Author(s)		Type	Stimulus	Decoding	Notes
Martínez-Cagigal,	Thielen,	article			Editorial c-VEP
Hornero, and Desain					

Author(s)	Type	Stimulus	Decoding	Notes
Ahmadi, Desain, and Thielen	article	modulated Gold codes	reconvolution, CCA	Bayesian dynamic stopping
Cantürk and Volosyak	proceedings	m-sequence	CCA	Language model (ChatGPT)
Cabrera Castillos and Dehais	dataset	burst codes		Grating stimuli
Dehais, Castillos, Ladouce, and Clisson	article	burst codes	Riemannian, logistic regression	Grating stimuli, dry EEG, comfort, eyestrain
Fodor, Herschel, Cantürk, Heisenberg, and Volosyak	article	m-sequence	CCA	Classification certainty feedback
E. Lai, Mai, Ji, Li, and Meng	proceedings	DIBS	filterbank task related component analysis (FBTRCA), LSTM	Asynchronous
Martínez-Cagigal, Álvaro Fernández-Rodríguez, Santamaría-Vázquez, Martín-Fernández, and Hornero	proceedings	non-binary m-sequence	CCA	Learning curve
Y. Miao et al.	article	white noise	TDCA, linear modeling, transfer learning	Minimal calibration, subject-to-subject transfer
Z. Miao, Meunier, Žák, and Grosse-Wentrup	proceedings	m-sequence	EEG2Code, EEGNet, Shallow-ConvNet, DeepConvNet, ShallowNet	Deep learning, transfer-learning, gine- tuning
Narayanan, Ahmadi, Desain, and Thielen	proceedings	modulated Gold codes	CCA	Gaze-independent, covert attention
Qu et al.	article	m-sequence	CCA	Biometrics
Scheppink, Ahmadi, Desain, Tangermann, and Thielen	proceedings	modulated Gold codes	CCA	Auditory, c-AEP
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 Tangermann	proceeding	modulated Gold codes	reconvolution, CCA, UMM	Calibration-free
Thielen, Farquhar, and Desain	dataset	modulated Gold codes		
Velut, Chevallier, Corsi, and Dehais	proceedings	burst codes	CNN, SPDNet, transfer learning	Subject-to-subject transfer
Zheng, Dong, et al. Zheng, Tian, et al.	article dataset	NRBS NRBS	FBCCA	Calibration-free, c-VEP versus SSVEP c-VEP versus SSVEP

Author(s)	Type	Stimulus	Decoding	Notes	
Ahmadi and Desain	preprint	modulated Gold codes	CCA	Bayesian dynamic stopping	
Cabrera Castillos, Ladouce,	article	m-sequence, burst codes	CNN		
Darmet, and Dehais					
Cabrera 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	1		CCA	D. d.,	
Henke et al.	proceedings	m-sequence	CCA TRCA	Background music	
C. Huang et al.	preprint article	white noise	Combined EEGNet	Visual tracking Biometrics	
Z. Huang, Liao, Ou, Chen, and Zhang	article	m-sequence	Combined EEGNet	Diometrics	
E. Lai, Mai, and Meng	proceedings	DIBS	FBTRCA, LSTM	fatigue	
Martínez-Cagigal et al.	article	m-sequence, non-binary m-	CCA	Eyestrain, fatigue	
Wartinez Cagigar et al.	articic	sequences	CCH	Lyestrant, rangae	
Martínez-Cagigal, Santamaría-	proceedings	non-binary m-sequence	CCA	Dynamic stopping	
Vázquez, and Hornero	r	yy		-)	
Moreno-Calderón et al.	article	m-sequence	CCA	Games	
Santamaría-Vázquez, Martínez-	proceedings	non-binary m-sequence,	EEG-inception		
Cagigal, and Hornero	1 0		•		
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	1.				
Volosyak et al.	proceedings	m-sequence	CCA	Gender	
Wolf and Götzelmann	article		TDCA	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	مسلنماء	<b>m</b> coguena	tuanafau laaunin a	Cubiact to aubiact transfer
Z. Huang, Zheng, Wu, and Wang	article	m-sequence	transfer-learning	Subject-to-subject transfer
Volosyak, Rezeika, Benda, Gem-	article	m-sequence	CCA	SSVEP, SSMVEP, c-VEP, BCI illiteracy
bler, and Stawicki				<b></b>
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,	proceedings	modulated Gold codes	CCA	Presentation rate (40, 60, 90, 120 Hz)
and Desain	-			
Desain, Thielen, van den Broek,	patent	modulated Gold codes	CCA	
and Farquhar	-			
Gembler and Volosyak	article	m-sequence	CCA	Language model
Gembler, Stawicki, Rezeika, and	proceedings	m-sequence	FBCCA	Presentation rate (30, 60, 120 Hz), age
Volosyak		-		(young, elderly)
Gembler, Stawicki, Saboor, and	article	m-sequence	FBCCA	Language model, dynamic stopping
Volosyak		-		
Gembler, Benda, Saboor, and	proceedings	m-sequence	FBCCA	Language model, dynamic stopping
Volosyak	-	-		
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		_		tracker
Kaya, Bohorquez, and Ozdamar	proceedings	quasi steady-state	CLAD	QSSVEP
Kaya, Bohórquez, and Özdamar	article	quasi steady-state	CLAD	QSSVEP
Kaya	dissertation			QSSVEP
Luo and Huang	proceedings	m-sequence	LDA, transfer learning	Subject-to-subject transfer
Matsuno, Itakura, Mizuno, and	proceedings	-	_	frequency-hopping VEP
Mito	-			
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 beam	- Fatigue
		_	forming	-
Turi and Clerc	article	m-sequence	<u> </u>	Static stopping number of cycles
Zhao, Wang, Liu, Pei, and Chen	article	m-sequence	FBCCA. FBTRCA	Biometrics
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

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 Spüler Thielen, Marsman, Farquhar, and Desain	proceedings article chapter	m-sequence, random codes m-sequence modulated Gold codes	CCA CCA reconvolution, CCA	Dry EEG, static and dynamic stopping 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 a	1.		proceedings	Gold code	correlation	Wheelchair
Nezamfar, S mus	Salehi, and E	rdog-	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, var quhar, and D	n den Broek, Jesain	Far-	article	modulated Gold codes	reconvolution, CCA	
Waytowich a	nd Krusienski	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-	proceedings	temporally jittered SSVEP	CLAD	
Nielsen, and Özdamar				
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)	Type	Stimulus	Decoding	Notes
Bin, Gao, Wang, Hong, and Gao	article	m-sequence	correlation	ERP versus SSVEP versus c-VEP

