SCC.NRG.AI4ME: Self-optimising distributed encoding nodes.

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#### Acknowledgements

Supervisors: Dr Haris Rotsos, Prof Nick Race

#### Overview of Today's Presentation

#### Object Based Media

Traditiona

Dynamic Objects

High Level Overview Worked Example

Toy Example

Node Distribution

#### Testing the hypotheses

Generation of Objects/Cross-correlation
Optimisation of Objects

Network Distribution

Encoding/Decoding Weights Objects as Code (OaC)

Any Questions

References/Inspiration





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$$A = \{S_0^A \cdots S_n^A\}, B = \{S_0^B \cdots S_m^B\}$$

where for a given object  $S_x$  assume.

$$S_x \in A, S_x \in B$$

$$A = \{S_x | \Sigma_1\}, B = \{S_x | \Sigma_2\}$$

such that.

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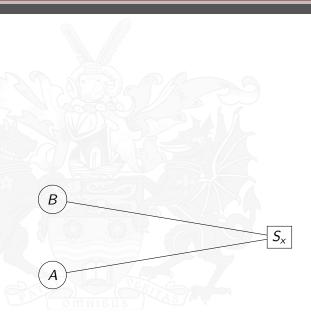
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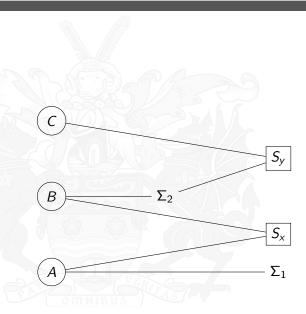
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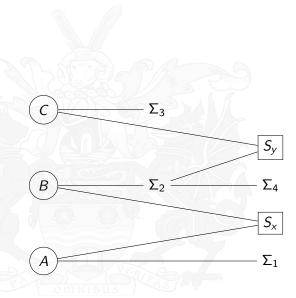
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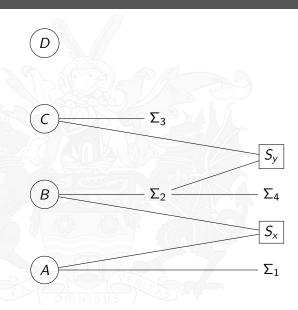
$$B = \Sigma_2 = \{S_y | \Sigma_3\}, C = \{S_y | \Sigma_4\}$$

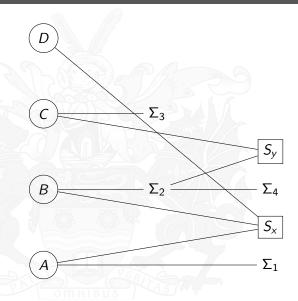
Figuring out if  $S_v$  is not in A or  $S_x$ ? No Idea.

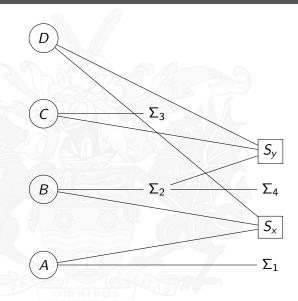


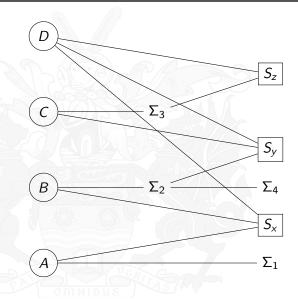




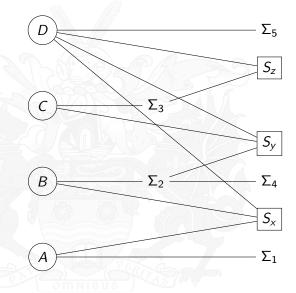


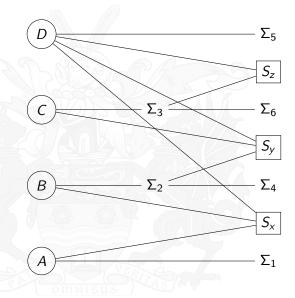


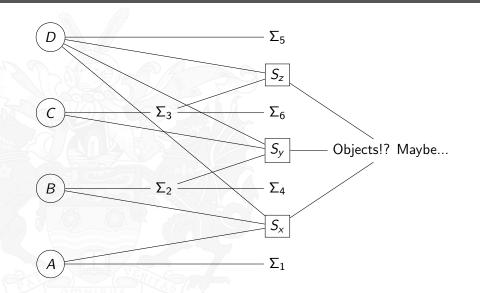


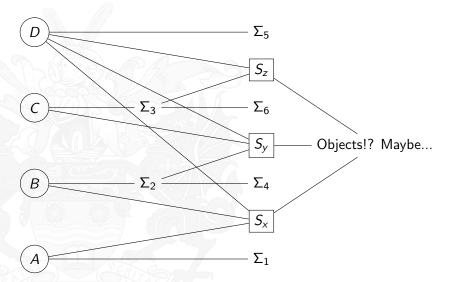












How do you optimise the ordering for encoder objects? No idea.



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Α	$\Sigma_1, S_{x}$
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assume that if the set is an itemised set of objects.

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Discard  $\Sigma_n$  as remainder Over large number of nodes.

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How to Assess if  $\Sigma_n$  is just artefact or truly unique? No idea.



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Add a new Client E. Existing parts of E that are already cached closer to the client fetch faster / optimisation to identify similar signals from a lower-quality/partially computed version.

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But how to identify  $S_z$  in E without full render? No idea.

Assuming that both signal A and B are correlated if they contain the same object with a different single change.

- Cross-Correlation?
- ► Steerable Pyramids?

Current Idea...

# Optimisation of an Object Tree How do you optimise extraction of signals?









