

CAUSAL INTERPRETATION RULES FOR ENCODING AND DECODING MODELS IN NEUROIMAGING

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Motivation



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)



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)

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



encoding



e.g. mean difference between conditions

decoding



e.g. classifier for experimental conditions





"Feature shows significant variation across experimental conditions?"

"Feature helpful for predicting the experimental condition?"







"Feature shows significant variation across experimental conditions?"

"Feature helpful for predicting the experimental condition?"





relevant feature ⟨→⟩ cognitive process

Causal interpretation of encoding and decoding models

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



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

stimulus- vs response-based

feature relevance ↔ marginal/conditional dependence

→ 16 causal interpretation rules

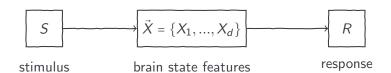


Let's set out the causal component of already performed analyses..

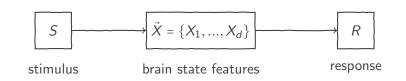
stimulus- vs response-based

feature relevance ↔ marginal/conditional dependence

→ 16 causal interpretation rules simple

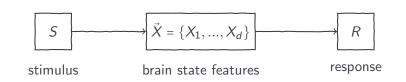






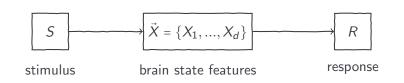
	stimulus-based		response-based	
$p(\vec{X} S)$		encoding		$p(\vec{X} R)$
$p(S \vec{X})$		decoding		$p(R \vec{X})$





	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









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







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







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





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





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





$$p(C|\vec{X}) \stackrel{?}{\neq} p(C|\vec{X} \setminus X_i)$$
 $X_i \not\perp C|\vec{X} \setminus X_i$





Causal interpretation rules (1)



Causal interpretation rules (1)

	Feature X_i relevant?		
	Encoding	Decoding	Causal interpretation
peg	×		
Stimulus-based	\checkmark		
mult		×	
Sti		\checkmark	
sed	×		
Response-based	√		
bons		×	
Res			



	$X_i \not\perp R$ $X_i \leftarrow h \rightarrow$	R
	$X_i \to R$	

Causal interpretation rules (1)

	Feature X_i relevant?		
	Encoding	Decoding	Causal interpretation
peg	×		
Stimulus-based	\checkmark		
mult		×	
Sti		\checkmark	
sed	×		
Response-based	√		
bons		×	
Res			



Causal interpretation rules (1)

	Feature X_i relevant?		
	Encoding	Decoding	Causal interpretation
pes	×		
Stimulus-based	\checkmark		
mulu		×	
Stir		√	
ed	×		
Response-based	\checkmark		inconclusive
bons		×	
Res		\checkmark	



	Feature X_i relevant?		
	Encoding	Decoding	Causal interpretation
peg	×		
Stimulus-based	\checkmark		
mulu		×	inconclusive
Stir		\checkmark	inconclusive
sed	×		
Response-based	\checkmark		inconclusive
bons		×	inconclusive
Res		\checkmark	inconclusive



	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
peg	×		no effect of S
Stimulus-based	\checkmark		effect of <i>S</i>
mulu		×	inconclusive
Stir		\checkmark	inconclusive
pes	×		no cause of <i>R</i>
Response-based	\checkmark		inconclusive
pons		×	inconclusive
Res		\checkmark	inconclusive



Causal interpretation rules (2)

	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
sed	\checkmark	\checkmark	
Stimulus-based	\checkmark	×	
mulu	×	\checkmark	
Stil	×	×	
sed	√	√	
Response-based		×	
pons	×	\checkmark	
Res	×	×	



	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
sed	\checkmark	\checkmark	
s-ba	\checkmark	×	
Stimulus-based	×	\checkmark	
Stii	×	×	
sed	\checkmark	\checkmark	inconclusive
Response-based	\checkmark	×	
pons	×	\checkmark	
Res	×	×	



× X _i ‡	S and X_i \bot	$ S X \setminus X_i$		
	$S \rightarrow X_i$ indi	rectly		
	√ √	inconclusive		

