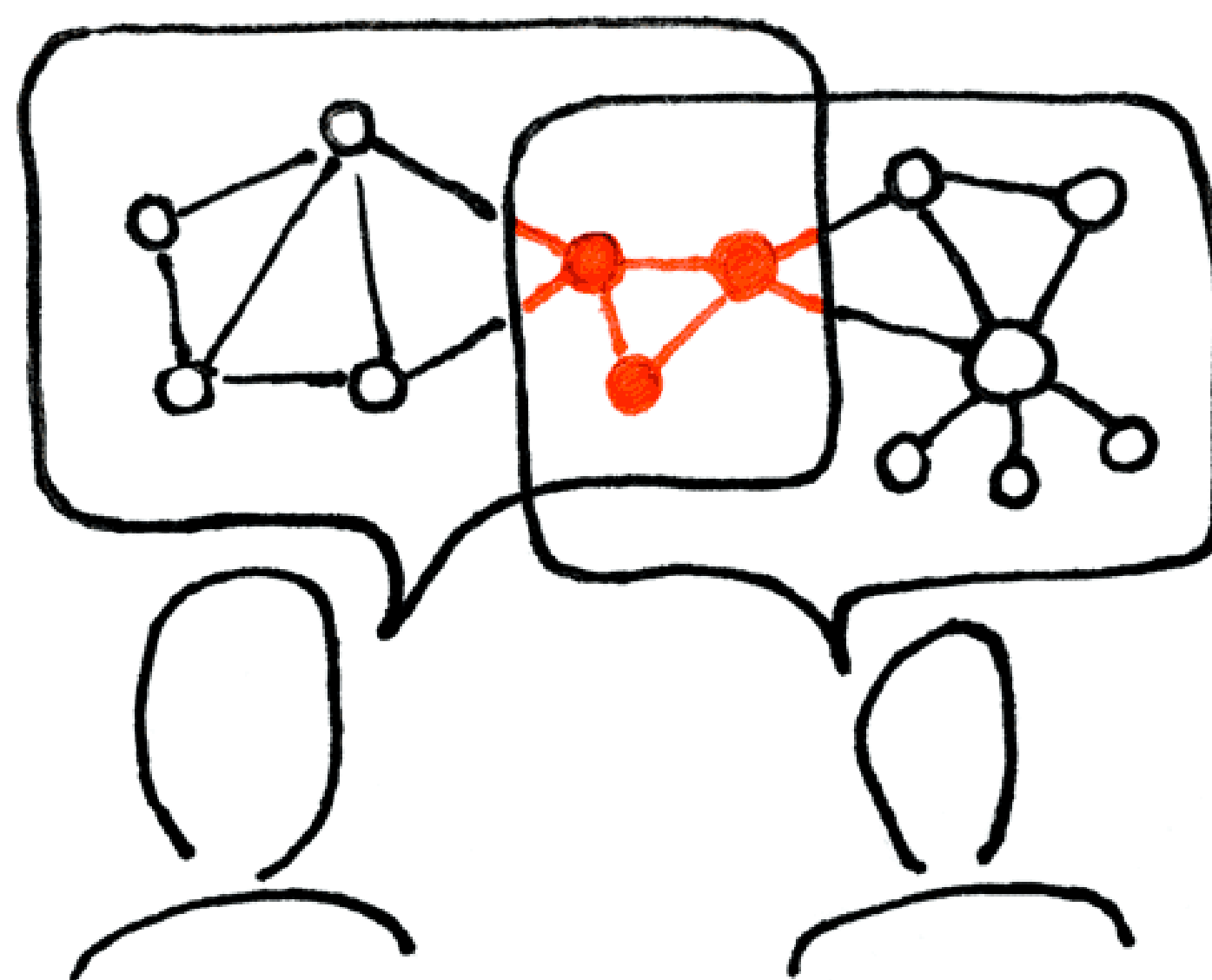


3DCuts.com

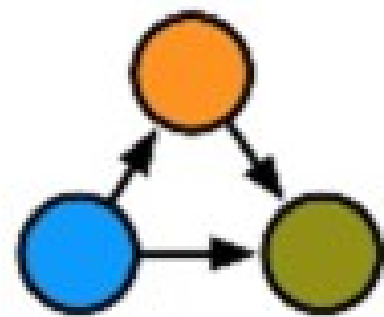




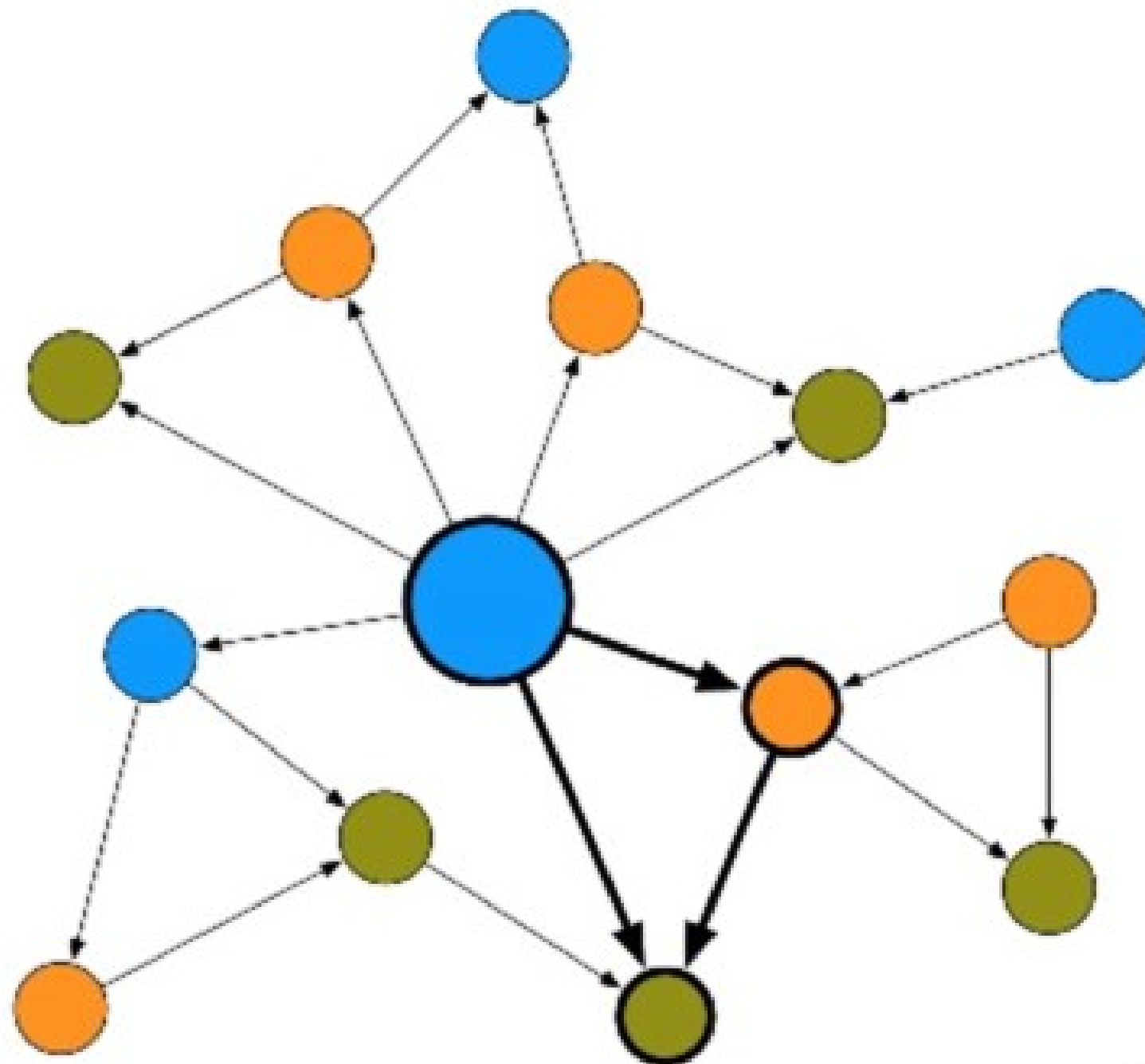




