Author(s)	Type	Stimulus	Decoding	Notes
Cantürk and Volosyak	proceedings	m-sequence	CCA	Language model (ChatGPT)
Cabrera Castillos and Dehais	dataset	burst codes		Grating stimuli
Dehais, Cabrera Castillos,	preprint	burst codes	Riemannian, logistic regression	Grating stimuli, dry EEG, comfort, eye-
Ladouce, and Clisson				strain
Fodor, Herschel, Cantürk,	article	m-sequence	CCA	Classification certainty feedback
Heisenberg, and Volosyak				
E. Lai, Mai, Ji, Li, and Meng	proceedings	DIBS	filterbank task related compo-	Asynchronous
			nent analysis (FBTRCA), LSTM	
Martínez-Cagigal, Ál-	proceedings	non-binary m-sequence	CCA	Learning curve
varo Fernández-Rodríguez,				
Santamaría-Vázquez, Martín-				
Fernández, and Hornero	1	1.0	TTDCA 1: 1 1: (No. 1 19 00 11 00
Miao et al.	article	white noise	TDCA, linear modeling, transfer	Minimal calibration, subject-to-subject
Name and Alexa di Dessia		modulated Gold codes	learning	transfer
Narayanan, Ahmadi, Desain, and Thielen	proceedings	modulated Gold codes	CCA	Gaze-independent, covert attention
Scheppink, Ahmadi, Desain,	proceedings	modulated Gold codes	CCA	Auditory, c-AEP
Tangermann, and Thielen	proceedings	modulated Gold codes	CCA	Auditory, C-AET
Shi et al.	article	white-noise	TDCA	Maximum information rate
Sun et al.	article	m-sequence	TDCA	Small stimuli (0.5, 1, 2, 3 visual degrees)
Thielen, Sosulski, and Tanger-	proceeding	modulated Gold codes	reconvolution, CCA, UMM	Calibration-free
mann	proceeding	modulated Gold codes	reconvolution, early entity	Campration nee
Thielen, Farquhar, and Desain	dataset	modulated Gold codes		
Velut, Chevallier, Corsi, and De-	proceedings	burst codes	CNN, SPDNet, transfer learning	Subject-to-subject transfer
hais	1 0		, ,	,
Zheng, Dong, et al.	article	NRBS	FBCCA	Calibration-free, c-VEP versus SSVEP
Zheng, Tian, et al.	dataset	NRBS		c-VEP versus SSVEP

Author(s)	Type	Stimulus	Decoding	Notes
Ahmadi and Desain	preprint	modulated Gold codes	CCA	Bayesian dynamic stopping
Castillos, Ladouce, Darmet, and	article	m-sequence, burst codes	CNN	
Dehais		-		
Castillos	dataset	m-sequence, burst codes		
Darmet, Ladouce, and Dehais	proceedings	m-sequence	TRCA, EEG2Code, CNN	
Fernández-Rodríguez,	article	m-sequence	CCA	Eyestrain spatial frequency
Martínez-Cagigal, Santamaría-		-		
Vázquez, Ron-Angevin, and				
Hornero				
Henke et al.	proceedings	m-sequence	CCA	Background music
C. Huang et al.	preprint	white noise	TRCA	Visual tracking
E. Lai, Mai, and Meng	proceedings	DIBS	FBTRCA, LSTM	fatigue
Martínez-Cagigal et al.	article	m-sequence, non-binary m-	CCA	Eyestrain, fatigue
		sequences		
Martínez-Cagigal, Santamaría-	proceedings	non-binary m-sequence	CCA	Dynamic stopping
Vázquez, and Hornero				
Moreno-Calderón et al.	article	m-sequence	CCA	Games
Santamaría-Vázquez, Martínez-	proceedings	non-binary m-sequence,	EEG-inception	
Cagigal, and Hornero				
Thielen	proceedings	m-sequence, APA sequence,	reconvolution, CCA	Simulated EEG
		Gold codes, Golay sequence, de		
		Bruijn sequence		
Thielen, Cornielje, van der	proceedings	m-sequence, Gold codes, Golay	reconvolution, CCA	Empirical EEG
Werff, and Desain		sequence, de Bruijn sequence,		
		modulated codes		
Thielen, Marsman, Farquhar,	dataset	modulated Gold codes		
and Desain				
Volosyak et al.	proceedings	m-sequence	CCA	Gender
Wolf and Götzelmann	article			VEPdgets, Dry EEG
Xu et al.	article	m-sequence	TRCA	c-VEP versus SSVEP

