

### CAUSAL INTERPRETATION RULES FOR ENCODING AND DECODING MODELS IN NEUROIMAGING

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## Motivation



We tested [...] whether pre-stimulus alpha oscillations measured with electroencephalography (EEG) **influence** the encoding of items into working memory.

(Myers et al., Journal of Neuroscience, 2014)



Hippocampal activity in this study was correlated with amygdala activity, supporting the view that the amygdala enhances explicit memory by modulating activity in the hippocampus.

(S. Hamann, Trends in Cognitive Sciences, 2001)

- 1. Motivation
- 2. Approach
- 3. Encoding and decoding models in neuroimaging
- 4. Causal Bayesian Networks
- 5. Causal interpretation of encoding and decoding models
- 6. Empirical example
- 7. Wrap-up

### Approach









That sounds intriguing! So, what do you do?





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I present stimuli to subjects or observe their behaviour while recording their brain activity. ...[explains common analysis methods]...





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 $Ha,\ interesting!\ ...[rephrases\ what\ has\ been\ said\ in\ causal\ inference\ slang]...$ 





That sounds intriguing! So, what do you do?

I present stimuli to subjects or observe their behaviour while recording their brain activity. ...[explains common analysis methods]...

 $Ha,\ interesting!\ ...[rephrases\ what\ has\ been\ said\ in\ causal\ inference\ slang]...$ 



Yeah, a solid basis for our interpretations! It also clarifies problems that we recently discussed in the community.



Encoding and decoding models in neuroimaging

Trial 3

Trial 4

Trial 5

Trial 6

Trial 6

Trial 7

R

R

6

encoding



e.g. mean difference between conditions

decoding



e.g. classifier for experimental conditions





"Significant variation across experimental conditions?"

"Does removal impair decoding performance?"







"Significant variation across experimental conditions?"

"Does removal impair decoding performance?"





relevant feature 

✓ cognitive process

- $X \rightarrow Y \iff p(Y) \neq p(Y|do\{X = x\})$
- ▶ Causal Markov Condition: d-separation → independence
- ► Faithfulness: d-separation ← independence



- $X \rightarrow Y \iff p(Y) \neq p(Y|do\{X = x\})$
- ▶ Causal Markov Condition: d-separation → independence
- ► Faithfulness: d-separation ~ independence

$$\begin{array}{ccccc} \text{chain} & \text{fork} & \text{collider} \\ X \to Y \to Z & X \leftarrow Y \to Z & X \to Y \leftarrow Z \\ X \not \perp Z & \end{array}$$



- $X \rightarrow Y \iff p(Y) \neq p(Y|do\{X = x\})$
- ▶ Causal Markov Condition: d-separation → independence
- ► Faithfulness: d-separation ← independence

chain fork collider 
$$X \to Y \to Z$$
  $X \leftarrow Y \to Z$   $X \to Y \leftarrow Z$   $X \downarrow Z$   $X \perp Z \mid Y$ 



- $X \rightarrow Y \iff p(Y) \neq p(Y|do\{X = x\})$
- ► Causal Markov Condition: d-separation → independence
- ► Faithfulness: d-separation ← independence

chain fork collider 
$$X \to Y \to Z$$
  $X \leftarrow Y \to Z$   $X \to Y \leftarrow Z$   $X \not\perp Z$   $X \not\perp Z \mid X \mid Z \mid Y$ 



- $X \rightarrow Y \iff p(Y) \neq p(Y|do\{X = x\})$
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chain	fork	collider	
$X \to Y \to Z$	$X \leftarrow Y \rightarrow Z$	$X \to Y \leftarrow Z$	
$X \not\perp Z$	$X \not\perp Z$	$X \perp \!\!\! \perp Z$	
$X \perp Z   Y$	$X \perp Z Y$	$X \not\perp Z Y$	



### Causal Bayesian Networks in neuroimaging (A)



We tested [...] whether pre-stimulus alpha oscillations measured with electroencephalography (EEG) **influence** the encoding of items into working memory.

(Myers et al., Journal of Neuroscience, 2014)





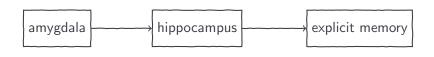
We tested [...] whether pre-stimulus alpha oscillations measured with electroencephalography (EEG) **influence** the encoding of items into working memory.

(Myers et al., Journal of Neuroscience, 2014)

$$\alpha \not\perp \mathsf{WM}$$



### Causal Bayesian Networks in neuroimaging (B)

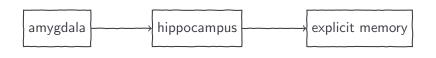


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### Causal Bayesian Networks in neuroimaging (B)



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AM ≠ EM AM ± EM | HC



# Causal interpretation of encoding and decoding models

Let's set out the causal component of already performed analyses..