_	n	n	1
- 2	U	U	ľ

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	gs			

#### 

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

#### 

Author(s)	Type	Stimulus	Decoding	Notes	
Sutter	proceedings				

### References

- Ahmadi, S. (2019a). High density EEG measurment. Radboud University. Retrieved from https://doi.org/10.34973/psaf-mq72
- Ahmadi, S. (2019b). Sensor tying. Radboud University. Retrieved from https://doi.org/10.34973/ehq6-b836 doi: 10.34973/ehq6-b836
- Ahmadi, S., Borhanazad, M., Tump, D., Farquhar, J., & Desain, P. (2019). Low channel count montages using sensor tying for VEP-based BCI. *Journal of Neural Engineering*. Retrieved from https://doi.org/10.1088/1741-2552/ab4057 doi: 10.1088/1741-2552/ab4057
- Ahmadi, S., & Desain, P. (2023). A model-based dynamic stopping method for c-VEP BCI. In N. Mrachacz-Kersting, J. Collinger, D. Mattia, D. Valeriani, M. Vansteensel, & G. Müller-Putz (Eds.), *Proceedings of the 10th International Brain-Computer Interface Meeting 2023: Balancing Innovation and Translation.* Retrieved from https://doi.org/10.3217/978-3-85125-962-9-168 doi: 10.3217/978-3-85125-962-9-168
- Ahmadi, S., Desain, P., & Thielen, J. (2024). A bayesian dynamic stopping method for evoked response brain-computer interfacing. *Frontiers in Human Neuroscience*, 18. Retrieved from https://doi.org/10.3389/fnhum.2024.1437965 doi: 10.3389/fnhum.2024.1437965
- Aminaka, D., Makino, S., & Rutkowski, T. M. (2015a). Chromatic and high-frequency cVEP-based BCI paradigm. In 2015 37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 1906–1909). Retrieved from https://doi.org/10.1109/EMBC.2015.7318755 doi: 10.1109/EMBC.2015.7318755
- Aminaka, D., Makino, S., & Rutkowski, T. M. (2015b). Classification accuracy improvement of chromatic and high–frequency code–modulated visual evoked potential–based BCI. In *International Conference on Brain Informatics and Health* (pp. 232–241). Retrieved from https://doi.org/10.1007/978-3-319-23344-4\_23 doi: 10.1007/978-3-319-23344-4\_23
- Aminaka, D., Makino, S., & Rutkowski, T. M. (2015c). EEG filtering optimization for code–modulated chromatic visual evoked potential–based brain–computer interface. In *International Workshop on Symbiotic Interaction* (pp. 1–6). Retrieved from https://doi.org/10.1007/978-3-319-24917-9\_1 doi: 10.1007/978-3-319-24917-9\_1
- Aminaka, D., Makino, S., & Rutkowski, T. M. (2015d). SVM classification study of code-modulated visual evoked potentials. In 2015 Asia-Pacific Signal and Information Processing Association Annual Summit and Conference (APSIPA) (pp. 1065–1070). Retrieved from https://doi.org/10.1109/APSIPA.2015.7415435 doi: 10.1109/APSIPA.2015.7415435
- Aminaka, D., & Rutkowski, T. M. (2017). A sixteen-command and 40 Hz carrier frequency code-modulated visual evoked potential BCI. In *Brain-Computer Interface Research* (pp. 97–104). Springer. Retrieved from https://doi.org/10.1007/978-3-319-64373-1\_10 doi: 10.1007/978-3-319-64373-1\_10
- Başaklar, T., İder, Y. Z., & Tuncel, Y. (2018). Effects of high stimulus presentation rate on c-VEP based BCIs. In 2018 Medical Technologies National Congress (TIPTEKNO) (pp. 1–4). Retrieved from https://doi.org/10.1088/2057-1976/ab0cee doi: 10.1088/2057-1976/ab0cee
- Başaklar, T., Tuncel, Y., & İder, Y. Z. (2019). Effects of high stimulus presentation rate on EEG template characteristics and performance of c-VEP based BCIs. *Biomedical Physics & Engineering Express*. Retrieved from https://doi.org/10.1088/2057-1976/ab0cee doi: 10.1088/2057-1976/ab0cee
- Behboodi, M., Mahnam, A., Marateb, H., & Rabbani, H. (2020). Optimization of visual stimulus sequence in a brain-computer interface based on code modulated visual evoked potentials. *IEEE Transactions on Neural Systems and Rehabilitation Engineering*. Retrieved from https://doi.org/10.1109/TNSRE.2020.3044947 doi: 10.1109/TNSRE.2020.3044947
- Bin, G., Gao, X., Wang, Y., Hong, B., & Gao, S. (2009). VEP-based brain-computer interfaces: time, frequency, and code modulations. *IEEE Computational Intelligence Magazine*, 4(4), 22–26. Retrieved from https://doi.org/10.1109/MCI.2009.934562 doi: 10.1109/MCI.2009.934562
- Bin, G., Gao, X., Wang, Y., Li, Y., Hong, B., & Gao, S. (2011). A high-speed BCI based on code modulation VEP. *Journal of Neural Engineering*, 8(2), 025015. Retrieved from https://doi.org/10.1088/1741-2560/8/2/025015 doi: 10.1088/1741-2560/8/2/025015
- Bohórquez, J., Lozano, S., Kao, A., Toft-Nielsen, J., & Özdamar, Ö. (2013). Deconvolution and modeling of overlapping visual evoked potentials. In 2013 29th southern biomedical engineering conference (pp. 31–32). Retrieved from https://doi.org/10.1109/SBEC.2013.24 doi: 10.1109/SBEC.2013.24
- Bohórquez, J., & Özdamar, Ö. (2006). Signal to noise ratio analysis of maximum length sequence deconvolution of overlapping evoked potentials. *The Journal of the Acoustical Society of America*, 119(5), 2881–2888. Retrieved from https://doi.org/10.1121/1.2191609 doi: 10.1121/1.2191609

- Borhanazad, M., Thielen, J., Farquhar, J., & Desain, P. (2019). The effect of high and low frequencies in c-VEP BCI. In 8th Graz Brain-Computer Interface Conference 2019.