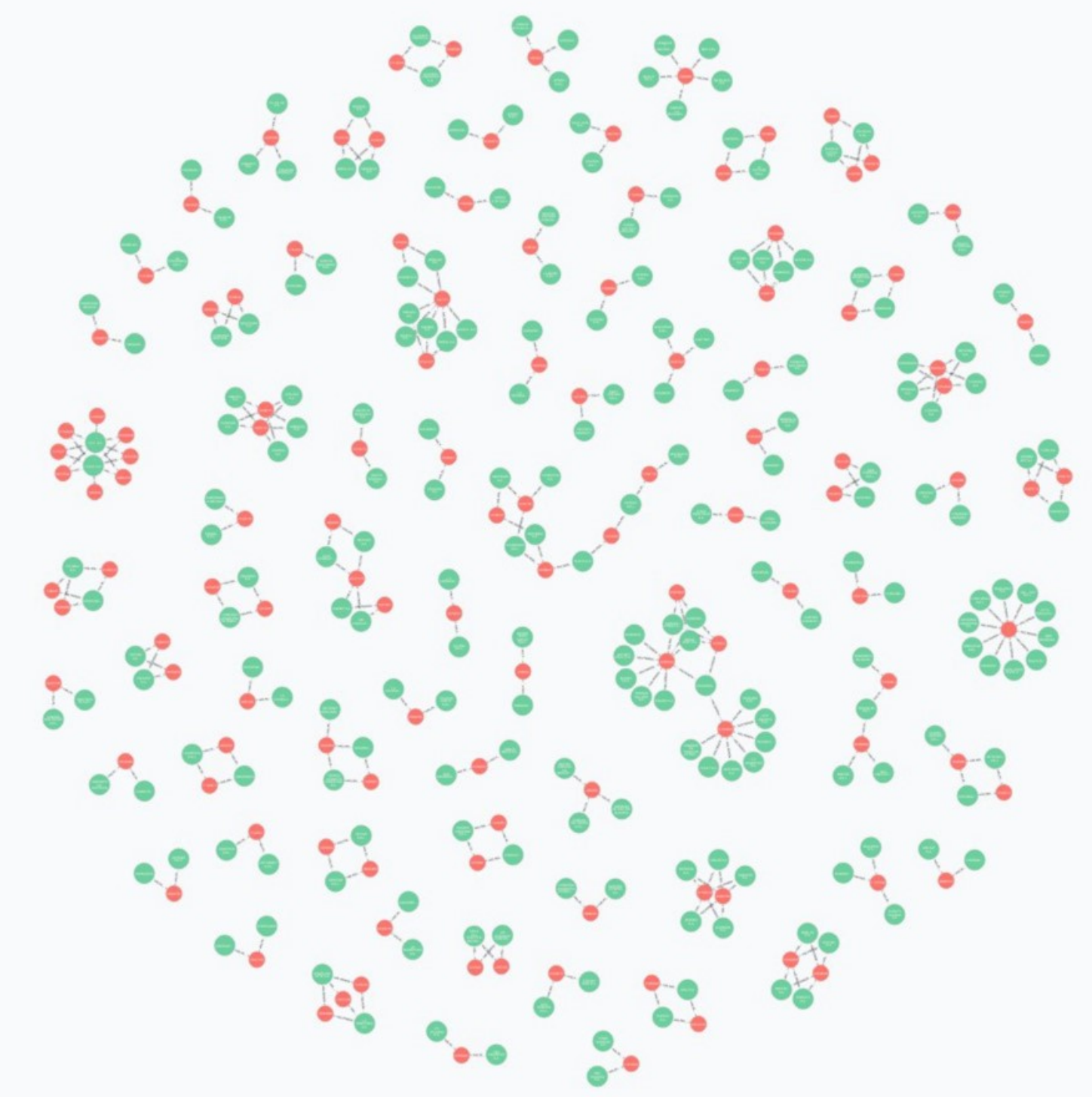
Match



Pattern



#neo4j

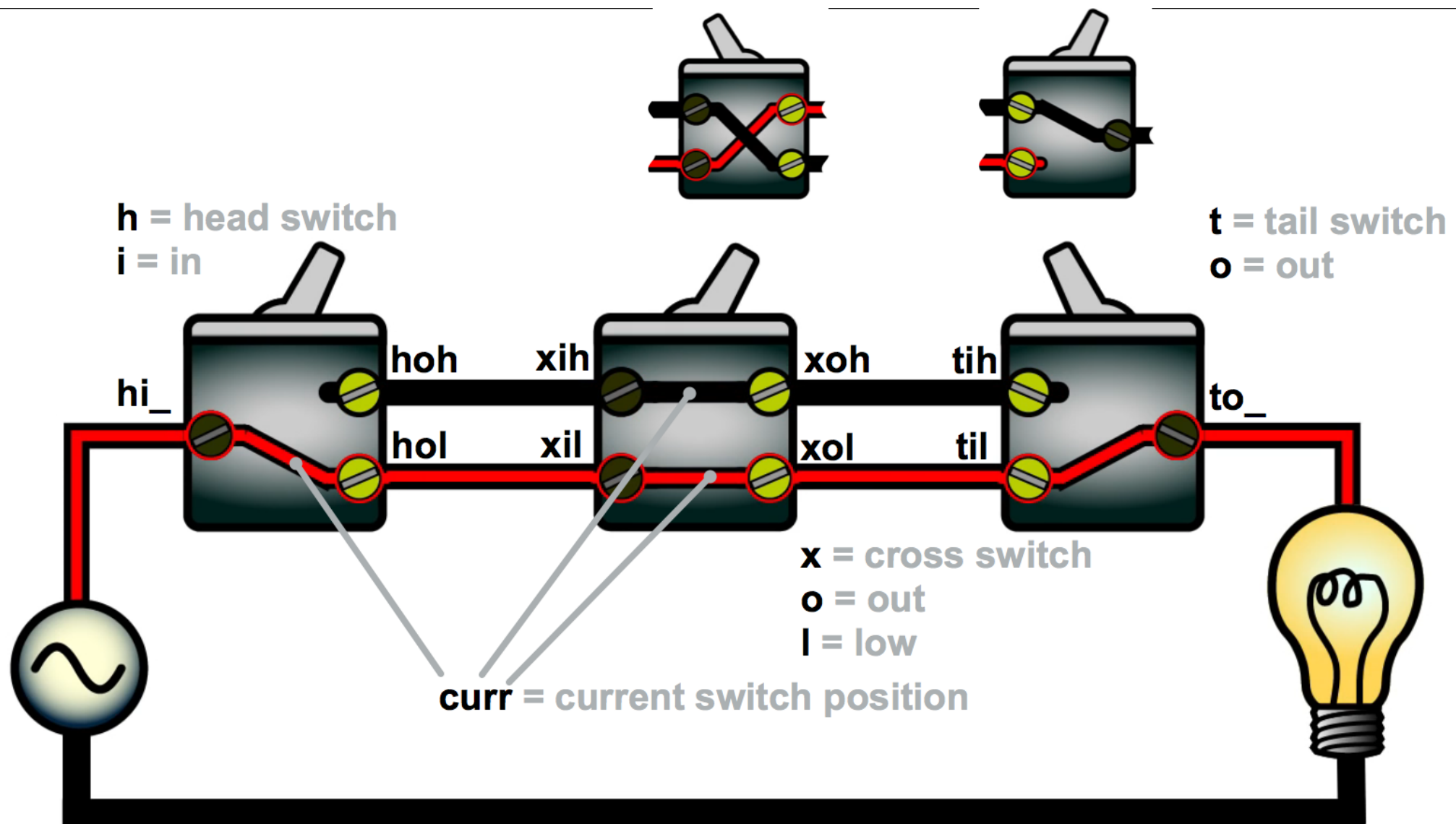