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stimulus- vs response-based

feature relevance ↔ marginal/conditional dependence

→ 16 causal interpretation rules

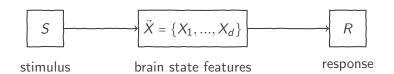


Let's set out the causal component of already performed analyses..

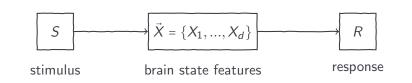
stimulus- vs response-based

feature relevance ↔ marginal/conditional dependence

→ 16 causal interpretation rules simple

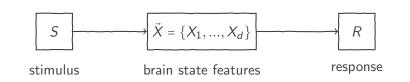






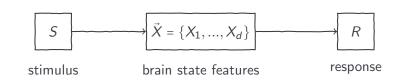
stimulus-base	ed	response-based
$p(\vec{X} S)$	encoding	$p(\vec{X} R)$
$p(S \vec{X})$	decoding	$p(R \vec{X})$





	stimulus-based		response-based	
$p(\vec{X} S)$	causal	encoding		$p(\vec{X} R)$
$p(S \vec{X})$		decoding	causal	$p(R \vec{X})$





	stimulus-based		response-based	
$p(\vec{X} S)$	causal	encoding	<i>anti-</i> causal	$p(\vec{X} R)$
$p(S \vec{X})$	<i>anti-</i> causal	decoding	causal	$p(R \vec{X})$



#### Relevance and dependence









$$p(X_i|C = c_1) \stackrel{?}{\neq} p(X_i|C = c_2)$$







$$p(X_i|C=c_1) \stackrel{?}{\neq} p(X_i|C=c_2) \qquad X_i \not\perp C$$









$$p(X_i|C = c_1) \stackrel{?}{\neq} p(X_i|C = c_2)$$

$$X_i \not\perp C$$

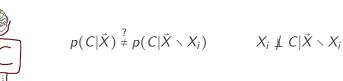


$$p(C|\vec{X}) \stackrel{?}{\neq} p(C|\vec{X} \setminus X_i)$$





$$p(X_i|C=c_1) \stackrel{?}{\neq} p(X_i|C=c_2)$$



$$X_i \not\perp C | \vec{X} \setminus X_i$$

 $X_i \not\perp C$ 





	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
pes	×		
Stimulus-based	<b>√</b>		
mulu		×	
Sti		$\checkmark$	
sed	×		
Response-based			
pons		×	
Res		$\checkmark$	



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
pes	×		no effect of $S$
Stimulus-based	$\checkmark$		
mulu		×	
Stii		$\checkmark$	
sed	×		
Response-based	$\checkmark$		
pons		×	
Res		$\checkmark$	



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based			effect of $S$
mulu		×	
Stii		$\checkmark$	
sed	×		
Response-based			
pons		×	
Res		<b>√</b>	



	$X_i \perp S   \vec{X} \setminus$	$X_i$
	$S \to X_j \to$	$X_i$
	5 X	i

Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based			effect of $S$
mulu		×	
Stii		$\checkmark$	
sed	×		
Response-based			
pons		×	
Res		<b>√</b>	



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of <i>S</i>
mulu		×	inconclusive
Stii		$\checkmark$	
sed	×		
Response-based	$\checkmark$		
pons		×	
Res			



	$X_i \not\perp S   \vec{X} \setminus$	X <sub>i</sub> inconclusive
	$S \to X_i$	
	$S \to X_j \leftarrow$	$X_i$



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of <i>S</i>
mulu		×	inconclusive
Stii		$\checkmark$	
sed	×		
Response-based	$\checkmark$		
pons		×	
Res			



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of $S$
mulu		×	inconclusive
Stii		$\checkmark$	inconclusive
sed	×		
Response-based	$\checkmark$		
pons		×	
Res		$\checkmark$	



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of S
mulu		×	inconclusive
Sti		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		
pons		×	
Res		$\checkmark$	



	X <sub>i</sub> ↓ R	
	$X_i \leftarrow h \rightarrow X_i \rightarrow R$	R nconclusive inconclusive
	example ( $\alpha \rightarrow WN$	A)



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of S
mulu		×	inconclusive
Sti		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		
pons		×	
Res		$\checkmark$	



Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
s-ba	$\checkmark$		effect of <i>S</i>
Stimulus-based		×	inconclusive
Stii		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		inconclusive
pons		×	
Res		$\checkmark$	



	$X_i \perp R   \vec{X} \setminus$	X <sub>inconclusive</sub>
	$X_i \to X_j \to$	Rnconclusive
	$X_i$ F	R no cause of $R$

Causal interpretation rules (1)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
s-ba	$\checkmark$		effect of <i>S</i>
Stimulus-based		×	inconclusive
Stii		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		inconclusive
pons		×	
Res		$\checkmark$	