  Retrieved from https://doi.org/10.3217/978-3-85125-682-6-24 doi: 10.3217/978-3-85125-682-6-24
- Cabrera Castillos, K. (2023). 4-class code-VEP EEG data. Zenodo. Retrieved from https://doi.org/10.5281/zenodo.8255618 doi: 10.5281/zenodo.8255618
- Cabrera Castillos, K., & Dehais, F. (2024). 5-class burst c-VEP with dry EEG. Zenodo. Retrieved from https://doi.org/10.5281/zenodo.13838406 doi: 10.5281/zenodo.13838406
- Cabrera Castillos, K., Ladouce, S., Darmet, L., & Dehais, F. (2023). Burst c-VEP based BCI: Optimizing stimulus design for enhanced classification with minimal calibration data and improved user experience. *NeuroImage*, 284, 120446. Retrieved from https://doi.org/10.1016/j.neuroimage.2023.120446 doi: 10.1016/j.neuroimage.2023.120446
- Cantürk, A., & Volosyak, I. (2024). A novel ChatGPT-driven communication aid based on code-modulated visual evoked potentials (cVEP). In 9th Graz Brain-Computer Interface Conference 2024. Retrieved from https://doi.org/10.3217/978-3-99161-014-4-008 doi: 10.3217/978-3-99161-014-4-008
- Darmet, L., Ladouce, S., & Dehais, F. (2023). Shortened calibration of code-VEP based BCI by improved deep learning architecture and golden subjects pre-training. In 11th International IEEE EMBS Conference on Neural Engineering. Retrieved from https://hal.science/hal-03984091/
- Dehais, F., Castillos, K. C., Ladouce, S., & Clisson, P. (2024). Leveraging textured flickers: a leap toward practical, visually comfortable, and high-performance dry EEG code-VEP BCI. *Journal of Neural Engineering*, 21(6), 066023. Retrieved from https://doi.org/10.1088/1741-2552/ad8ef7 doi: 10.1088/1741-2552/ad8ef7
- Dehais, F., Ladouce, S., Darmet, L., Nong, T.-V., Ferraro, G., Torre Tresols, J., ... Labedan, P. (2022). Dual passive reactive brain-computer interface: A novel approach to human-machine symbiosis. *Frontiers in Neuroergonomics*, *3*, 824780. Retrieved from https://doi.org/10.3389/fnrgo.2022.824780 doi: 10.3389/fnrgo.2022.824780
- Desain, P., Farquhar, J., Blankespoor, J., & Gielen, S. (2008). Detecting spread spectrum pseudo random noise tags in EEG/MEG using a structure-based decomposition. In 4th International Brain-Computer Interface Workshop and Training Course 2008. Retrieved from https://www.tugraz.at/fileadmin/user\_upload/Institute/INE/Proceedings/Proceedings\_BCI\_Conference\_2008.pdf
- Desain, P., Thielen, J., van den Broek, P., & Farquhar, J. (2016). Brain computer interface using broadband evoked potentials. Google Patents. Retrieved from https://patents.google.com/patent/W02016012390A1/en (US Patent App. 15/328,083)
- Desain, P., Thielen, J., van den Broek, P., & Farquhar, J. (2019). Brain computer interface using broadband evoked potentials. Google Patents. (US Patent App. 10/314,508)
- Dimitriadis, S. I., & Marimpis, A. D. (2018). Enhancing performance and bit rates in a brain–computer interface system with phase-to-amplitude cross-frequency coupling: evidences from traditional c-VEP, fast c-VEP, and SSVEP designs. *Frontiers in Neuroinformatics*, 12, 19. Retrieved from https://doi.org/10.3389/fninf.2018.00019 doi: 10.3389/fninf.2018.00019
- Farquhar, J., Blankespoor, J., Vlek, R., & Desain, P. (2008). Towards a noise-tagging auditory BCI-paradigm. In 4th International Brain-Computer Interface Workshop and Training Course 2008. Retrieved from https://www.tugraz.at/fileadmin/user\_upload/Institute/INE/Proceedings/Proceedings\_BCI Conference 2008.pdf
- Fernández-Rodríguez, Á., Martínez-Cagigal, V., Santamaría-Vázquez, E., Ron-Angevin, R., & Hornero, R. (2023). Influence of spatial frequency in visual stimuli for cVEP-based BCIs: evaluation of performance and user experience. *Frontiers in Human Neuroscience*, 17. Retrieved from https://doi.org/10.3389/fnhum.2023.1288438 doi: 10.3389/fnhum.2023.1288438
- Fodor, M. A., Herschel, H., Cantürk, A., Heisenberg, G., & Volosyak, I. (2024). Evaluation of different visual feedback methods for brain—computer interfaces (BCI) based on code-modulated visual evoked potentials (cVEP). *Brain Sciences*, 14(8), 846. Retrieved from https://doi.org/10.3390/brainsci14080846 doi: 10.3390/brainsci14080846
- Gembler, F. (2020). Parameter optimization for brain-computer interfaces based on visual evoked potentials (Unpublished doctoral dissertation). Universität Bielefeld.
- Gembler, F., Benda, M., Rezeika, A., Stawicki, P. R., & Volosyak, I. (2020). Asynchronous c-VEP communication tools—efficiency comparison of low-target, multi-target and dictionary-assisted BCI spellers. *Scientific Reports*, 10(1), 1–13. Retrieved from https://doi.org/10.1038/s41598-020-74143-4 doi: 10.1038/s41598-020-74143-4

- Gembler, F., Benda, M., Saboor, A., & Volosyak, I. (2019). A multi-target c-VEP-based BCI speller utilizing n-gram word prediction and filter bank classification. In 2019 IEEE International Conference on Systems, Man and Cybernetics (SMC) (pp. 2719–2724). Retrieved from https://doi.org/10.1109/SMC.2019.8914235 doi: 10.1109/SMC.2019.8914235
- Gembler, F., Rezeika, A., Benda, M., & Volosyak, I. (2020). Five shades of grey: Exploring quintary m-sequences for more user-friendly c-VEP-based BCIs. *Computational Intelligence and Neuroscience*, 2020. Retrieved from https://doi.org/10.1155/2020/7985010 doi: 10.1155/2020/7985010
- Gembler, F., Stawicki, P., Rezeika, A., Benda, M., & Volosyak, I. (2020). Exploring session-to-session transfer for brain-computer interfaces based on code-modulated visual evoked potentials. In 2020 ieee international conference on systems, man, and cybernetics (smc) (pp. 1505–1510). Retrieved from https://doi.org/10.1109/SMC42975.2020.9282826 doi: 10.1109/SMC42975.2020.9282826
- Gembler, F., Stawicki, P., Rezeika, A., Saboor, A., Benda, M., & Volosyak, I. (2018). Effects of monitor refresh rates on c-VEP BCIs. In *International Workshop on Symbiotic Interaction* (pp. 53–62). Retrieved from https://doi.org/10.1007/978-3-319-91593-7\_6 doi: 10.1007/978-3-319-91593-7\_6
- Gembler, F., Stawicki, P., Rezeika, A., & Volosyak, I. (2019). A comparison of cVEP-based BCI-performance between different age groups. In *International Work-Conference on Artificial Neural Networks* (pp. 394–405). Retrieved from https://doi.org/10.1007/978-3-030-20521-8\_33 doi: 10.1007/978-3-030-20521-8\_33
- Gembler, F., Stawicki, P., Saboor, A., Benda, M., Grichnik, R., Rezeika, A., & Volosyak, I. (2018). A dictionary driven mental typewriter based on code-modulated visual evoked potentials (cVEP). In 2018 IEEE International Conference on Systems, Man, and Cybernetics (SMC) (pp. 619–624). Retrieved from https://doi.org/10.1109/SMC.2018.00114 doi: 10.1109/SMC.2018.00114
- Gembler, F., Stawicki, P., Saboor, A., & Volosyak, I. (2019). Dynamic time window mechanism for time synchronous VEP-based BCIs—performance evaluation with a dictionary-supported BCI speller employing SSVEP and c-VEP. *PLoS ONE*, 14(6), e0218177. Retrieved from https://doi.org/10.1371/journal.pone.0218177 doi: 10.1371/journal.pone.0218177
- Gembler, F., & Volosyak, I. (2019). A novel dictionary-driven mental spelling application based on code-modulated visual evoked potentials. *Computers*, 8(2), 33. Retrieved from https://doi.org/10.3390/computers8020033 doi: 10.3390/computers8020033
- Grigoryan, R., Filatov, D., & Kaplan, A. (2019). High-speed brain-computer communication interface based on code-modulated visual evoked potentials. *Bulletin of Russian State Medical University*(2). Retrieved from https://doi.org/10.24075/brsmu.2019.019 doi: 10.24075/brsmu.2019.019
- Hanagata, J., & Momose, K. (2002). A method for detecting gazed target using visual evoked potentials elicited by pseudorandom stimuli. In *Proc. 5th asia pacific conf. medical and biological engineering and 11th int. conf. biomedical engineering (icbme).*
- Henke, L., Rulffs, P., Adepoju, F., Stawicki, P., Cantürk, A., & Volosyak, I. (2023). Investigating the influence of background music on the performance of a cVEP-based BCI. In 2023 IEEE International Conference on Systems, Man, and Cybernetics (SMC) (pp. 1104–1109). Retrieved from https://doi.org/10.3390/healthcare11071014 doi: 10.3390/healthcare11071014
- Huang, C., Shi, N., Miao, Y., Chen, X., Wang, Y., & Gao, X. (2023). Visual tracking brain computer interface. arXiv preprint arXiv:2311.12592. Retrieved from https://doi.org/10.48550/arXiv.2311.12592 doi: 10.48550/arXiv.2311.12592
- Huang, Z., Liao, Z., Ou, G., Chen, L., & Zhang, Y. (2023). Authentication using c-vep evoked in a mild-burdened cognitive task. *Frontiers in Human Neuroscience*, 17, 1240451. Retrieved from https://doi.org/10.3389/fnhum.2023.1240451 doi: 10.3389/fnhum.2023.1240451
- Huang, Z., Zheng, W., Wu, Y., & Wang, Y. (2020). Ensemble or pool: A comprehensive study on transfer learning for c-VEP BCI during interpersonal interaction. *Journal of Neuroscience Methods*, 343, 108855. Retrieved from https://doi.org/10.1016/j.jneumeth.2020.108855 doi: 10.1016/j.jneumeth.2020.108855
- Isaksen, J., Mohebbi, A., & Puthusserypady, S. (2016). A comparative study of pseudorandom sequences used in a c-VEP based BCI for online wheelchair control.