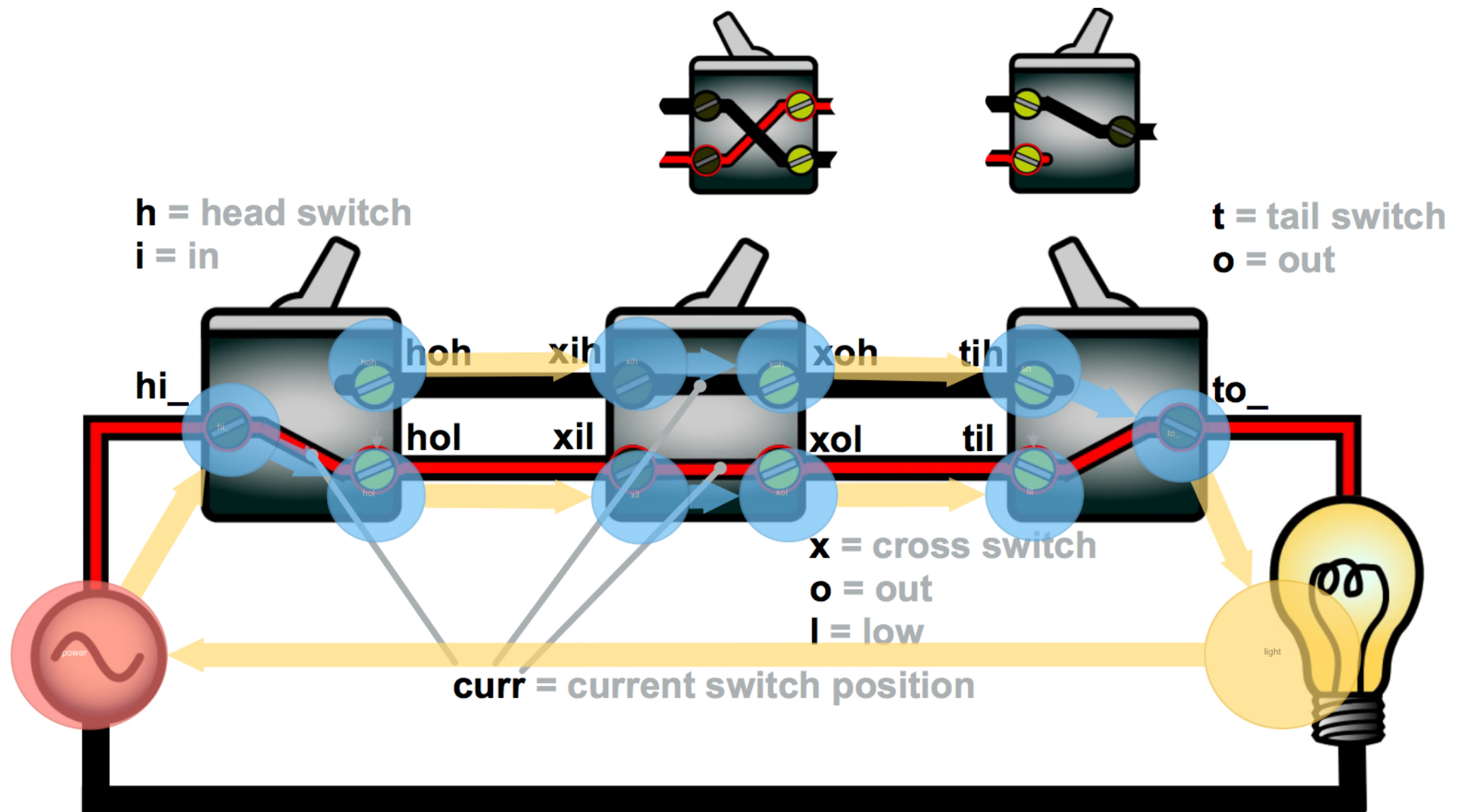


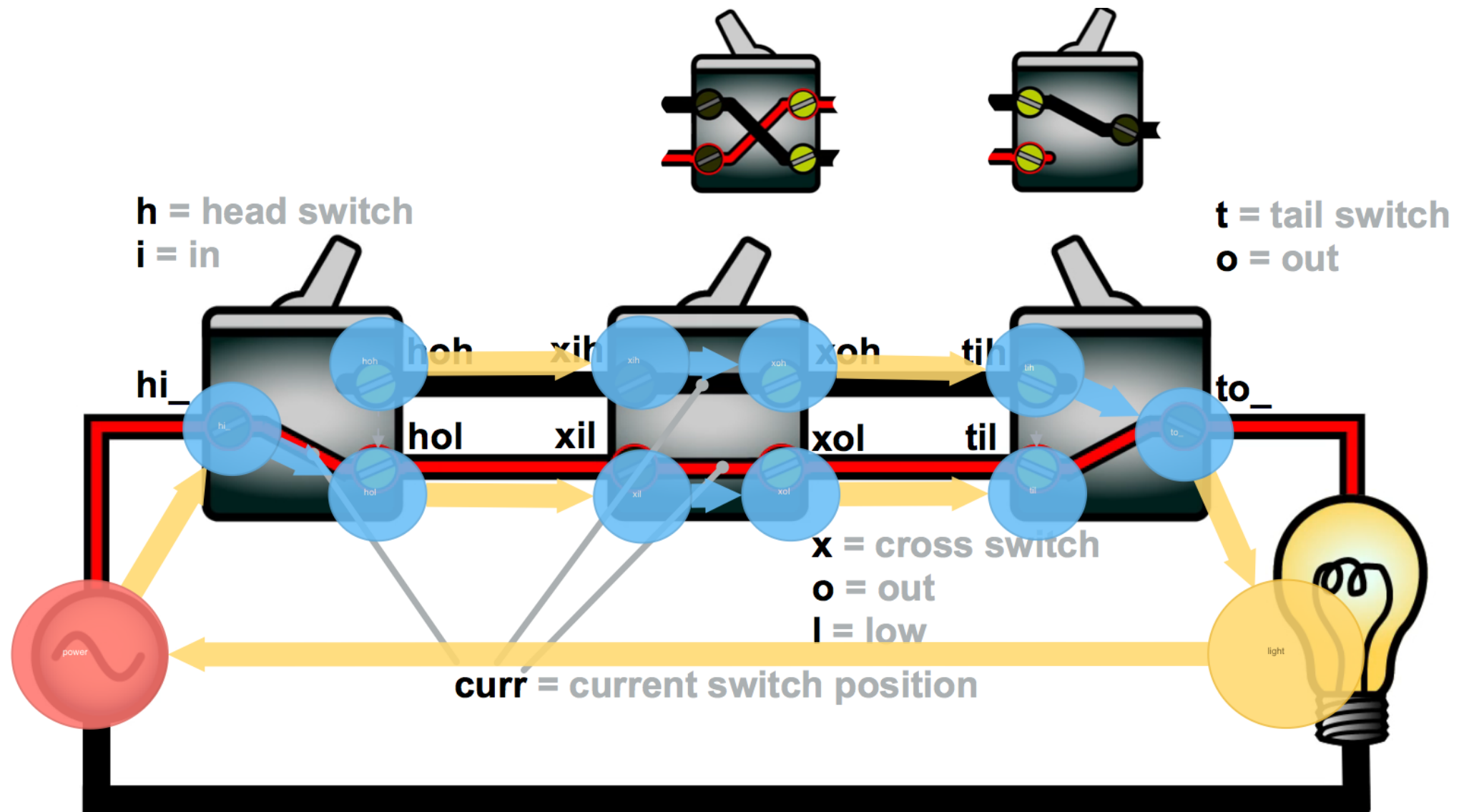
3 relationships.

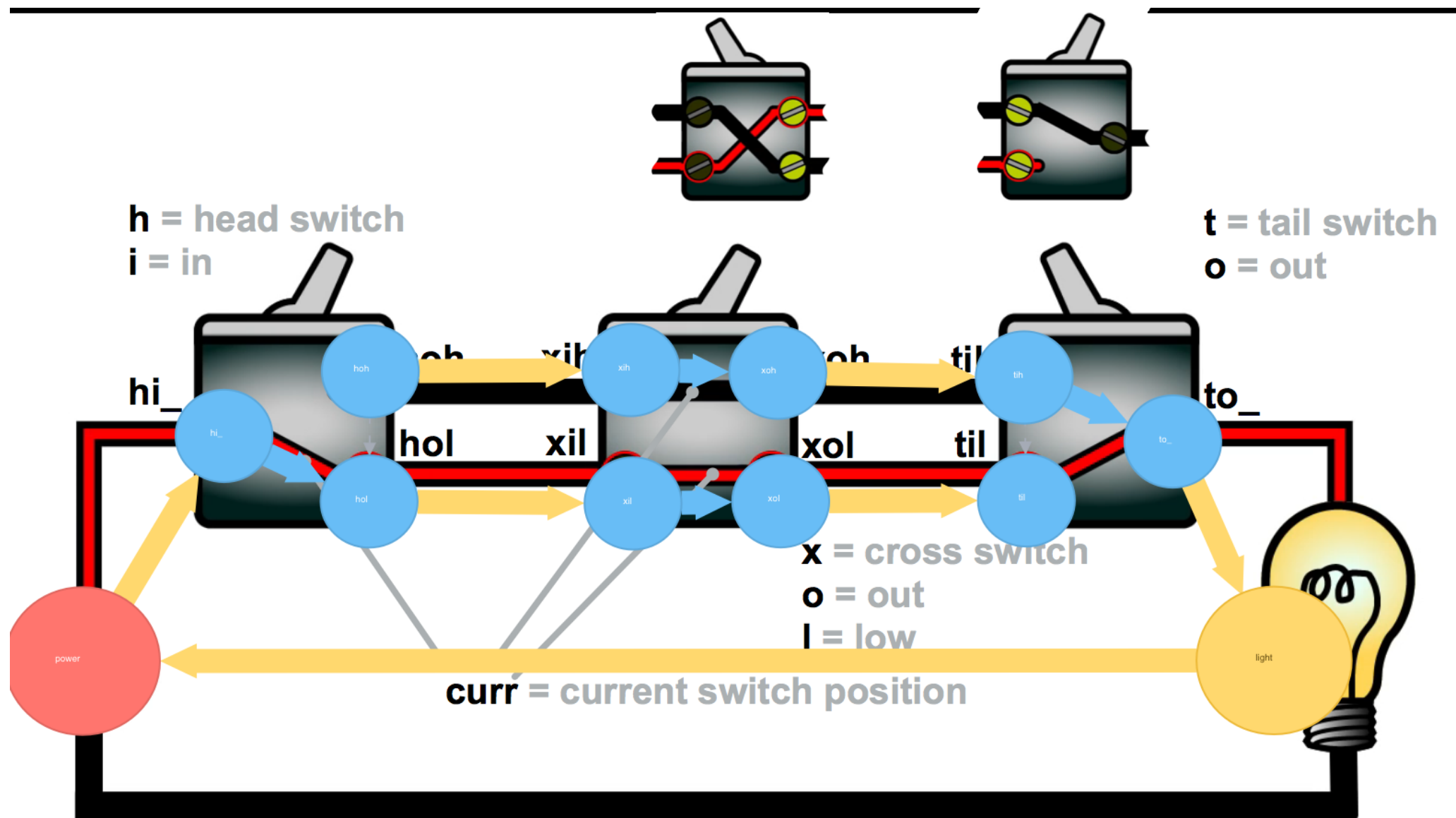
A true story.