Author(s)	Type	Stimulus	Decoding	Notes
Dehais et al.	article			Dry EEG, flight simulator, active and passive BCI
Stawicki and Volosyak	article	m-sequence	transfer learning	Session-to-session transfer
Sun, Zheng, Pei, Gao, and Wang	article	shifted Gold code	FBTRCA	120 targets
Ying, Wei, and Zhou	article	m-sequence	Riemannian, transfer learning	Subject-to-subject transfer
Zarei and Asl	article	m-sequence	spatiotemporal beamformer	
Zarei and Asl	article	m-sequence	spatiotemporal beamformer	Improved covariance estimator
Zheng, Pei, Gao, Zhang, and Wang	article	Gold codes	TRCA	Brain-switch

2021

Author(s)	Type	Stimulus	Decoding	Notes
Kaya, Bohorquez, and Özdamar Martínez-Cagigal et al.	article article	quasi steady-state	CLAD	Review c-VEP
Thielen, Marsman, Farquhar, and Desain	article	modulated Gold codes	reconvolution, CCA	Zero-training
Torres and Daly	article	APA sequence, de Bruijn sequence, Golay sequence, m-sequence, Gold code, Kasami sequence	CCA, ICA, PCA, MLP	Synthetic EEG
Verbaarschot et al.	article	modulated Gold codes	CCA	ALS versus healthy participants

Author(s)	Type	Stimulus	Decoding	Notes
Behboodi, Mahnam, Marateb, and Rabbani	article	m-sequence, TFO, 6FO	CCA	
Gembler, Rezeika, Benda, and Volosyak	article	m-sequence, quintary m- sequence	FBCCA	Presentation rate (60, 120, 240), comfort
Gembler, Benda, Rezeika, Stawicki, and Volosyak	article	m-sequence	CCA	Asynchronous, language model
Gembler	dissertation			c-VEP
Gembler, Stawicki, Rezeika,	proceedings	m-sequence	FBCCA	Asynchronous, multi-session
Benda, and Volosyak		•		•
Z. Huang, Zheng, Wu, and	article	m-sequence	transfer-learning	Subject-to-subject transfer
Wang		•	<u> </u>	,
Volosyak, Rezeika, Benda, Gem-	article	m-sequence	CCA	SSVEP, SSMVEP, c-VEP, BCI illiteracy
bler, and Stawicki		-		•
Shirzhiyan et al.	article	periodic, quasi-periodic, chaotic codes	CCA	Fatigue
Turi, Gayraud, and Clerc	article	m-sequence	CCA	Auto-calibration, language model, zero training
Yasinzai and Ider	article	m-sequence, random sequence, SOP sequences	CCA	