  In 2016 38th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 1512–1515). Retrieved from https://doi.org/10.1109/EMBC.2016.7590997 doi: 10.1109/EMBC.2016.7590997
- Isaksen, J., Mohebbi, A., & Puthusserypady, S. (2017). Optimal pseudorandom sequence selection for online c-VEP based BCI control applications. *PLoS ONE*, 12(9), e0184785. Retrieved from https://doi.org/10.1371/journal.pone.0184785 doi: 10.1371/journal.pone.0184785

- Kadıoğlu, B., Yıldız, İ., Closas, P., Fried-Oken, M. B., & Erdoğmuş, D. (2019). Robust fusion of c-VEP and gaze. *IEEE Sensors Letters*, 3(1), 1–4. Retrieved from https://doi.org/10.1109/LSENS.2018.2878705 doi: 10.1109/LSENS.2018.2878705
- Kapeller, C., Hintermüller, C., Abu-Alqumsan, M., Prückl, R., Peer, A., & Guger, C. (2013). A BCI using VEP for continuous control of a mobile robot. In 2013 35th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 5254–5257). Retrieved from https://doi.org/10.1109/EMBC.2013.6610734 doi: 10.1109/EMBC.2013.6610734
- Kapeller, C., Kamada, K., Ogawa, H., Prueckl, R., Scharinger, J., & Guger, C. (2014). An electrocorticographic BCI using code-based VEP for control in video applications: a single-subject study. *Frontiers in Systems Neuroscience*, 8, 139. Retrieved from https://doi.org/10.3389/fnsys.2014.00139 doi: 10.3389/fnsys.2014.00139
- Kaya, I. (2019). High rate quasi-steady-state pattern visual evoked potentials for brain-computer interface (Unpublished doctoral dissertation). University of Miami.
- Kaya, I., Bohórquez, J., & Özdamar, Ö. (2019). A BCI gaze sensing method using low jitter code modulated VEP. Sensors, 19(17), 3797. Retrieved from https://doi.org/10.3390/s19173797 doi: 10.3390/s19173797
- Kaya, I., Bohorquez, J., & Özdamar, Ö. (2021). BCI performance improvement by special low jitter quasi-steady-state VEP paradigm. *Modern Approaches to Augmentation of Brain Function*, 121–139. Retrieved from https://doi.org/10.1007/978-3-030-54564-2\_7 doi: 10.1007/978-3-030-54564-2\_7
- Kaya, I., Bohorquez, J. E., & Ozdamar, O. (2019). Brain computer interface switch based on quasi-steady-state visual evoked potentials. In 2019 9th international ieee/embs conference on neural engineering (ner) (pp. 1175–1178).
- Lai, E., Mai, X., Ji, M., Li, S., & Meng, J. (2024). High-frequency discrete-interval binary sequence in asynchronous c-VEP-based BCI for visual fatigue reduction. *IEEE Journal of Biomedical and Health Informatics*. Retrieved from https://doi.org/10.1109/BIBM58861.2023.10385977 doi: 10.1109/BIBM58861.2023.10385977
- Lai, E., Mai, X., & Meng, J. (2023). High-performance deep neural network pretrained with contrastive learning for asynchronous high-frequency c-VEP detection. In 2023 IEEE International Conference on Bioinformatics and Biomedicine (BIBM) (pp. 2515–2521). Retrieved from https://doi.org/10.1109/BIBM58861.2023.10385977 doi: 10.1109/BIBM58861.2023.10385977
- Lai, S. M., Zhang, Z., Hung, Y. S., Niu, Z., & Chang, C. (2011). A chromatic transient visual evoked potential based encoding/decoding approach for brain-computer interface. *IEEE Journal on Emerging and Selected Topics in Circuits and Systems*, 1(4), 578–589. Retrieved from https://doi.org/10.1109/JETCAS.2011.2178734 doi: 10.1109/JETCAS.2011.2178734
- Liu, Y., Wei, Q., & Lu, Z. (2018). A multi-target brain-computer interface based on code modulated visual evoked potentials. *PLoS ONE*, 13(8), e0202478. Retrieved from https://doi.org/10.1371/journal.pone.0202478 doi: 10.1371/journal.pone.0202478
- Luo, D., & Huang, Z. (2019). A subject-transfer study on detecting c-VEP. In 2019 Chinese Automation Congress (CAC) (pp. 2956–2959). Retrieved from https://doi.org/10.1109/CAC48633.2019.8996478 doi: 10.1109/CAC48633.2019.8996478
- Martínez-Cagigal, V., Santamaría-Vázquez, E., & Hornero, R. (2023). Toward early stopping detection for non-binary c-vep-based bcis: A pilot study. In I. Rojas, G. Joya, & A. Catala (Eds.), *Advances in Computational Intelligence* (pp. 580–590). Cham: Springer Nature Switzerland. Retrieved from https://doi.org/10.1007/978-3-031-43078-7\_47 doi: 10.1007/978-3-031-43078-7\_47
- Martínez-Cagigal, V., Santamaría-Vázquez, E., Pérez-Velasco, S., Marcos-Martínez, D., Moreno-Calderón, S., & Hornero, R. (2023). Non-binary m-sequences for more comfortable brain–computer interfaces based on c-VEPs. *Expert Systems with Applications*, 120815. Retrieved from https://doi.org/10.1016/j.eswa.2023.120815
- Martínez-Cagigal, V., Thielen, J., Hornero, R., & Desain, P. (2025). The role of code-modulated evoked potentials in next-generation brain-computer interfacing. *Frontiers in Human Neuroscience*, 19, 1548183. Retrieved from https://doi.org/10.3389/fnhum.2025.1548183 doi: 10.3389/fnhum.2025.1548183
- Martínez-Cagigal, V., Thielen, J., Santamaría-Vázquez, E., Pérez-Velasco, S., Desain, P., & Hornero, R. (2021). Brain–computer interfaces based on code-modulated visual evoked potentials (c-VEP): a literature review. *Journal of Neural Engineering*. Retrieved from https://doi.org/10.1088/1741-2552/ac38cf doi: 10.1088/1741-2552/ac38cf
- Martínez-Cagigal, V., Álvaro Fernández-Rodríguez, Santamaría-Vázquez, E., Martín-Fernández, A., & Hornero, R. (2024). Assessing calibration durations for c-