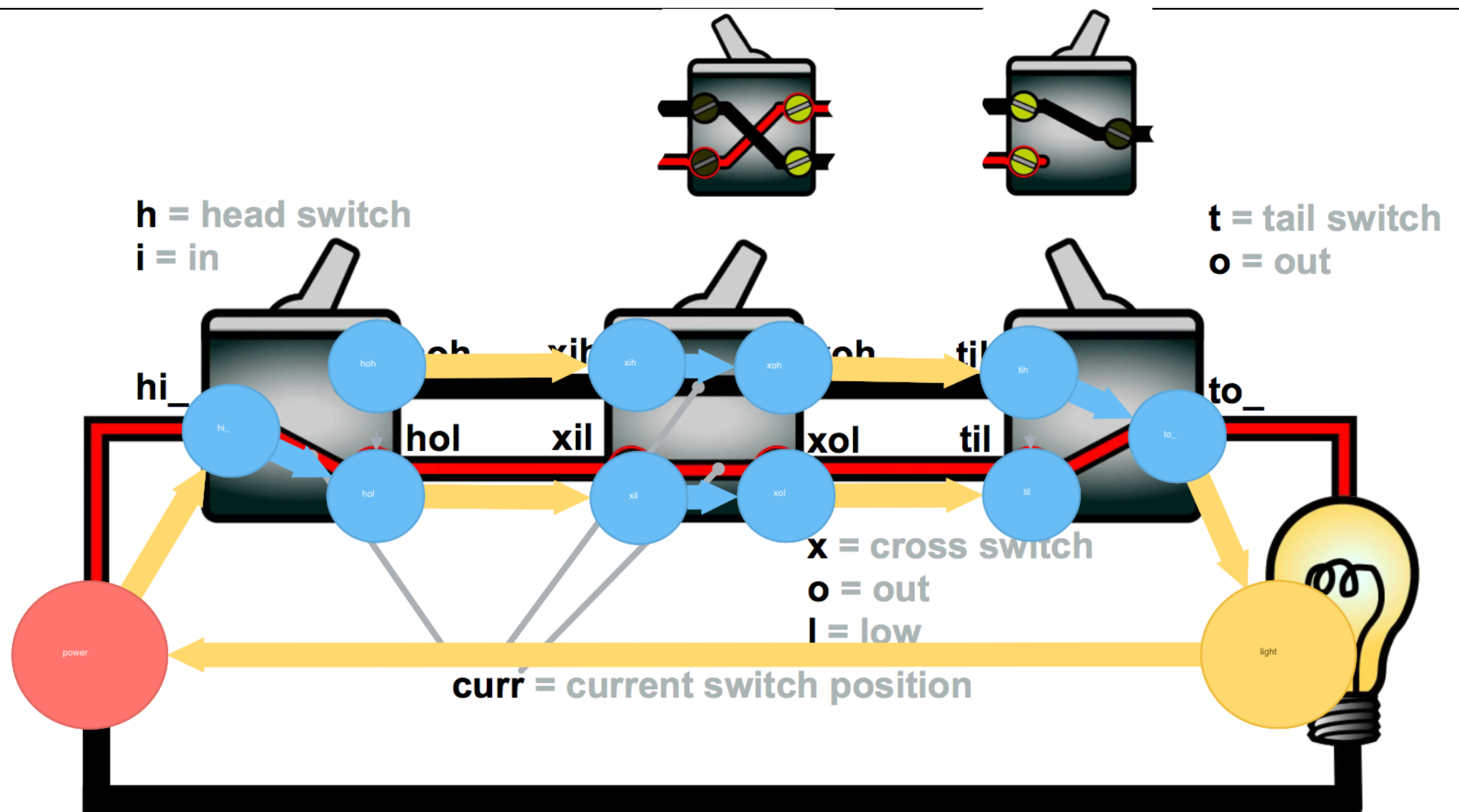


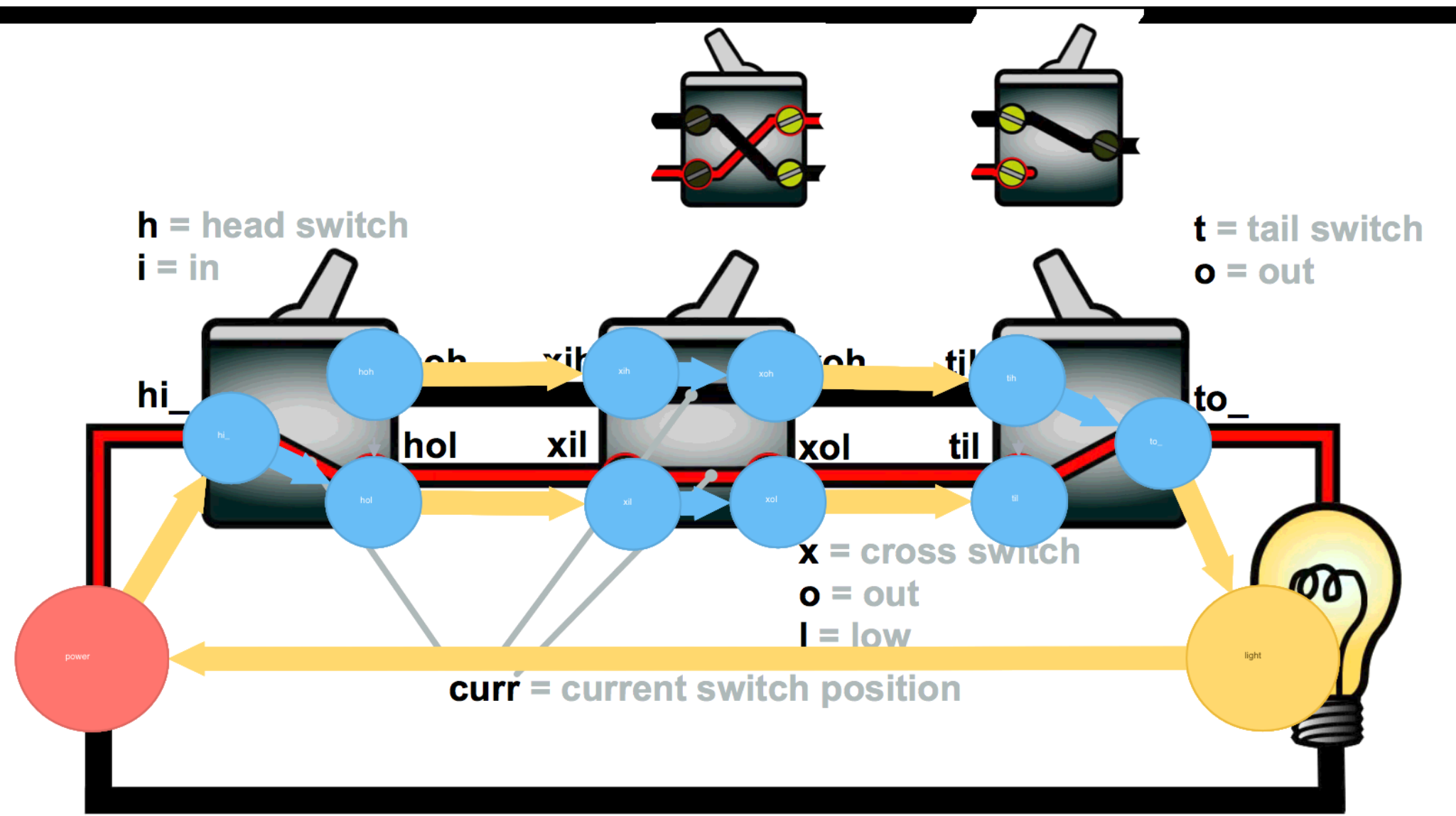


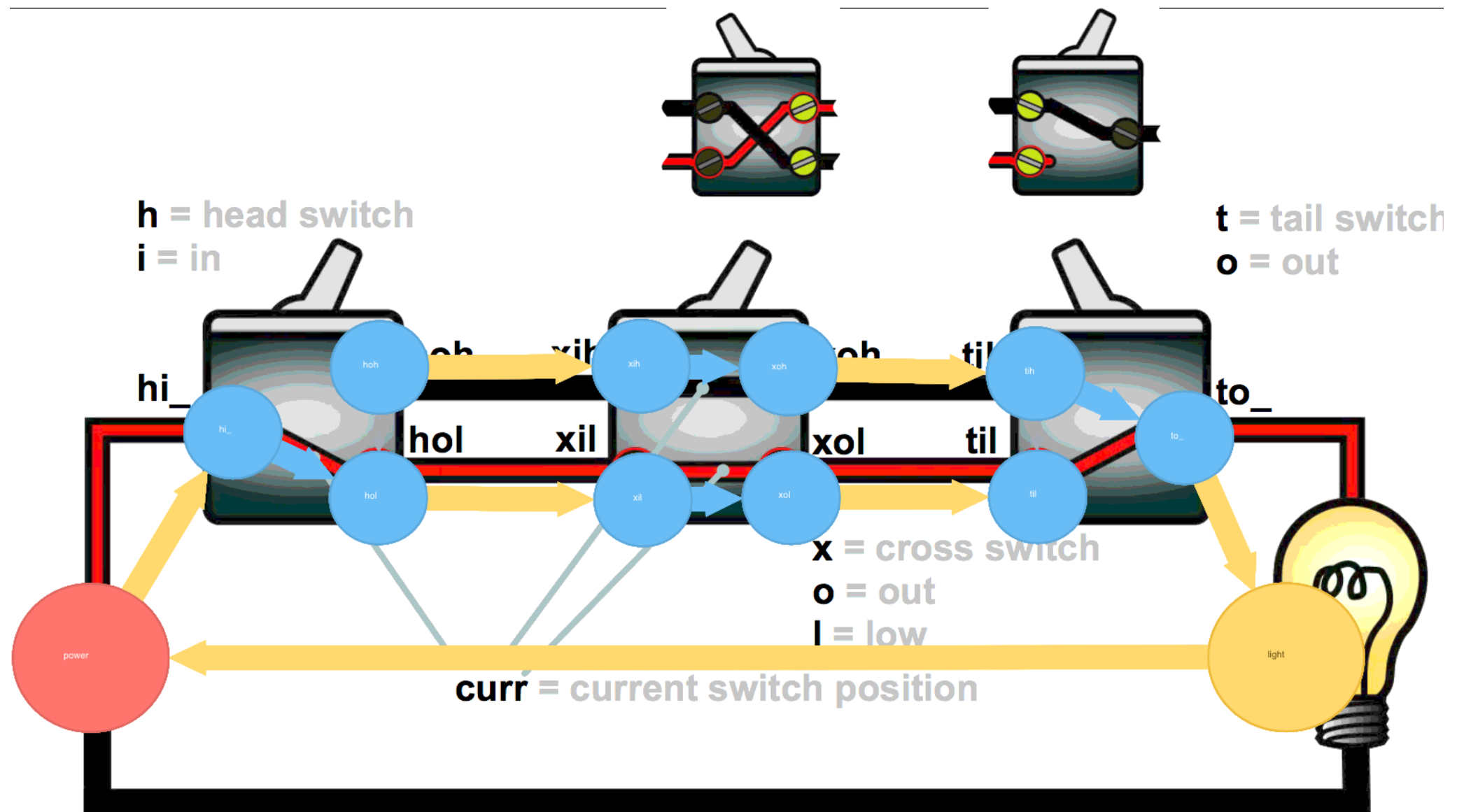


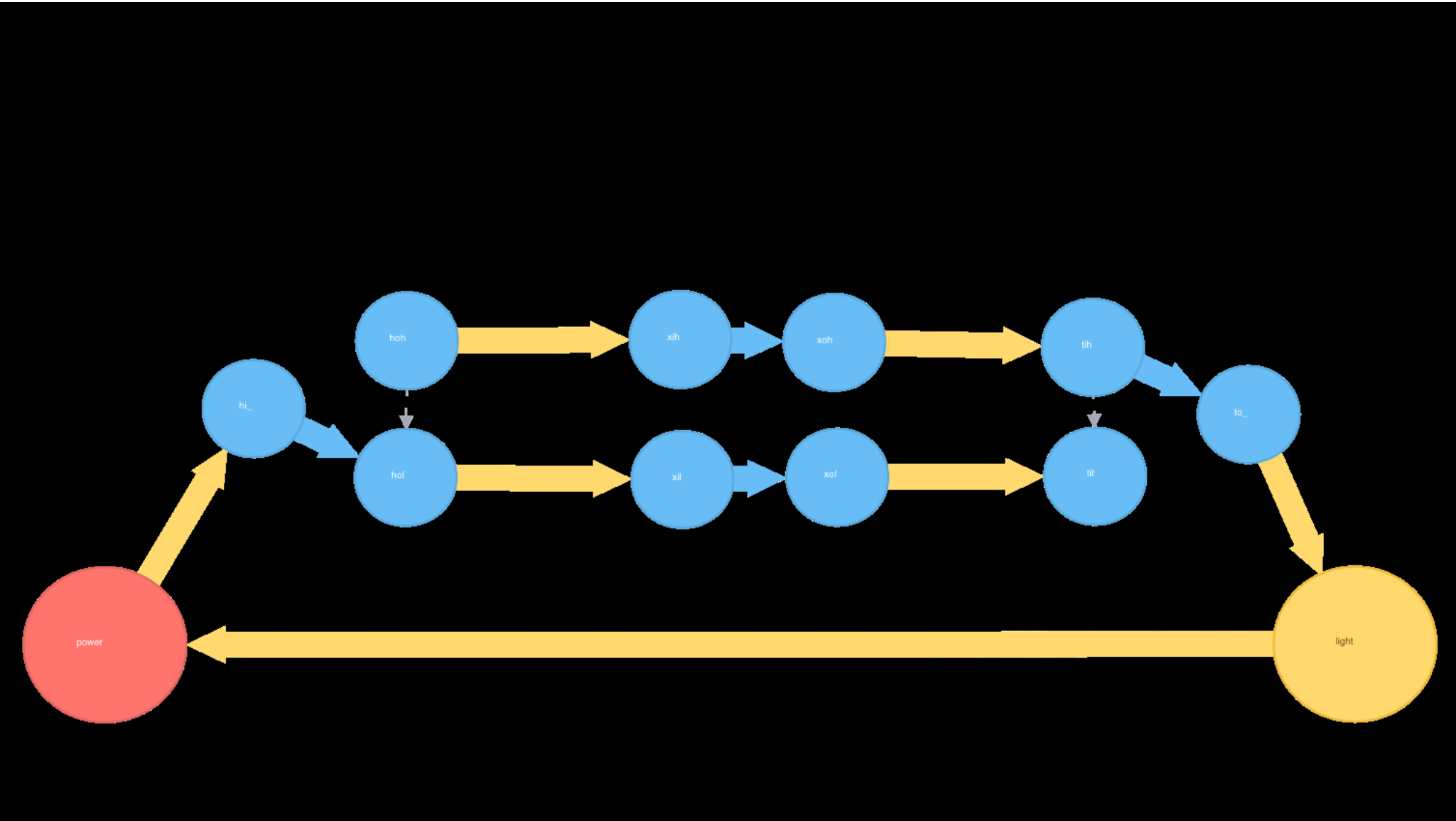


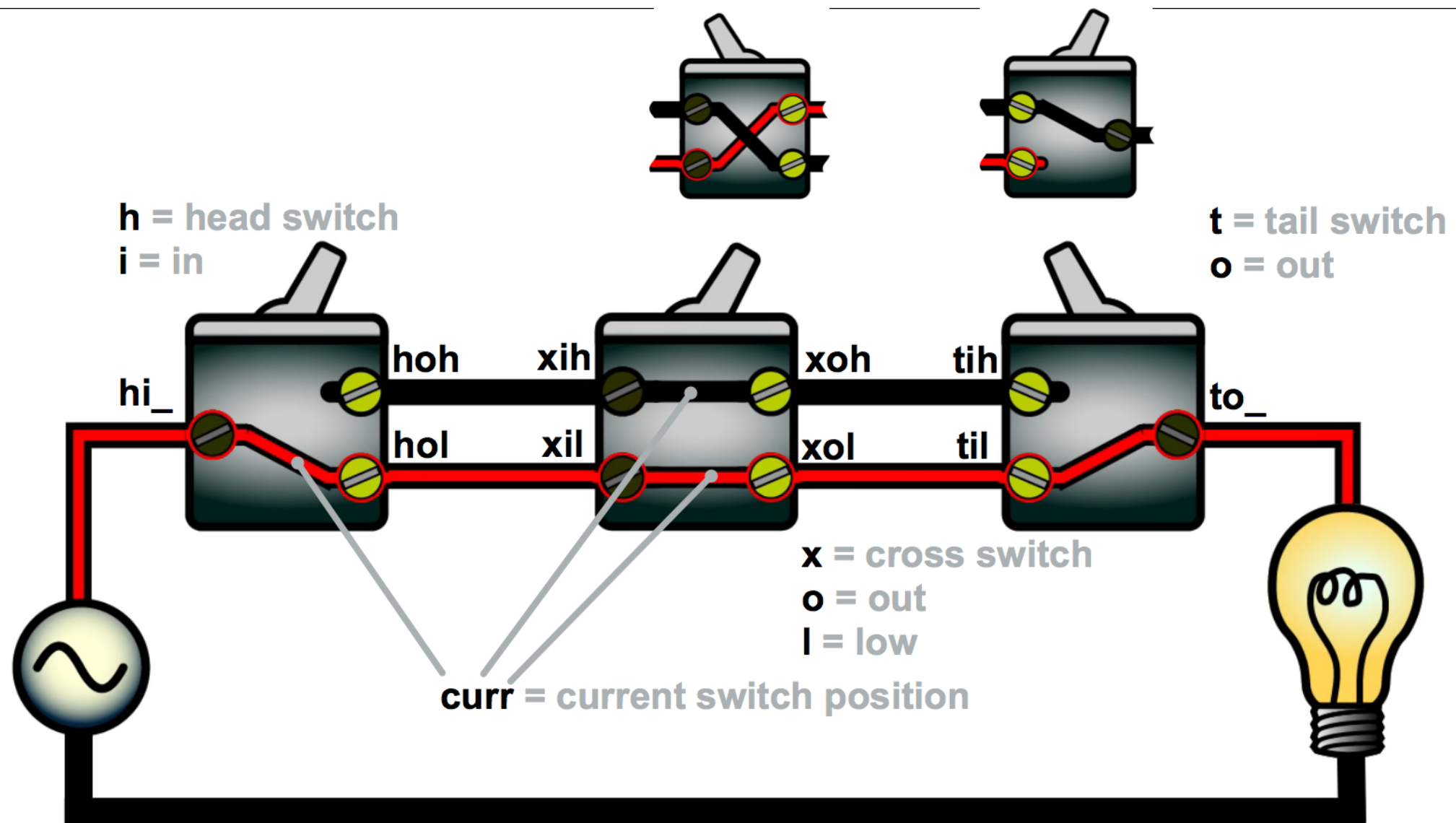


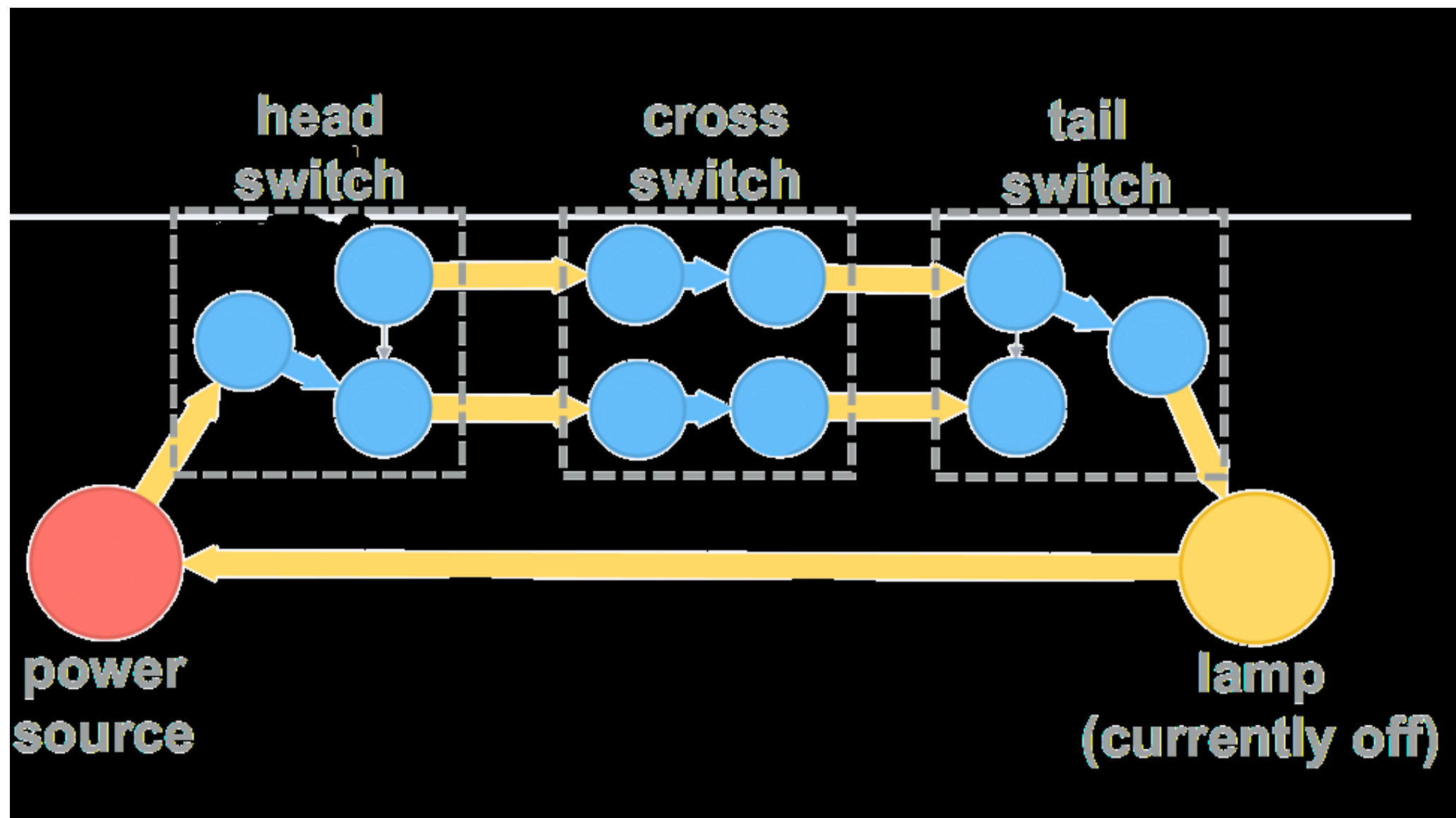