Author(s)	Type	Stimulus	Decoding	Notes
Ahmadi	dataset	modulated Gold codes		
Ahmadi	dataset	modulated Gold codes		
Ahmadi, Borhanazad, Tump,	article	modulated Gold codes	CCA	Low channel count
Farquhar, and Desain				
Başaklar, Tuncel, and İder	article	m-sequence	CCA	Presentation rate (60, 120, 240 Hz)
Borhanazad, Thielen, Farquhar, and Desain	proceedings	modulated Gold codes	CCA	Presentation rate (40, 60, 90, 120 Hz)
Desain, Thielen, van den Broek, and Farquhar	patent	modulated Gold codes	CCA	
Gembler and Volosyak	article	m-sequence	CCA	Language model
Gembler, Stawicki, Rezeika, and Volosyak	proceedings	m-sequence	FBCCA	Presentation rate (30, 60, 120 Hz), age (young, elderly)
Gembler, Stawicki, Saboor, and Volosyak	article	m-sequence	FBCCA	Language model, dynamic stopping
Gembler, Benda, Saboor, and Volosyak	proceedings	m-sequence	FBCCA	Language model, dynamic stopping
Grigoryan, Filatov, and Kaplan	article	m-sequence	CCA	Presentation rate (30, 60, 120 Hz)
Kadıoğlu, Yıldız, Closas, Fried-	article	m-sequence	Maximum likelihood	Color (green-red), fusion of c-VEP and eye
Oken, and Erdoğmuş		1		tracker
Kaya, Bohorquez, and Ozdamar	proceedings	quasi steady-state	CLAD	OSSVEP
Kaya, Bohórquez, and Özdamar	article	quasi steady-state	CLAD	OSSVEP
Kaya	dissertation	ı y		QSSVEP
Luo and Huang	proceedings	m-sequence	LDA, transfer learning	Subject-to-subject transfer
Matsuno, Itakura, Mizuno, and	proceedings	1		frequency-hopping VEP
Mito	1			7 11 0
Nagel and Spüler	article	optimized random sequences	EEG2Code	Asynchronous, non-control state
Nagel and Spüler	article	random sequences	EEG2Code	
Nagel	dissertation	•		c-VEP
Peng and Huang	proceedings	m-sequence	sLDA	For psychological experiments (button presses without behavior)
Shirzhiyan et al.	article	m-sequence, chaotic codes	CCA, spatiotemporal bear forming	± '
Turi and Clerc	article	m-sequence		Static stopping number of cycles
Zheng, Wang, Pei, and Chen	proceedings	Gold codes	TRCA	Brain switch

Author(s)	Type	Stimulus	Decoding	Notes
Başaklar, İder, and Tuncel	proceedings	m-sequence		Presentation rate (60, 120, 240 Hz)
Dimitriadis and Marimpis	article	m-sequence	SVM	PAC, healthy and patients
Gembler, Stawicki, Saboor, et al.	proceedings	m-sequence	CCA	Presentation rate (60, 120, 200 Hz)
Gembler, Stawicki, Rezeika, et al.	proceedings	m-sequence	CCA	Language model
Liu, Wei, and Lu	article	Golay sequence, APA sequence	CCA	
Nagel, Dreher, Rosenstiel, and Spüler	article	m-sequence		Monitor raster latency, P300, SSVEP, c-VEP
Nagel, Rosenstiel, and Spüler	proceedings	optimized random sequences	CCA, regression	
Nagel and Spüler	article	random and optimized sequences	Ridge regression, EEG2Code	
Nezamfar, Mohseni Salehi, Hig- ger, and Erdogmus	article	m-sequence	RDA	Color (green-red), c-VEP versus eye tracker
Spüler and Kurek	article	m-sequence	CCA, SVM	ASSR versus c-AEP
Turi, Gayraud, and Clerc	preprint	m-sequence		Zero-training, language model
Wei et al.	article	grouping modulation, Golay complementary sequences, APA sequence	CCA	

2017

2017				
Author(s)	Type	Stimulus	Decoding	Notes
Aminaka and Rutkowski Isaksen, Mohebbi, and Puthusserypady	chapter article	m-sequence m-sequence, Gold code, Barker code	CCA, SVM correlation	Color (green-blue), 40 Hz
Nagel, Rosenstiel, and Spüler	proceedings	m-sequence, random codes	CCA	
Spüler	article	m-sequence	CCA	Dry EEG, static and dynamic stopping
Thielen, Marsman, Farquhar, and Desain	chapter	modulated Gold codes	reconvolution, CCA	Zero-training
Wei, Gong, and Lu	article	grouping modulation, Golay sequence, APA sequence	CCA	
Wittevrongel, Van Wolputte, and Van Hulle	article	m-sequence	beamformer	