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of <i>S</i>
mulu		×	inconclusive
Stii		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		inconclusive
bons		×	inconclusive
Res		$\checkmark$	



	$X_i \not\perp R   \vec{X} \setminus$	. X <sub>i</sub>
	$X_i \to R$	
	$X_i \leftarrow h \rightarrow \downarrow$	R no cause of R
	$X_j$	



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of <i>S</i>
mulu		×	inconclusive
Stii		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		inconclusive
bons		×	inconclusive
Res		$\checkmark$	



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×		no effect of $S$
Stimulus-based	$\checkmark$		effect of <i>S</i>
mulu		×	inconclusive
Sti		$\checkmark$	inconclusive
sed	×		no cause of <i>R</i>
Response-based	$\checkmark$		inconclusive
pons		×	inconclusive
Res		$\checkmark$	inconclusive



Are dec	oding models	useful at all?
Are dec	oding models	useful at all?
Are dec	oding models	useful at all?
Are dec	oding models	



Causal interpretation rules (2)

	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×	×	
Stimulus-based	$\checkmark$	×	
mulu	×		
Stil	$\checkmark$	<b>√</b>	
sed	×	×	
Response-based		×	
bons	×		
Res	$\checkmark$		



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
peg	×	×	
s-ba	$\checkmark$	×	
Stimulus-based	×	$\checkmark$	
Stii	$\checkmark$	$\checkmark$	
sed	×	×	
Response-based	$\checkmark$	×	
	×	$\checkmark$	
Res	$\checkmark$	$\checkmark$	



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
sed	×	×	no effect of $S$
Stimulus-based	$\checkmark$	×	
mulu	×	<b>√</b>	provides context
Stii	$\checkmark$	$\checkmark$	effect of S
sed	×	×	no cause of R
Response-based	$\checkmark$	×	
	×	$\checkmark$	provides context
Res	$\checkmark$		inconclusive



V J	×	CIV V	
× X <sub>i</sub> ↓	$S$ and $X_i$ $\blacksquare$	$LS X \setminus X_i$	
	$S \rightarrow X_i$ indi	rectly	
	e.g. $S \to X_i$	$\rightarrow X_i$	
	×		

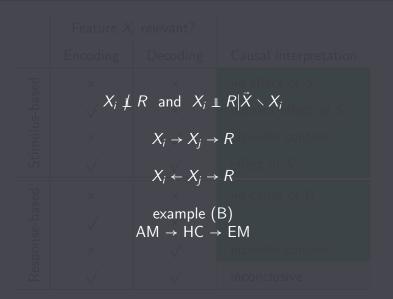


	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
Stimulus-based	×	×	no effect of $S$
	$\checkmark$	×	
nulu	×	$\checkmark$	provides context
Stir	$\checkmark$	$\checkmark$	effect of S
Response-based	×	×	no cause of R
	$\checkmark$	×	
	×	$\checkmark$	provides context
Res	$\checkmark$	$\checkmark$	inconclusive



	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
Stimulus-based	×	×	no effect of $S$
	$\checkmark$	×	indirect effect of $S$
	×	$\checkmark$	provides context
Stii	$\checkmark$	$\checkmark$	effect of S
Response-based	×	×	no cause of <i>R</i>
	$\checkmark$	×	
	×	$\checkmark$	provides context
Res	$\checkmark$	$\checkmark$	inconclusive





С

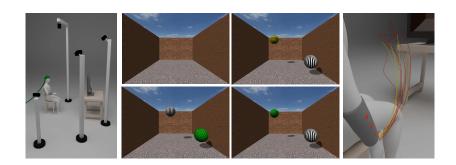
	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
Stimulus-based	×	×	no effect of $S$
	$\checkmark$	×	indirect effect of $S$
	×	$\checkmark$	provides context
Stii	$\checkmark$	$\checkmark$	effect of S
Response-based	×	×	no cause of <i>R</i>
	$\checkmark$	×	
	×	$\checkmark$	provides context
Res	$\checkmark$	$\checkmark$	inconclusive



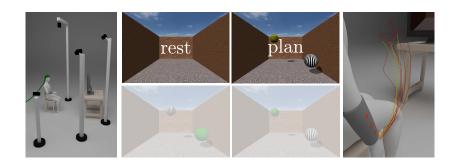
	Feature $X_i$ relevant?		
	Encoding	Decoding	Causal interpretation
peg	×	×	no effect of $S$
s-ba	$\checkmark$	×	indirect effect of $S$
Stimulus-based	×	$\checkmark$	provides context
Stii	$\checkmark$	$\checkmark$	effect of S
Response-based	×	×	no cause of R
	$\checkmark$	×	no direct cause of R
	×	$\checkmark$	provides context
Res	$\checkmark$	$\checkmark$	inconclusive