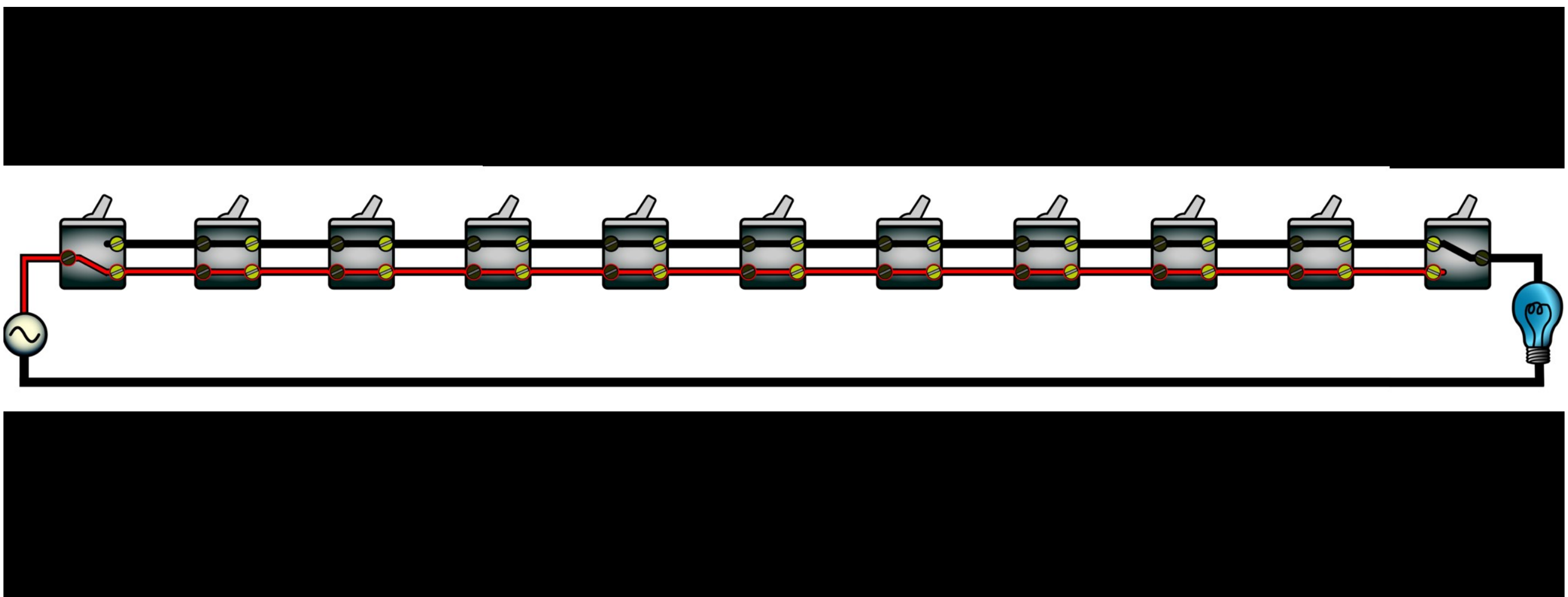








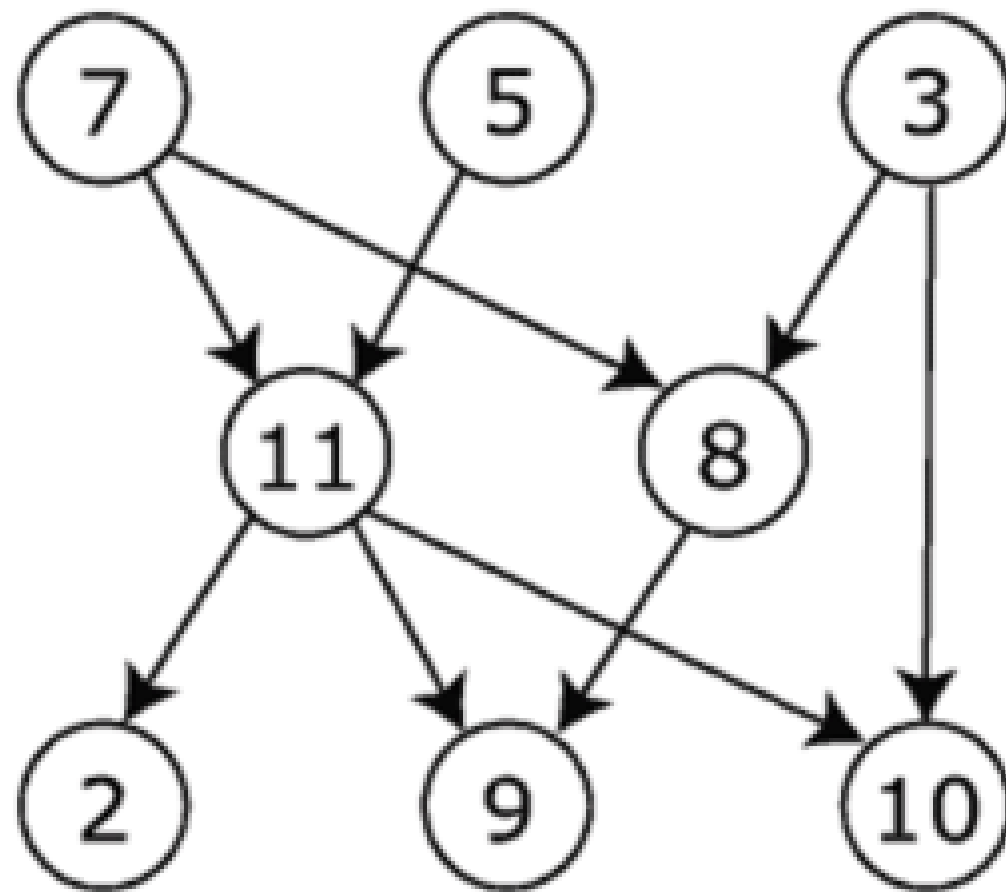








“Graph” is a mathematical concept, which was first described in 1736 by Euler. Do not think pie charts, the picture in your head should look like this:



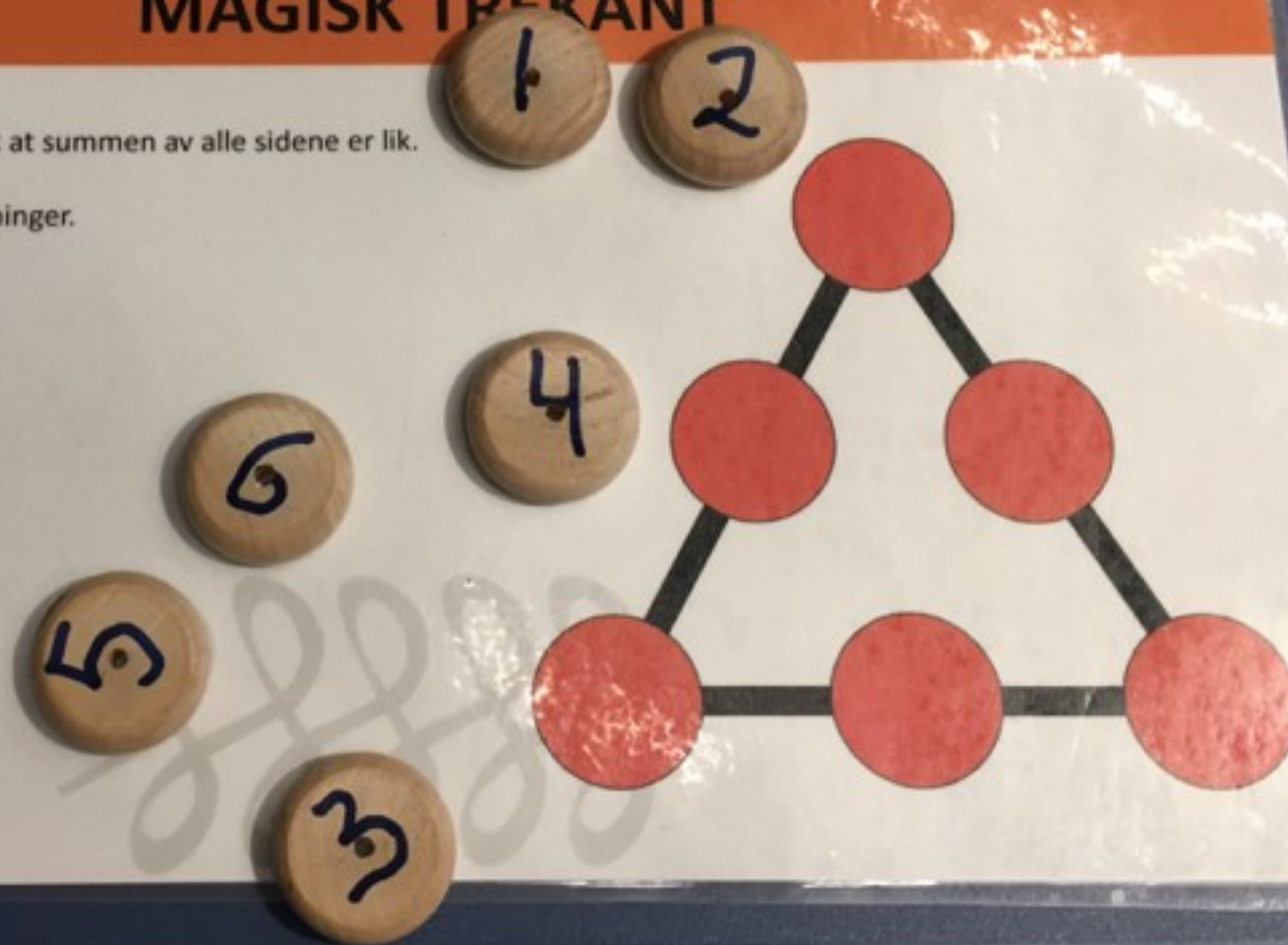
$$\frac{N}{Z} = \left\{ \frac{\beta}{\mu} \right\} a^{\beta}$$

VITENFABRIKKEN
JERNMUSEET

MAGISK TREKANT

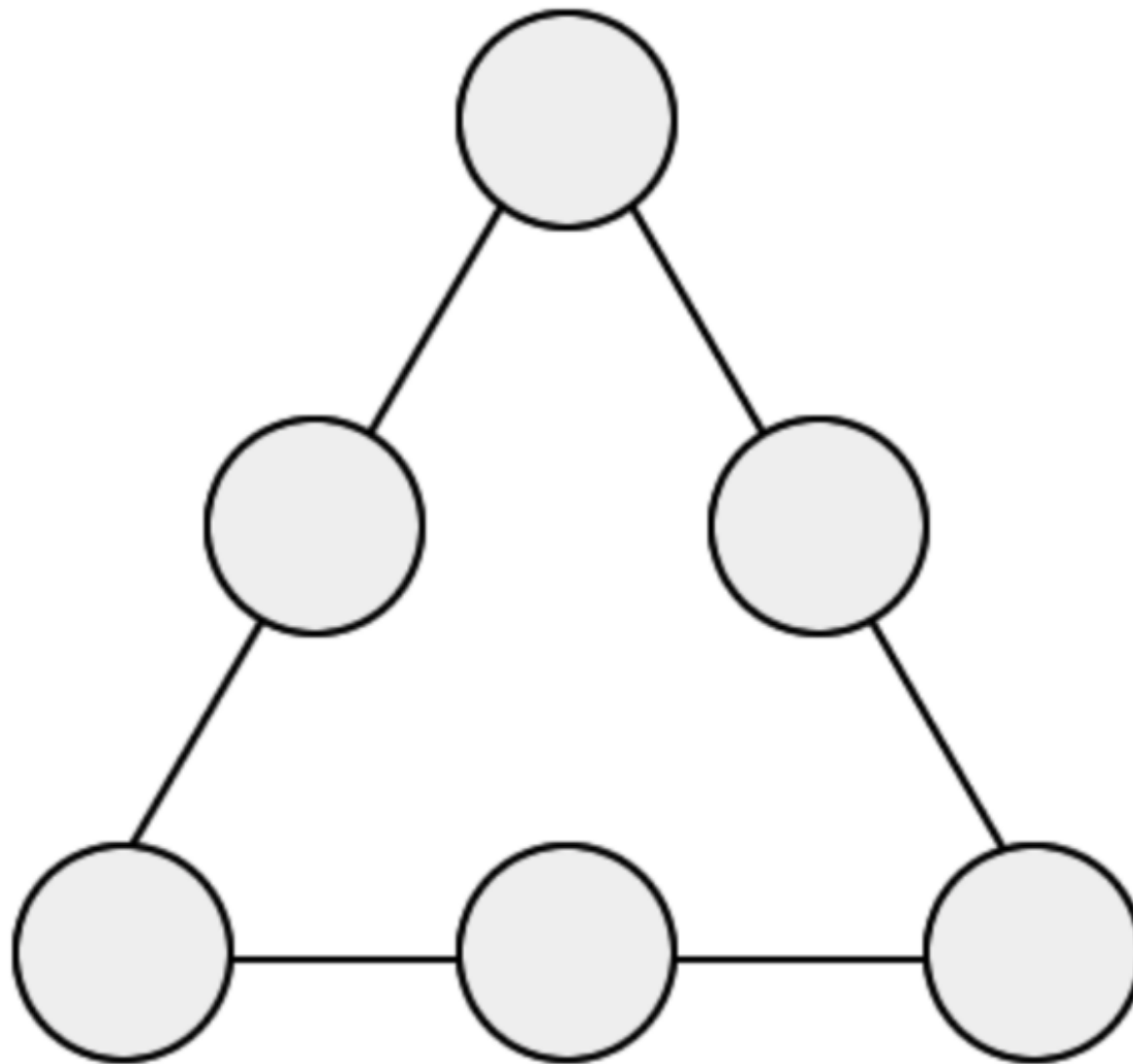
Legg tallene 1-6 slik at summen av alle sidene er lik.