Author(s)	Type	Stimulus	Decoding	Notes
Desain, Thielen, van den Broek, and Farquhar	patent	modulated Gold codes	CCA	
Isaksen, Mohebbi, and Puthusserypady	proceedings	m-sequence	Barker code, Gold code	
Nezamfar, Salehi, Moghadamfalahi, and Er- dogmus	article	m-sequence		FlashType, color (red-green), 110 Hz, language model
Riechmann, Finke, and Ritter	article	m-sequence	SVM (linear)	Color (green-red, black-white), shape, background, 120 Hz, virtual agent
Sato and Washizawa	proceedings	m-sequence	CCA, MLP, Lasso regression, Linear regression	Ç Ç
Thielen, Farquhar, and Desain	proceedings	modulated Gold codes	reconvolution, CCA	
Wei, Feng, and Lu	article	m-sequence	CCA	Stimulus characteristics: size (0.67, 1.7, 2.8, 3.8, 5.4, 7.1, 8.9 dva), color (white, red, green, blue, yellow), proximity (3.8, 4.8, 5.8, 6.8 dva), length (15, 31, 63, 127 bits), lag (2, 4, 6, 8, 10 bits)
Wei, Huang, Li, and Lu	article	m-sequence, Golay sequence	CCA	

Author(s)			Type	Stimulus	Decoding	Notes
Aminaka, Rutkowski	Makino,	and	proceedings	m-sequence	CCA	Color (green-blue, white-black), presentation rate (30, 40 Hz)
Aminaka, Rutkowski	Makino,	and	proceedings	m-sequence	CCA, SVM	Color (green-blue, white-black), presentation rate (30, 40 Hz), CCA versus SVM
Aminaka, Rutkowski	Makino,	and	proceedings	m-sequence	SVM	Color (green-blue, white-black), presentation rate (30, 40 Hz), pass-band optimization (6-21 Hz)
Aminaka, Rutkowski	Makino,	and	proceedings	m-sequence	SVM	Color (green-blue, white-black), presentation rate (30, 40 Hz), SVM (linear, polynomial, rbf, sigmoid)
Mohebbi et al	1.		proceedings	Gold code	correlation	Wheelchair
Nezamfar, S mus	Salehi, and I	Erdog-	proceedings	m-sequence	maximum likelihood	Color (red-green, blue-yellow, black-white), presentation rate (30, 60, 110 Hz)
Sato and Was	shizawa		proceedings	m-sequence	correlation	Automatic repeat request
Spüler			proceedings	m-sequence	CCA, SVM	Windows applications
Thielen, van quhar, and D		, Far-	article	modulated Gold codes	reconvolution, CCA	
Waytowich a	nd Krusiensk	i	article	m-sequence	CCA	Foveal versus peripheral stimulation

2014

Author(s)	Type	Stimulus	Decoding	Notes
Kapeller et al.	article	m-sequence	CCA, LDA	Invasive, video application
Tu et al.	article		CSP, SVM, Naive Bayes, LDA	Color (red-green), CTVEP

Author(s)	Type	Stimulus	Decoding	Notes
Bohórquez, Lozano, Kao, Toft- Nielsen, and Özdamar	proceedings	temporally jittered SSVEP	CLAD	
Kapeller et al.	proceedings	m-sequence	CCA, LDA	Robot, SSVEP versus c-VEP
Riechmann, Finke, and Ritter	proceedings	hierarchical codebook	SVM	Color (red-green, black-white)
Spüler, Rosenstiel, and Bogdan	proceedings	m-sequence	OCSVM,	Unsupervised online calibration
Spüler, Rosenstiel, and Bogdan	proceedings	m-sequence	OCSVM	Unsupervised online calibration
Spüler, Walter, Rosenstiel, and	article	m-sequence	CCA, OCSVM	c-VEP, ERN, P300, TMSEP, CCEP
Bogdan		-		