	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
sed	\checkmark	√	
s-ba	\checkmark	×	
Stimulus-based	×	\checkmark	
Stii	×	×	
sed	\checkmark	\checkmark	inconclusive
Response-based	\checkmark	×	
pons	×	\checkmark	
Res	×	×	



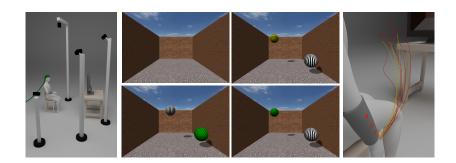
	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
sed	\checkmark	√	
Stimulus-based	\checkmark	×	indirect effect of S
mulu	×	\checkmark	
Sti	×	×	
sed	\checkmark	√	inconclusive
e-ba	\checkmark	×	
Response-based	×	\checkmark	
Res	×	×	



	Feature X	; relevant?	
	Encoding	Decoding	Causal interpretation
ed	\checkmark	\checkmark	effect of S
Stimulus-based	\checkmark	×	indirect effect of S
nulu	×	\checkmark	provides context
Stir	×	×	no effect of S
pes	√	√	inconclusive
e-ba	\checkmark	×	no direct cause of R
Response-based	×	\checkmark	provides context
Res	×	×	no cause of R



Empirical example

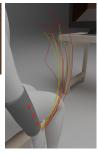




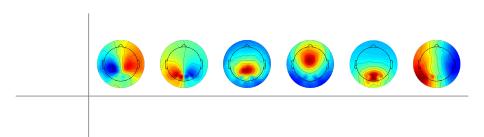




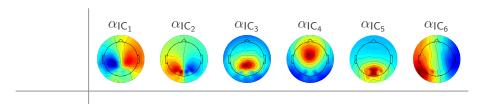




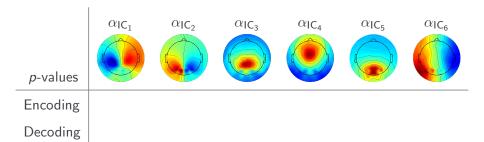














<i>p</i> -values	α_{IC_1}	α_{IC_2}	α_{IC_3}	α _{IC4}	α _{IC5}	α_{IC_6}
Encoding	0	0	0	0	0	0
Decoding	0	0	0.50	0.34	0.79	0.13



<i>p</i> -values	α_{IC_1}	α_{IC_2}	α_{IC_3}	α _{IC4}	α_{IC_5}	α_{IC_6}
Encoding	0	0	0	0	0	0
Decoding	0	0	0.50	0.34	0.79	0.13



Causal analysis

<i>p</i> -values	α_{IC_1}	α_{IC_2}	α_{IC_3}	α _{IC4}	α_{IC_5}	α_{IC_6}
Encoding	0	0	0	0	0	0
Decoding	0	0	0.50	0.34	0.79	0.13





Causal analysis

<i>p</i> -values	α_{IC_1}	α_{IC_2}	α_{IC_3}	α_{IC_4}	α_{IC_5}	α_{IC_6}
Encoding	0	0	0	0	0	0
Decoding	0	0	0.50	0.34	0.79	0.13

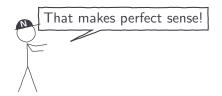
- ullet instruction to plan a reaching movement is causal for all $lpha_{{
 m IC}_i}$
- $\alpha_{{\rm IC_3}},...,\alpha_{{\rm IC_6}}$ are only indirect effects





<i>p</i> -values	α_{IC_1}	α_{IC_2}	α_{IC_3}	α_{IC_4}	α_{IC_5}	α_{IC_6}
Encoding	0	0	0	0	0	0
Decoding	0	0	0.50	0.34	0.79	0.13

- instruction to plan a reaching movement is causal for all $\alpha_{{\rm IC}_i}$
- ${}^{\blacktriangleright}$ $\alpha_{\rm IC_3},...,\alpha_{\rm IC_6}$ 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.



sweichwald.de/neuroimage2015