- VEP-based BCIs: Insights from non-binary patterns and spatial frequency variations. In 9th Graz Brain-Computer Interface Conference 2024. Retrieved from https://doi.org/10.3217/978-3-99161-014-4-041 doi: 10.3217/978-3-99161-014-4-041
- Matsuno, S., Itakura, N., Mizuno, T., & Mito, K. (2019). Examination of multi-optioning for cVEP-based BCI by fluctuation of indicator lighting intervals and luminance. In 2019 IEEE International Conference on Systems, Man and Cybernetics (SMC) (pp. 2743–2747). Retrieved from https://doi.org/10.1109/SMC.2019.8914445 doi: 10.1109/SMC.2019.8914445
- Miao, Y., Shi, N., Huang, C., Song, Y., Chen, X., Wang, Y., & Gao, X. (2024). High-performance c-VEP-BCI under minimal calibration. *Expert Systems with Applications*, 249, 123679. Retrieved from https://doi.org/10.1016/j.eswa.2024.123679 doi: 10.1016/j.eswa.2024.123679
- Miao, Z., Meunier, A., Žák, M. R., & Grosse-Wentrup, M. (2024). Exploring artificial neural network models for c-VEP decoding in a brain-artificial intelligence interface. In 2024 ieee international conference on bioinformatics and biomedicine (bibm) (pp. 4178–4183). Retrieved from https://eprints.cs.univie.ac.at/8197/
- Mohebbi, A., Engelsholm, S. K., Puthusserypady, S., Kjaer, T. W., Thomsen, C. E., & Sorensen, H. B. (2015). A brain computer interface for robust wheelchair control application based on pseudorandom code modulated visual evoked potential. In 2015 37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 602–605). Retrieved from https://doi.org/10.1109/EMBC.2015.7318434 doi: 10.1109/EMBC.2015.7318434
- Momose, K. (2007). Evaluation of an eye gaze point detection method using VEP elicited by multi-pseudorandom stimulation for brain computer interface. In 2007 29th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (pp. 5063–5066). Retrieved from https://doi.org/10.1109/IEMBS.2007.4353478 doi: 10.1109/IEMBS.2007.4353478
- Momose, K. (2008). Simultaneous detection method of P300 event-related potentials and eye gaze point using multi-pseudorandom and flash stimulation for brain computer interface. In 2008 30th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (pp. 666–669). Retrieved from https://doi.org/10.1109/IEMBS.2008.4649240 doi: 10.1109/IEMBS.2008.4649240
- Moreno-Calderón, S., Martínez-Cagigal, V., Santamaría-Vázquez, E., Pérez-Velasco, S., Marcos-Martínez, D., & Hornero, R. (2023). Combining brain-computer interfaces and multiplayer video games: an application based on c-VEPs. *Frontiers in Human Neuroscience*, 17. Retrieved from https://doi.org/10.3389/fnhum.2023.1227727 doi: 10.3389/fnhum.2023.1227727
- Nagel, S. (2019). Towards a home-use BCI: fast asynchronous control and robust non-control state detection (Unpublished doctoral dissertation). Universität Tübingen.
- Nagel, S., Dreher, W., Rosenstiel, W., & Spüler, M. (2018). The effect of monitor raster latency on VEPs, ERPs and brain-computer interface performance. *Journal of Neuroscience Methods*, 295, 45–50. Retrieved from https://doi.org.10.1016/j.jneumeth.2017.11.018 doi: 10.1016/j.jneumeth.2017.11.018
- Nagel, S., Rosenstiel, W., & Spüler, M. (2017). Random visual evoked potentials (rVEP) for brain-computer interface (BCI) control. In 7th Graz Brain-Computer Interface Conference 2017. Retrieved from https://doi.org/10.3217/978-3-85125-533-1-64 doi: 10.3217/978-3-85125-533-1-64
- Nagel, S., Rosenstiel, W., & Spüler, M. (2018). Finding optimal stimulation patterns for BCIs based on visual evoked potentials. In *Proceedings of the 7th International Brain-Computer Interface Meeting*. BCI Society (pp. 164–165). Retrieved from https://bcisociety.org/wp-content/uploads/2019/03/2018AbstractBook.pdf
- Nagel, S., & Spüler, M. (2018). Modelling the brain response to arbitrary visual stimulation patterns for a flexible high-speed brain-computer interface. *PLoS ONE*, 13(10), e0206107. Retrieved from https://doi.org/10.1371/journal.pone.0206107 doi: 10.1371/journal.pone.0206107
- Nagel, S., & Spüler, M. (2019a). Asynchronous non-invasive high-speed BCI speller with robust non-control state detection. *Scientific Reports*, 9(1), 1–9. Retrieved from https://doi.org/10.1038/s41598-019-44645-x doi: 10.1038/s41598-019-44645-x
- Nagel, S., & Spüler, M. (2019b). World's fastest brain-computer interface: combining EEG2Code with deep learning. *PLoS ONE*, 14(9), e0221909. Retrieved from https://doi.org/10.1371/journal.pone.0221909 doi: 10.1371/journal.pone.0221909
- Nakanishi, M., & Mitsukura, Y. (2012). Periodicity detection for BCI based on periodic code modulation visual evoked potentials. In 2012 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP) (pp. 665–668). Retrieved from https://doi.org/10.1109/ICASSP.2012.6287971 doi: 10.1109/ICASSP.2012.6287971
- Narayanan, S., Ahmadi, S., Desain, P., & Thielen, J. (2024). Towards gaze-independent c-VEP BCI: A pilot study. In 9th Graz Brain-Computer Interface Conference 2024. Retrieved from https://doi.org/10.3217/978-3-99161-014-4-060 doi: 10.3217/978-3-99161-014-4-060