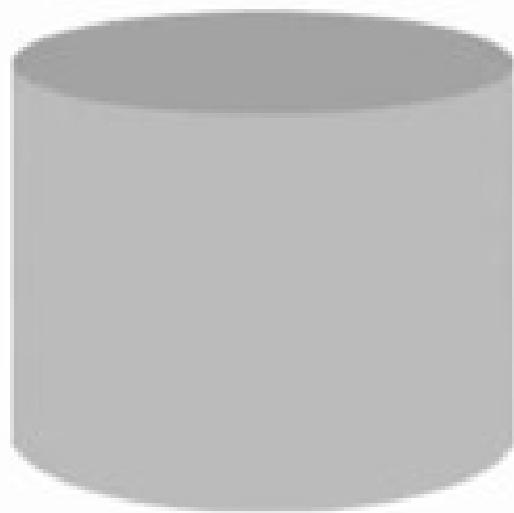
Det finnes flere løsninger.



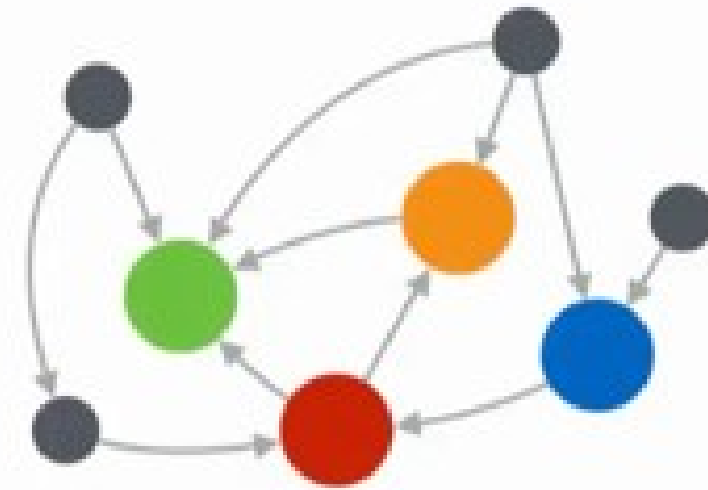
Magic Triangle

Directions: Arrange the numbers 1, 2, 3, 4, 5, 6 so each row has a sum of 9. Use all six numbers and each number once.



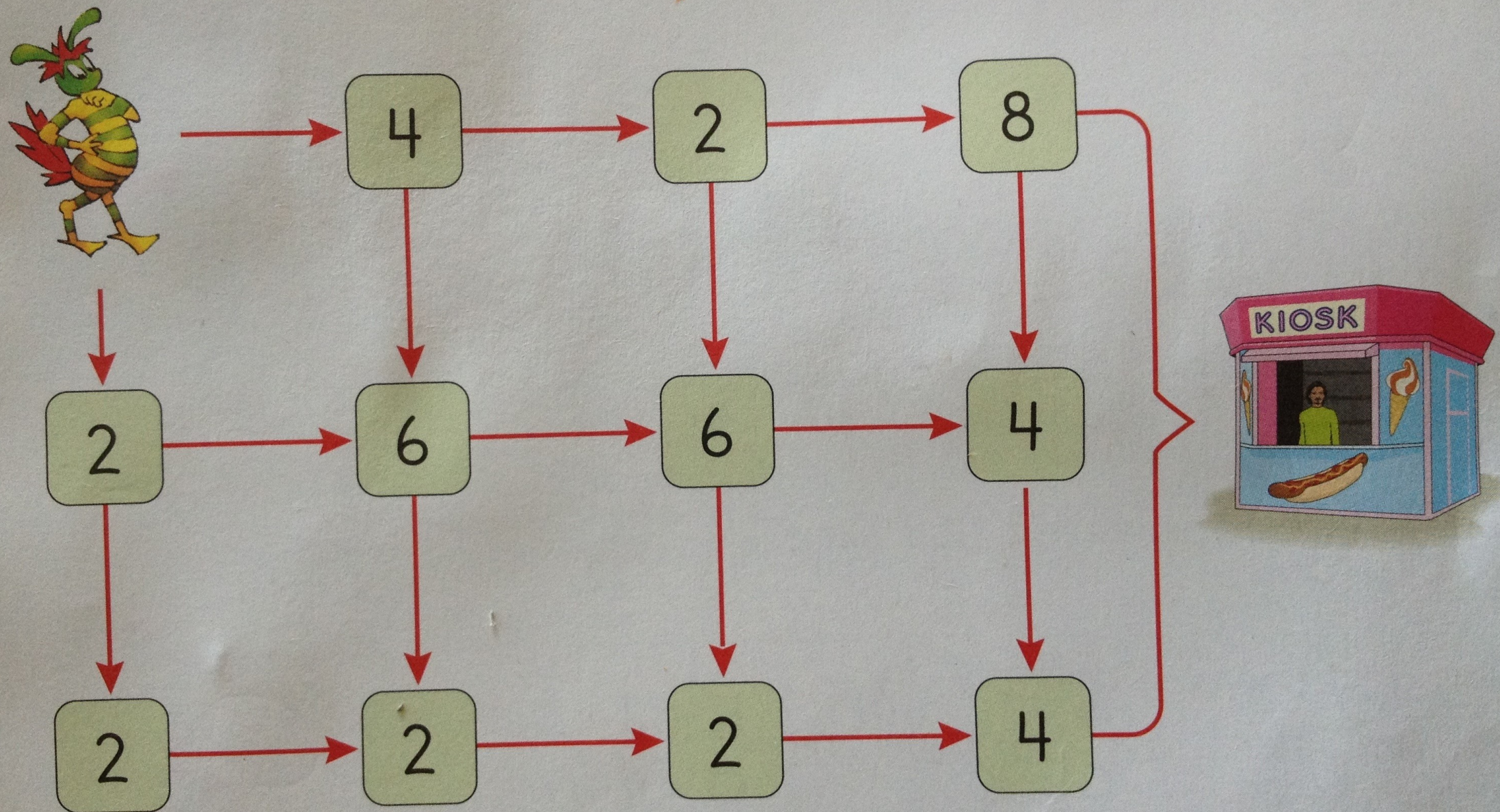


Store / Retrieve



Actionable Insights

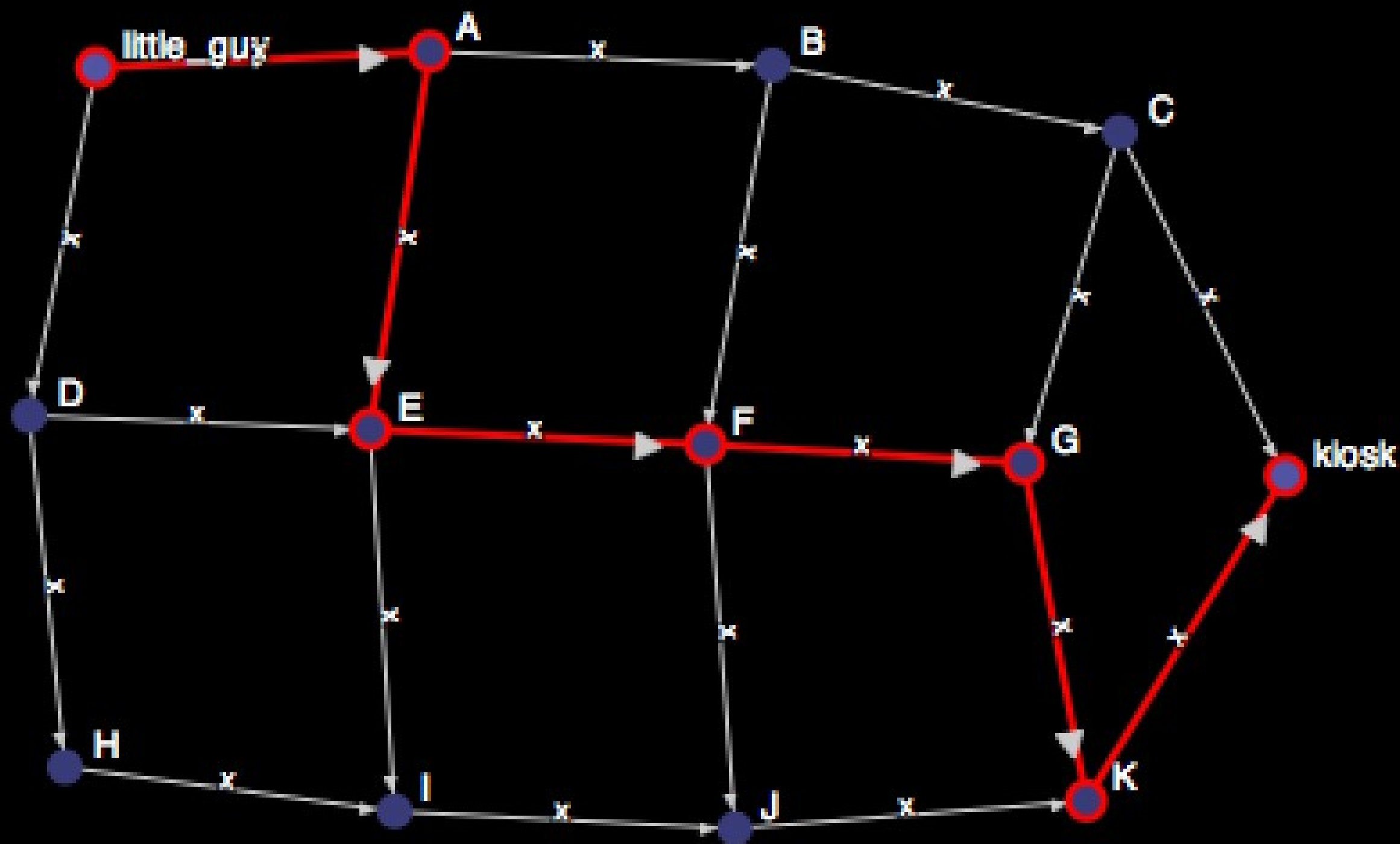
Følg pilene, og lag en vei til kiosken der summen blir 24.