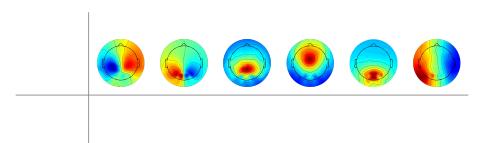
# Empirical example



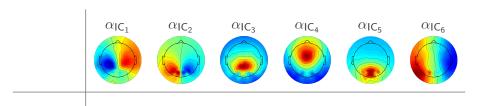




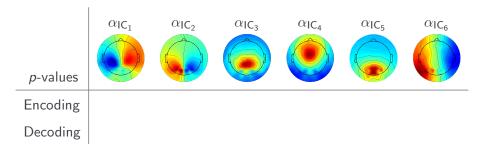














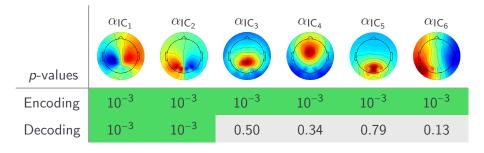
<i>p</i> -values	$\alpha_{IC_1}$	$\alpha_{IC_2}$	$\alpha_{IC_3}$	$\alpha_{IC_4}$	$\alpha_{IC_5}$	$\alpha_{IC_6}$
Encoding	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$
Decoding	$10^{-3}$	$10^{-3}$	0.50	0.34	0.79	0.13



<i>p</i> -values	$\alpha_{IC_1}$	$\alpha_{IC_2}$	$\alpha_{IC_3}$	$\alpha_{IC_4}$	$\alpha_{IC_5}$	$\alpha_{IC_6}$
Encoding	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$
Decoding	$10^{-3}$	$10^{-3}$	0.50	0.34	0.79	0.13



## Causal analysis







#### Causal analysis

<i>p</i> -values	$\alpha_{IC_1}$	$\alpha_{IC_2}$	$\alpha_{IC_3}$	$\alpha_{IC_4}$	$\alpha_{IC_5}$	$\alpha_{IC_6}$
Encoding	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$
Decoding	$10^{-3}$	$10^{-3}$	0.50	0.34	0.79	0.13

- instruction to plan a reaching movement is causal for all  $\alpha_{{\rm IC}_i}$
- $\alpha_{\rm IC_3},...,\alpha_{\rm IC_6}$  are only indirect effects





<i>p</i> -values	$\alpha_{IC_1}$	$\alpha_{IC_2}$	$\alpha_{IC_3}$	$\alpha_{IC_4}$	$\alpha_{IC_5}$	$\alpha_{IC_6}$
Encoding	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$	$10^{-3}$
Decoding	$10^{-3}$	$10^{-3}$	0.50	0.34	0.79	0.13

- instruction to plan a reaching movement is causal for all  $\alpha_{{\rm IC}_i}$
- $\alpha_{\rm IC_3},...,\alpha_{\rm IC_6}$  are only indirect effects





Wrap-up





feature relevance





feature relevance  $\mathcal{L}^{7}$  (conditional) (in)dependence





feature relevance  $\mathcal{L}^{7}$  (conditional) (in)dependence  $\mathcal{L}^{7}$ 





feature relevance  $\mathcal{L}^{7}$  (conditional) (in)dependence  $\mathcal{L}^{7}$ 

- simple interpretation rules
- reinterpretation of previous results?
- resolve recently discussed issues





feature relevance  $\mathcal{L}^{7}$  (conditional) (in)dependence  $\mathcal{L}^{7}$ 

- simple interpretation rules
- reinterpretation of previous results?
- resolve recently discussed issues

It's an interesting application!





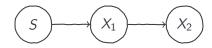
Sebastian Weichwald, Timm Meyer, Ozan Özdenizci, Bernhard Schölkopf, Tonio Ball, Moritz Grosse-Wentrup:

- Causal interpretation rules for encoding and decoding models in neuroimaging. NeuroImage, 2015.
- Causal and anti-causal learning in pattern recognition for neuroimaging. PRNI, 2014.





20



• Establish  $X_1 \to X_2$  even in the presence of latent confounders (Grosse-Wentrup et al., under revision)

• Identify  $X_2$  from a linear mixture of signals

Problem of confounds in MVPA

(Todd et al., Neurolmage, 2013; Woolgar et al., Neurolmage, 2014)

▶ Interpretation of weight vectors of linear models

(Haufe et al., NeuroImage, 2014)

▶ Type II errors

- Conditional independence tests
  - permutation-based → biased
     (Strobl et al., BMC Bioinformatics, 2008)
  - o unbiased → hard

(Zhang et al., *UAI*, 2011)

Untestable assumption: faithfulness