- Nezamfar, H., Mohseni Salehi, S., Higger, M., & Erdogmus, D. (2018). Code-VEP vs. eye tracking: A comparison study. *Brain Sciences*, 8(7), 130. Retrieved from https://doi.org/10.3390/brainsci8070130 doi: 10.3390/brainsci8070130
- Nezamfar, H., Orhan, U., Erdogmus, D., Hild, K., Purwar, S., Oken, B., & Fried-Oken, M. (2011). On visually evoked potentials in eeg induced by multiple pseudorandom binary sequences for brain computer interface design. In 2011 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP) (pp. 2044–2047). Retrieved from https://doi.org/10.1109/ICASSP.2011.5946914 doi: 10.1109/ICASSP.2011.5946914
- Nezamfar, H., Orhan, U., Purwar, S., Hild, K., Oken, B., & Erdogmus, D. (2011). Decoding of multichannel EEG activity from the visual cortex in response to pseudorandom binary sequences of visual stimuli. *International Journal of Imaging Systems and Technology*, 21(2), 139–147. Retrieved from https://doi.org/10.1002/ima.20288 doi: 10.1002/ima.20288
- Nezamfar, H., Salehi, S. S. M., & Erdogmus, D. (2015). Stimuli with opponent colors and higher bit rate enable higher accuracy for c-VEP BCI. In 2015 IEEE Signal Processing in Medicine and Biology Symposium (SPMB) (pp. 1–6). Retrieved from https://doi.org/10.1109/SPMB.2015.7405476 doi: 10.1109/SPMB.2015.7405476
- Nezamfar, H., Salehi, S. S. M., Moghadamfalahi, M., & Erdogmus, D. (2016). Flashtype<sup>TM</sup>: A context-aware c-VEP-based BCI typing interface using EEG signals. *IEEE Journal of Selected Topics in Signal Processing*, 10(5), 932–941. Retrieved from https://doi.org/10.1109/JSTSP.2016.2552140 doi: 10.1109/JSTSP.2016.2552140
- Peng, F., & Huang, Z. (2019). A c-VEP BCI system for psychological experiments. In 2019 12th International Congress on Image and Signal Processing, BioMedical Engineering and Informatics (CISP-BMEI) (pp. 1–5). Retrieved from https://doi.org/10.1109/CISP-BMEI48845.2019.8966018 doi: 10.1109/CISP-BMEI48845.2019.8966018
- Qu, H., Zhao, H., Wei, Q., Pei, W., Gao, X., & Wang, Y. (2024). Combing multiple visual stimuli to enhance the performance of VEP-based biometrics. *IEEE Transactions on Information Forensics and Security*. Retrieved from https://doi.org/10.1109/TIFS.2024.3452628 doi: 10.1109/TIFS.2024.3452628
- Riechmann, H., Finke, A., & Ritter, H. (2013). Hierarchical codebook visually evoked potentials for fast and flexible BCIs. In 2013 35th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 2776–2779). Retrieved from https://doi.org/10.1109/EMBC.2013.6610116 doi: 10.1109/EMBC.2013.6610116
- Riechmann, H., Finke, A., & Ritter, H. (2016). Using a cVEP-based brain-computer interface to control a virtual agent. *IEEE Transactions on Neural Systems and Rehabilitation Engineering*, 24(6), 692–699. Retrieved from https://doi.org/10.1109/TNSRE.2015.2490621 doi: 10.1109/TNSRE.2015.2490621
- Santamaría-Vázquez, E., Martínez-Cagigal, V., & Hornero, R. (2023). Bit-wise reconstruction of non-binary visual stimulation patterns from EEG using deep learning: A promising alternative for user-friendly high-speed c-VEP-based BCIs. In I. Rojas, G. Joya, & A. Catala (Eds.), *Advances in Computational Intelligence* (pp. 603–614). Cham: Springer Nature Switzerland. Retrieved from https://doi.org/10.1007/978-3-031-43078-7\_49 doi: 10.1007/978-3-031-43078-7\_49
- Sato, J.-i., & Washizawa, Y. (2015). Reliability-based automatic repeat request for short code modulation visual evoked potentials in brain computer interfaces. In 2015 37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 562–565). Retrieved from https://doi.org/10.1109/EMBC.2015.7318424 doi: 10.1109/EMBC.2015.7318424
- Sato, J.-i., & Washizawa, Y. (2016). Neural decoding of code modulated visual evoked potentials by spatio-temporal inverse filtering for brain computer interfaces. In 2016 38th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 1484–1487). Retrieved from https://doi.org/10.1109/EMBC.2016.7590990 doi: 10.1109/EMBC.2016.7590990
- Scheppink, H., Ahmadi, S., Desain, P., Tangermann, M., & Thielen, J. (2024). Towards auditory attention decoding with noise-tagging: A pilot study. In *9th Graz Brain-Computer Interface Conference* 2024. Retrieved from https://doi.org/10.3217/978-3-99161-014-4-059 doi: 10.3217/978-3-99161-014-4-059
- Shi, N., Miao, Y., Huang, C., Li, X., Song, Y., Chen, X., ... Gao, X. (2024). Estimating and approaching the maximum information rate of noninvasive visual brain-computer interface. *NeuroImage*, 120548. Retrieved from https://doi.org/10.1016/j.neuroimage.2024.120548 doi: 10.1016/j.neuroimage.2024.120548
- Shirzhiyan, Z., Keihani, A., Farahi, M., Shamsi, E., GolMohammadi, M., Mahnam, A., ... Jafari, A. H. (2019). Introducing chaotic codes for the modulation of code modulated visual evoked potentials (c-VEP) in normal adults for visual fatigue reduction. *PLoS ONE*, 14(3), e0213197. Retrieved from https://doi.org/