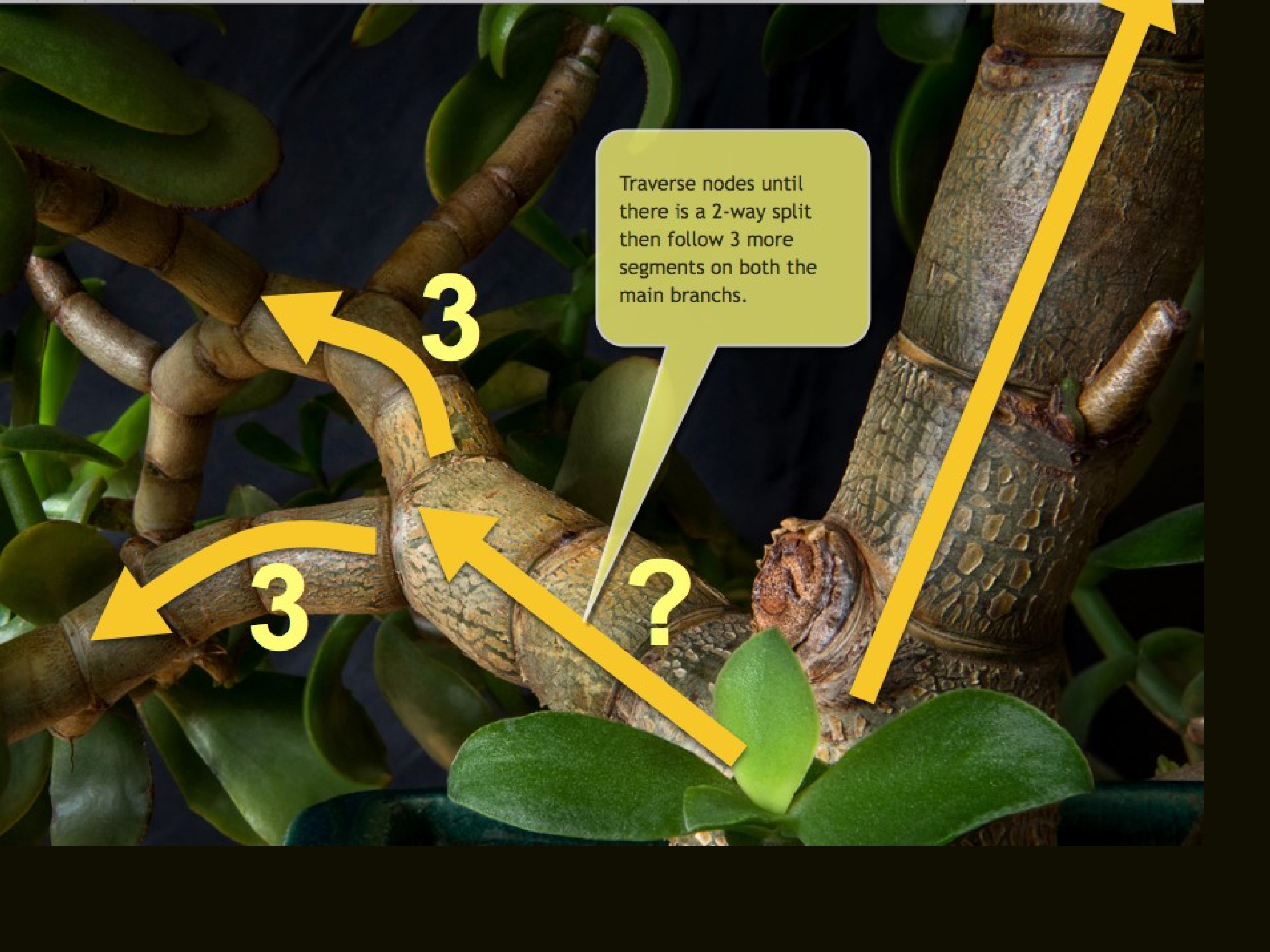



```

#clear existing graph~
START n=node(*) MATCH n-[r?]->m DELETE r,m~
~
~
START n=node(0) ~
CREATE (A{name:'A'}), (B{name:'B'}), (C{name:'C'}), (D{name:'D'}), (E{name:'E'}), (F{name:'F'}),
(G{name:'G'}), (H{name:'H'}), (I{name:'I'}), (J{name:'J'}), (K{name:'K'}), (kiosk{name:'kiosk'}),~
n-[x]->A, n-[x]->D, A-[x]->B, A-[x]->E, B-[x]->C, B-[x]->F, C-[x]->G, D-[x]->E, D-[x]->H,
E-[x]->F, E-[x]->I, F-[x]->G, F-[x]->J, G-[x]->K, H-[x]->I, I-[x]->J, J-[x]->K, C-[x]->kiosk,
K-[x]->kiosk~
SET n.name = 'little_guy', A.number = 4, B.number = 2, C.number = 8, D.number = 2, E.number = 6,
F.number = 6, G.number = 4, H.number = 2, I.number = 2, J.number = 2, K.number = 4~
~
~
START n=node(0), kiosk=node:node_auto_index(name = 'kiosk') ~
MATCH p = n-[*]->i-[*]->(kiosk) ~
WITH p, SUM(i.number) AS total, COLLECT(i.name) AS name_sequence, COLLECT(i.number) AS number_sequence
WHERE total=24 ~
RETURN p, name_sequence, number_sequence, total~
~

```





Traverse nodes until
there is a 2-way split
then follow 3 more
segments on both the
main branches.

[x IN range(1,10) WHERE x % 2 = 0 | x*x]

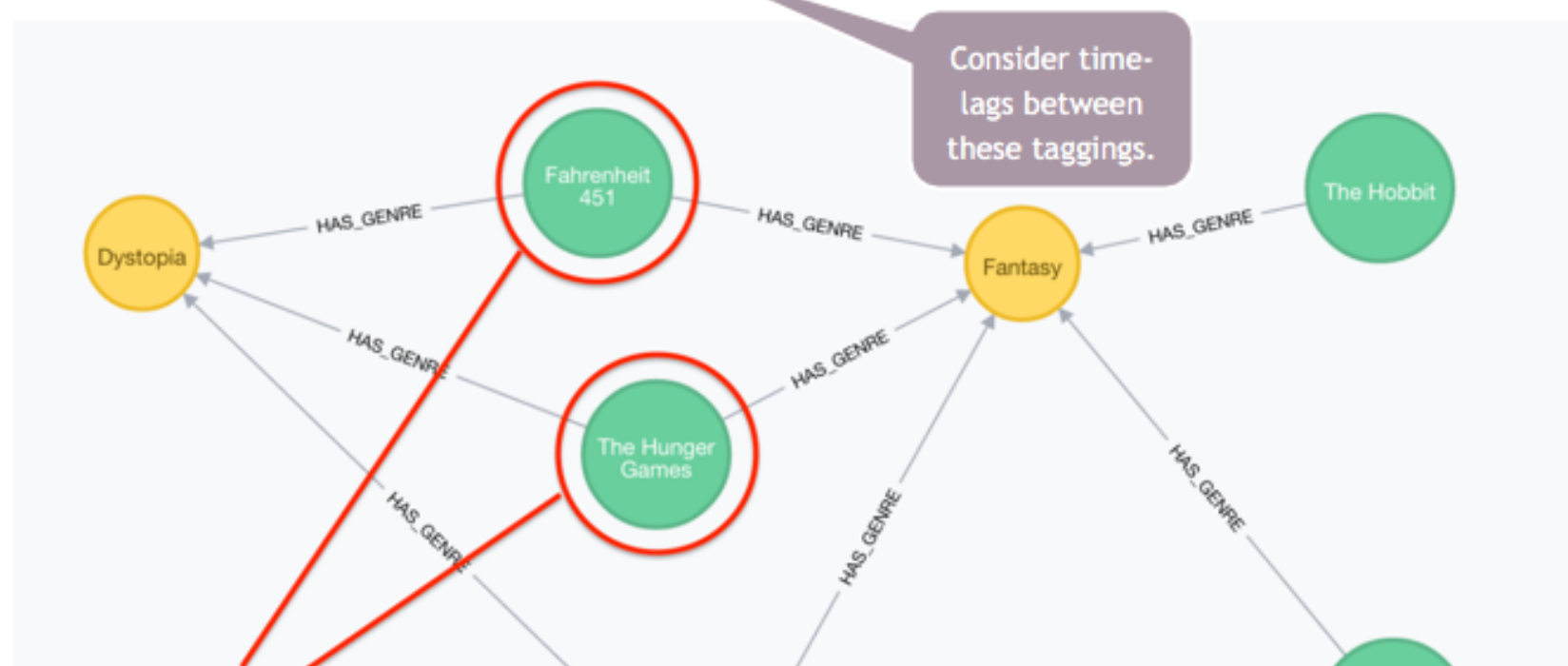
[4, 16, 36, 64, 100]

STEP1: Compute co-occurrence

Co-occurrence is the basic building block for the algorithm and is in itself a quite useful relationship because it indicates some degree of overlap between tags and therefore a certain degree of similarity which is something that can be exploited for query expansion or recommendation.

The co-occurrence index between two categories A and B is computed as the portion of items in category A that are also in category B, this is a simple division of the number of items tagged as both A and B divided by the number of items in tagged as A.

$$\text{COOC}(A,B) = \# \text{items tagged as both A and B} / \# \text{items tagged as A}$$



what is connected data?

It's almost as if we didn't used to know even how to collect data, while preserving its full value. We collected data for a specific set of reports, and when all reports were made, we weren't able to track details across those reports that came from the same data.

It cost money to store and sift through data. Like an old shirt. If you haven't worn it in the past year, get rid of it.

Now, we hold the data, and if it's not stored connected, at least we are capable of reconnecting it.

Questions