2012

Author(s)	Type	Stimulus	Decoding	Notes
Nakanishi and Mitsukura	proceedings	m-sequence, periodic codes	periodicity detection	Online unsupervised adaptation with ERN
Spüler, Rosenstiel, and Bogdan	proceedings	m-sequence	CCA, OCSVM	
Spüler, Rosenstiel, and Bogdan	article	m-sequence	CCA, OCSVM	

2011

Author(s)	Type	Stimulus	Decoding	Notes
Bin et al.	article	m-sequence	CCA	
S. M. Lai, Zhang, Hung, Niu, and Chang	article			Color (red-green), CTVEP
Nezamfar, Orhan, Purwar, et al.	article	m-sequence	template matching, Bayesian fusion	
Nezamfar, Orhan, Erdogmus, et al.	proceedings	m-sequence	correlation, naive Bayes	Presentation rate (15, 30 Hz)

2009

Author(s)	Туре	Stimulus	Decoding	Notes
Bin, Gao, Wang, Hong, and Gao	article	m-sequence	correlation	ERP versus SSVEP versus c-VEP

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Author(s)	Type	Stimulus	Decoding	Notes
Desain, Farquhar, Blankespoor, and Gielen	proceedings	Gold codes	reconvolution	Auditory
Farquhar, Blankespoor, Vlek, and Desain	proceedings	Gold codes		Auditory
Momose	proceedings	m-sequence		Hybrid P300 and c-VEP

Author(s)	Type	Stimulus	Decoding	Notes	
Momose	proceedings	m-sequence			

Author(s)	Type	Stimulus	Decoding	Notes	
Bohórquez and Özdamar	article	m-sequence	CLAD	Auditory	

Author(s)	Type	Stimulus	Decoding	Notes	
Hanagata and Momose	proceeding				

Author(s)	Type	Stimulus	Decoding	Notes
Sutter	article	m-sequence	correlation	Invasive, ALS patient

Author(s)	Type	Stimulus	Decoding	Notes	
Sutter	proceedings				

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Glossary

6FO 6 factor optimum. 4
ALS amyotrophic lateral sclerosis. 3, 10
APA almost perfect auto-correlation. 2, 3, 6
ASSR auditory steady-state response. 6
BCI brain computer interfacing. 4
c-AEP code-modulated auditory evoked potential. 1, 6
c-VEP code-modulated visual evoked potential. 1–6, 9, 10
CCA canonical correlation analysis. 1–9
CCEP contrico-cortical evoked potential. 9
CLAD continuous loop averaging deconvolution. 3, 5, 9, 10
CNN convolutional neural network. 1, 2
CSP common spatial patterns. 8
CTVEP chromatic transient visual evoked potential. 8, 9
DIBS discrete interval binary sequence. 1, 2
EEG electroencephalography. 1–3, 6
ERN error related negativity. 9
ERP event related potential. 9
FBCCA filterbank canonical correlation analysis. 1, 4, 5
FBTRCA filterbank task related component analysis. 1–3
ICA independent component analysis. 3
LDA linear discriminant analysis. 5, 8, 9

LSTM long short term memory. 1, 2 MLP multilayer perceptron. 3, 7 NRBS narrow-band random sequences. 1 OCSVM one class support vector machine. 9 PAC phase to amplitude coupling. 6 PCA principal component analysis. 3 QSSVEP quasi steady-state visual evoked potential. 5 RDA regularized discriminant analysis. 6 **sLDA** shrinkage linear discriminant analysis. 5 **SOP** superposition optimized pulse. 4 SSMVEP steady state motion visual evoked potential. 4 SSVEP steady state visual evoked potential. 1, 2, 4, 6, 9 **SVM** support vector machine. 6–9 TDCA task discriminative component analysis. 1 TFO time-factor optimum. 4 TMSEP transcranial magnetic stimulation evoked potential. 9 TRCA task related component analysis. 2, 3, 5 UMM unsupervised mean difference maximization. 1 **VEP** visual evoked potential. 5