- 10.1371/journal.pone.0213197 doi: 10.1371/journal.pone.0213197
- Shirzhiyan, Z., Keihani, A., Farahi, M., Shamsi, E., GolMohammadi, M., Mahnam, A., ... Jafari, A. H. (2020). Toward new modalities in VEP-based BCI applications using dynamical stimuli: Introducing quasi-periodic and chaotic VEP-based BCI. *Frontiers in Neuroscience*, 14. Retrieved from https://doi.org/10.3389/fnins.2020.534619 doi: 10.3389/fnins.2020.534619
- Spüler, M. (2015). A brain-computer interface (BCI) system to use arbitrary Windows applications by directly controlling mouse and keyboard. In 2015 37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 1087–1090). Retrieved from https://doi.org/10.1109/EMBC.2015.7318554 doi: 10.1109/EMBC.2015.7318554
- Spüler, M. (2017). A high-speed brain-computer interface (BCI) using dry EEG electrodes. *PLoS ONE*, 12(2), e0172400. Retrieved from https://doi.org/10.1371/journal.pone.0172400 doi: 10.1371/journal.pone.0172400
- Spüler, M., & Kurek, S. (2018). Alpha-band lateralization during auditory selective attention for brain–computer interface control. *Brain-Computer Interfaces*, 5(1), 23–29. Retrieved from https://doi.org/10.1080/2326263X.2017.1415496 doi: 10.1080/2326263X.2017.1415496
- Spüler, M., Rosenstiel, W., & Bogdan, M. (2012a). One class SVM and canonical correlation analysis increase performance in a c-VEP based brain-computer interface (BCI). In ESANN. Retrieved from https://www.esann.org/sites/default/files/proceedings/legacy/es2012-18.pdf
- Spüler, M., Rosenstiel, W., & Bogdan, M. (2012b). Online adaptation of a c-VEP brain-computer interface (BCI) based on error-related potentials and unsupervised learning. *PLoS ONE*, 7(12), e51077. Retrieved from https://doi.org/10.1371/journal.pone.0051077 doi: 10.1371/journal.pone.0051077
- Spüler, M., Rosenstiel, W., & Bogdan, M. (2013a). Unsupervised BCI calibration as possibility for communication in CLIS patients. In *Proceedings of the Fifth International Brain-Computer Interface Meeting* (Vol. 2013). Retrieved from https://doi.org/10.3217/978-3-85125-260-6-122 doi: 10.3217/978-3-85125-260-6-122
- Spüler, M., Rosenstiel, W., & Bogdan, M. (2013b). Unsupervised online calibration of a c-VEP brain-computer interface (BCI). In *International Conference on Artificial Neural Networks* (pp. 224–231). Retrieved from https://doi.org/10.1007/978-3-642-40728-4\_28 doi: 10.1007/978-3-642-40728-4\_28
- Spüler, M., Walter, A., Rosenstiel, W., & Bogdan, M. (2013). Spatial filtering based on canonical correlation analysis for classification of evoked or event-related potentials in EEG data. *IEEE Transactions on Neural Systems and Rehabilitation Engineering*, 22(6), 1097–1103. Retrieved from https://doi.org/10.1109/TNSRE.2013.2290870 doi: 10.1109/TNSRE.2013.2290870
- Stawicki, P., & Volosyak, I. (2022). cVEP training data validation towards optimal training set composition from multi-day data. *Brain Sciences*, 12(2). Retrieved from <a href="https://doi.org.10.3390/brainsci12020234">https://doi.org.10.3390/brainsci12020234</a> doi: 10.3390/brainsci12020234
- Sun, Q., Zhang, S., Dong, G., Pei, W., Gao, X., & Wang, Y. (2024). High-density electroencephalogram facilitates the detection of small stimuli in code-modulated visual evoked potential brain–computer interfaces. *Sensors*, 24(11), 3521. Retrieved from https://doi.org/10.3390/s24113521 doi: 10.3390/s24113521
- Sun, Q., Zheng, L., Pei, W., Gao, X., & Wang, Y. (2022). A 120-target brain-computer interface based on code-modulated visual evoked potentials. *Journal of Neuroscience Methods*, 109597. Retrieved from https://doi.org/10.1016/j.jneumeth.2022.109597 doi: 10.1016/j.jneumeth.2022.109597
- Sutter, E. E. (1984). The visual evoked response as a communication channel. In Proceedings of the IEEE Symposium on Biosensors (pp. 95–100).
- Sutter, E. E. (1992). The brain response interface: communication through visually-induced electrical brain responses. *Journal of Microcomputer Applications*, 15(1), 31–45. Retrieved from https://doi.org/10.1016/0745-7138(92)90045-7 doi: 10.1016/0745-7138(92)90045-7
- Thielen, J. (2023). Effects of stimulus sequences on brain-computer interfaces using code-modulated visual evoked potentials: An offline simulation. In I. Rojas, G. Joya, & A. Catala (Eds.), *Advances in Computational Intelligence* (pp. 555–568). Cham: Springer Nature Switzerland. Retrieved from https://doi.org/10.1007/978-3-031-43078-7\_45 doi: 10.1007/978-3-031-43078-7\_45
- Thielen, J., Cornielje, G., van der Werff, F., & Desain, P. (2023). A comparison of stimulus sequences for code-modulated visual evoked potential (c-VEP) based BCI. In N. Mrachacz-Kersting, J. Collinger, D. Mattia, D. Valeriani, M. Vansteensel, & G. Müller-Putz (Eds.), *Proceedings of the 10th International Brain-Computer Interface Meeting 2023: Balancing Innovation and Translation.* Retrieved from https://doi.org/10.3217/978-3-85125-962-9-139 doi: 10.3217/978-3-85125-962-9-139
- Thielen, J., Farquhar, J., & Desain, P. (2016). Re(con)volution: accurate response prediction for BBVEP-based BCI. In G. R. Müller-Putz, J. E. Huggins, & D. Steyrl (Eds.), *Proceedings of the Sixth International Brain-Computer Interface Meeting: BCI Past, Present, and Future.* Retrieved from https://doi.org/10.3217/978-3

- -85125-467-9-34 doi: 10.3217/978-3-85125-467-9-34
- Thielen, J., Farquhar, J., & Desain, P. (2024). *Broad-Band Visually Evoked Potentials: Re(con)volution in Brain-Computer Interfacing*. Radboud University. Retrieved from <a href="https://doi.org/10.34973/1ecz-1232">https://doi.org/10.34973/1ecz-1232</a> doi: 10.34973/1ecz-1232
- Thielen, J., Marsman, P., Farquhar, J., & Desain, P. (2017). Re(con)volution: Accurate response prediction for broad-band evoked potentials-based brain computer interfaces. In *Brain-Computer Interface Research* (pp. 35–42). Springer. Retrieved from https://doi.org/10.1007/978-3-319-64373-1\_4 doi: 10.1007/978-3-319-64373-1\_4
- Thielen, J., Marsman, P., Farquhar, J., & Desain, P. (2021). From full calibration to zero training for a code-modulated visual evoked potentials for brain-computer interface. *Journal of Neural Engineering*, 18(5), 056007. Retrieved from https://doi.org/10.1088/1741-2552/abecef doi: 10.1088/1741-2552/abecef
- Thielen, J., Marsman, P., Farquhar, J., & Desain, P. (2023). From full calibration to zero training for a code-modulated visual evoked potentials brain computer interface. Radboud University. Retrieved from https://doi.org/10.34973/9txv-z787 doi: 10.34973/9txv-z787
- Thielen, J., Sosulski, J., & Tangermann, M. (2024). Exploring new territory: Calibration-free decoding for c-VEP BCI. In 9th Graz Brain-Computer Interface Conference 2024. Retrieved from https://doi.org/10.3217/978-3-99161-014-4-057 doi: 10.3217/978-3-99161-014-4-057
- Thielen, J., van den Broek, P., Farquhar, J., & Desain, P. (2015). Broad-band visually evoked potentials: re(con)volution in brain-computer interfacing. *PLoS ONE*, 10(7), e0133797. Retrieved from https://doi.org/10.1371/journal.pone.0133797 doi: 10.1371/journal.pone.0133797
- Torres, J. A. R., & Daly, I. (2021). How to build a fast and accurate code-modulated brain-computer interface. *Journal of Neural Engineering*. Retrieved from https://doi.org/10.1088/1741-2552/abfaac doi: 10.1088/1741-2552/abfaac
- Tu, Y., Hung, Y. S., Hu, L., Huang, G., Hu, Y., & Zhang, Z. (2014). An automated and fast approach to detect single-trial visual evoked potentials with application to brain-computer interface. *Clinical Neurophysiology*, 125(12), 2372–2383. Retrieved from https://doi.org/10.1016/j.clinph.2014.03.028 doi: 10.1016/j.clinph.2014.03.028
- Turi, F., & Clerc, M. (2019). Adaptive parameter setting in a code modulated visual evoked potentials BCI. *HAL Open Science*. Retrieved from https://inria.hal.science/hal-02303562/document
- Turi, F., Gayraud, N., & Clerc, M. (2018). Zero-calibration cVEP BCI using word prediction: a proof of concept. *arXiv preprint arXiv:1810.03428*. Retrieved from https://doi.org/10.48550/arXiv.1810.03428 doi: 10.48550/arXiv.1810.03428
- Turi, F., Gayraud, T., & Clerc, M. (2020). Auto-calibration of c-VEP BCI by word prediction. *HAL Open Science*. Retrieved from https://hal.science/hal-02844024/document
- Velut, S., Chevallier, S., Corsi, M.-C., & Dehais, F. (2024). Deep Riemannian neural architectures for domain adaptation in burst cVEP-based brain computer interface. In *Esann* 2024 (pp. 571–576). Retrieved from https://hal.science/hal-04720928/
- Verbaarschot, C., Tump, D., Lutu, A., Borhanazad, M., Thielen, J., van den Broek, P., ... others (2021). A visual brain-computer interface as communication aid for patients with amyotrophic lateral sclerosis. *Clinical Neurophysiology*, 132(10), 2404–2415. Retrieved from https://doi.org/10.1016/j.clinph.2021.07.012 doi: 10.1016/j.clinph.2021.07.012
- Volosyak, I., Adepoju, F., Stawicki, P., Rulffs, P., Cantürk, A., & Henke, L. (2023). Gender influence on cvep-based bci performance. In I. Rojas, G. Joya, & A. Catala (Eds.), *Advances in Computational Intelligence* (pp. 591–602). Cham: Springer Nature Switzerland. Retrieved from https://doi.org/10.1007/978-3-031-43078-7\_48 doi: 10.1007/978-3-031-43078-7\_48
- Volosyak, I., Rezeika, A., Benda, M., Gembler, F., & Stawicki, P. (2020). Towards solving of the illiteracy phenomenon for VEP-based brain-computer interfaces. *Biomedical Physics & Engineering Express*, 6(3), 035034. Retrieved from https://doi.org/10.1088/2057-1976/ab87e6 doi: 10.1088/2057-1976/ab87e6
- Waytowich, N. R., & Krusienski, D. J. (2015). Spatial decoupling of targets and flashing stimuli for visual brain–computer interfaces. *Journal of Neural Engineering*, 12(3), 036006. Retrieved from https://doi.org/10.1088/1741-2560/12/3/036006 doi: 10.1088/1741-2560/12/3/036006
- Wei, Q., Feng, S., & Lu, Z. (2016). Stimulus specificity of brain-computer interfaces based on code modulation visual evoked potentials. *PLoS ONE*, 11(5), e0156416. Retrieved from https://doi.org/10.1371/journal.pone.0156416 doi: 10.1371/journal.pone.0156416

- Wei, Q., Gong, H., & Lu, Z. (2017). Grouping modulation with different codes for improving performance in cVEP-based brain–computer interfaces. *Electronics Letters*, 53(4), 214–216. Retrieved from https://doi.org/10.1049/el.2016.4006 doi: 10.1049/el.2016.4006
- Wei, Q., Huang, Y., Li, M., & Lu, Z. (2016). VEP-based brain-computer interfaces modulated by Golay complementary series for improving performance. *Technology and Health Care*, 24(s2), S541–S549. Retrieved from https://doi.org/10.3233/THC-161180 doi: 10.3233/THC-161180
- Wei, Q., Liu, Y., Gao, X., Wang, Y., Yang, C., Lu, Z., & Gong, H. (2018). A novel c-VEP BCI paradigm for increasing the number of stimulus targets based on grouping modulation with different codes. *IEEE Transactions on Neural Systems and Rehabilitation Engineering*, 26(6), 1178–1187. Retrieved from https://doi.org/10.1109/TNSRE.2018.2837501 doi: 10.1109/TNSRE.2018.2837501
- Wittevrongel, B., Van Wolputte, E., & Van Hulle, M. M. (2017). Code-modulated visual evoked potentials using fast stimulus presentation and spatiotemporal beamformer decoding. *Scientific Reports*, 7(1), 15037. Retrieved from https://doi.org/10.1038/s41598-017-15373-x doi: 10.1038/s41598-017-15373-x
- Wolf, P., & Götzelmann, T. (2023). VEPdgets: Towards richer interaction elements based on visually evoked potentials. *Sensors*, 23(22), 9127. Retrieved from https://doi.org/10.3390/s23229127 doi: 10.3390/s23229127
- Xu, H., Hsu, S.-H., Nakanishi, M., Lin, Y., Jung, T.-P., & Cauwenberghs, G. (2023). Stimulus design for visual evoked potential based brain-computer interfaces. *IEEE Transactions on Neural Systems and Rehabilitation Engineering*. Retrieved from https://doi.org/10.1109/TNSRE.2023.3280081 doi: 10.1109/TNSRE.2023.3280081
- Yasinzai, M. N., & Ider, Y. Z. (2020). New approach for designing cVEP BCI stimuli based on superposition of edge responses. *Biomedical Physics & Engineering Express*, 6(4), 045018. Retrieved from https://doi.org/10.1088/2057-1976/ab98e7 doi: 10.1088/2057-1976/ab98e7
- Ying, J., Wei, Q., & Zhou, X. (2022). Riemannian geometry-based transfer learning for reducing training time in c-VEP BCIs. *Scientific Reports*, 12(1), 1–15. Retrieved from https://doi.org/10.1038/s41598-022-14026-y doi: 10.1038/s41598-022-14026-y
- Zarei, A., & Asl, B. M. (2022a). Automatic detection of code-modulated visual evoked potentials using novel covariance estimators and short-time EEG signals. Computers in Biology and Medicine, 105771. Retrieved from https://doi.org/10.1016/j.compbiomed.2022.105771 doi: 10.1016/j.compbiomed.2022.105771
- Zarei, A., & Asl, B. M. (2022b). Classification of code-modulated visual evoked potentials using adaptive modified covariance beamformer and EEG signals. *Computer Methods and Programs in Biomedicine*, 221, 106859. Retrieved from https://doi.org/10.1016/j.cmpb.2022.106859 doi: 10.1016/j.cmpb.2022.106859
- Zhao, H., Wang, Y., Liu, Z., Pei, W., & Chen, H. (2019). Individual identification based on code-modulated visual-evoked potentials. *IEEE Transactions on Information Forensics and Security*, 14(12), 3206–3216. Retrieved from https://doi.org/10.1109/TIFS.2019.2912272 doi: 10.1109/TIFS.2019.2912272
- Zheng, L., Dong, Y., Tian, S., Pei, W., Gao, X., & Wang, Y. (2024). A calibration-free c-VEP based BCI employing narrow-band random sequences. *Journal of Neural Engineering*. Retrieved from https://doi.org/10.1088/1741-2552/ad3679 doi: 10.1088/1741-2552/ad3679
- Zheng, L., Pei, W., Gao, X., Zhang, L., & Wang, Y. (2022). A high-performance brain switch based on code-modulated visual evoked potentials. *Journal of Neural Engineering*. Retrieved from https://doi.org/10.1088/1741-2552/ac494f doi: 10.1088/1741-2552/ac494f
- Zheng, L., Tian, S., Dong, Y., Pei, W., Gao, X., & Wang, Y. (2024). A large dataset for VEP based brain-computer interfaces employing narrow-band code modulation and frequency-phase modulation. *Brain-Apparatus Communication: A Journal of Bacomics*, 3(1), 2383860. Retrieved from https://doi.org/10.1080/27706710.2024.2383860 doi: 10.1080/27706710.2024.2383860
- Zheng, L., Wang, Y., Pei, W., & Chen, H. (2019). A fast brain switch based on multi-class code-modulated VEPs. In 2019 41st Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC) (pp. 3058–3061). Retrieved from https://doi.org/10.1109/EMBC.2019.8857617 doi: 10.1109/EMBC.2019.8857617

# Glossary

<b>6FO</b> 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
<b>c-AEP</b> 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