

2025

| Author(s) | Type | Stimulus | Decoding | Notes |
|---|-------------|----------------------|----------------------|--|
| Dold et al. | article | modulated Gold codes | reconvolution, CCA | Dareplane, experiment platform |
| Dong, Zheng, Pei, Gao, and Wang | article | NBRS | EEG2Code | 240 classes |
| Fodor, Canturk, Le, Heisenberg, and Volosyak | proceedings | m-sequence | CCA | class augmentation |
| Fodor, Cantürk, Heisenberg, and Volosyak | article | m-sequence | CCA | number of electrodes, montage |
| Fodor and Volosyak | proceedings | m-sequence | GAT-CNN | Deep learning, multi-head attention and graph neural net, electrode importance |
| Fodor and Volosyak | proceedings | m-sequence | SVM, CNN | imperceptible stimulation with 60 Hz carrier |
| Gomel, Torre Tresols, Cimarost, Cabrera Castillo, and Dehais | proceedings | burst codes | xDAWN and Riemannian | dry EEG, invisible stimuli |
| Gomel, Torre Tresols, Cimarost, Cabrera Castillo, and Dehais | proceedings | burst codes | xDAWN, LDA | confidence feedback |
| Guyonnet-Hencke et al. | article | Gold code | CCA | auditory hearing diagnostics |
| Kiser, Cantürk, and Volosyak | proceedings | m-sequence | CCA | authentication |
| Le, Fodor, Cantürk, and Volosyak | proceedings | m-sequence | CCA | authentication |
| Martín-Fernández, Martínez-Cagigal, Moreno-Calderón, Santamaría-Vázquez, and Hornero | article | m-sequence | CCA | stimulus opacity |
| Martín-Fernández, Martínez-Cagigal, Moreno-Calderón, Santamaría-Vázquez, Pascual-Roa, and Hornero | proceedings | m-sequence | CCA | realistic background and opacity |
| Martínez-Cagigal, Thielen, Hornero, and Desain | article | | | Editorial c-VEP |
| Martínez-Cagigal et al. | proceedings | m-sequence | CCA | asynchronous, early stopping |
| Martínez-Cagigal, Santamaría-Vázquez, Pérez-Velasco, Martín-Fernández, and Hornero | article | binary and sequence | CCA | binary and non-binary codes, spatial frequency, calibration time, comfort |
| Moreno-Calderón, Martínez-Cagigal, Martín-Fernández, Santamaría-Vázquez, Pascual-Roa, and Hornero | proceedings | m-sequence | CCA | mixed reality versus screen |

| | | | | |
|--|-------------|---|------------------------|---|
| Moreno-Calderón, Martínez-Cagigal, Martín-Fernández, Santamaría-Vázquez, and Hornero | article | m-sequence | CCA | mixed reality versus screen |
| Nair and Cecotti Santamaría-Vázquez et al. | preprint | m-sequence | CCA, Bayesian LDA, CNN | deep learning, siamese network |
| Scheppink, Cantürk, and Volosyak | proceedings | m-sequence | CCA | calibration-free, deep learning scene/object based stimulation |
| Tangermann et al. | article | | | review learning from small datasets |
| Thielen, Tangermann, Aarnoutse, Ramsey, and Vansteensel | article | modulated Gold codes | reconvolution, CCA | implanted, invasive, sEEG |
| Thielen | article | m-sequence, de Bruijn sequence, Golay sequence, Gold sequence, Gold code set, modulated | reconvolution, CCA | BCI inefficiency, performance predictors, binary stimulus sequences |
| Thielen | dataset | m-sequence, de Bruijn sequence, Golay sequence, Gold sequence, Gold code set, modulated | | binary stimulus sequences, EEG, ECG, SART, resting-state |
| Vodila, van Lohuizen, and Thielen | proceedings | modulated Gold codes | reconvolution CCA | gaze-independent, covert attention, alpha, ERP, c-VEP |
| Yu, Rao, Chen, Liu, and Jiang | article | Gold code | DBFENet, EEGNet, etc. | Robot arm, scene segmentation, deep learning |
| Zheng, Dong, Pei, Gao, and Wang | preprint | NBRS | FBCCA | calibration-free, 504 targets |

| 2024 | | | | |
|---|---|--|---|--|
| Author(s) | Type | Stimulus | Decoding | Notes |
| Ahmadi, Desain, and Thielen Cantürk and Volosyak Cabrera Castillos and Dehais Dehais, Castillos, Ladouce, and Clisson Fodor, Herschel, Cantürk, Heisenberg, and Volosyak C. Huang et al. E. Lai, Mai, Ji, Li, and Meng | article proceedings dataset article article article proceedings | modulated Gold codes m-sequence burst codes burst codes m-sequence white noise DIBS non-binary m-sequence | reconvolution, CCA CCA Riemannian, logistic regression CCA TRCA filterbank task related component analysis (FBTRCA), LSTM CCA | Bayesian dynamic stopping language model (ChatGPT) grating stimuli grating stimuli, dry EEG, comfort, eye-strain classification certainty feedback visual tracking asynchronous learning curve |
| Martínez-Cagigal, Álvaro Fernández-Rodríguez, Santamaría-Vázquez, Martín-Fernández, and Hornero Y. Miao et al. | article | white noise | TDCA, linear modeling, transfer learning EEG2Code, EEGNet, Shallow-ConvNet, DeepConvNet, ShallowNet CCA | minimal calibration, subject-to-subject transfer deep learning, transfer-learning, fine-tuning |
| Z. Miao, Meunier, Žák, and Grosse-Wentrup | proceedings | m-sequence | | |
| Narayanan, Ahmadi, Desain, and Thielen Qu et al. Scheppink, Ahmadi, Desain, Tangermann, and Thielen Shi et al. Sun et al. Thielen, Sosulski, and Tangermann Thielen, Farquhar, and Desain Velut, Chevallier, Corsi, and Dehais Zheng, Dong, et al. Zheng, Tian, et al. | proceedings article proceedings article article proceeding dataset proceedings article dataset | modulated Gold codes m-sequence modulated Gold codes white-noise m-sequence modulated Gold codes modulated Gold codes burst codes NBRS NBRS | CCA CCA TDCA TDCA reconvolution, CCA, UMM CNN, SPDNet, transfer learning FBCCA | gaze-independent, covert attention biometrics auditory, c-AEP maximum information rate small stimuli (0.5, 1, 2, 3 visual degrees) calibration-free, instantaneous, cumulative subject-to-subject transfer calibration-free, c-VEP versus SSVEP c-VEP versus SSVEP |

| 2023 | | | | | |
|--|---|--|---|---|--|
| Author(s) | Type | Stimulus | Decoding | Notes | |
| Ahmadi and Desain Cabrera Castillos, Darmet, and Dehais Cabrera Castillos Darmet, Ladouce, and Dehais Fernández-Rodríguez, Martínez-Cagigal, Santamaría-Vázquez, Ron-Angevin, and Hornero Henke et al. Z. Huang, Liao, Ou, Chen, and Zhang E. Lai, Mai, and Meng Martínez-Cagigal et al. | preprint article dataset proceedings article proceedings article proceedings article proceedings article proceedings article proceedings article proceedings article dataset proceedings article article | modulated Gold codes m-sequence, burst codes m-sequence, burst codes m-sequence m-sequence m-sequence m-sequence m-sequence, non-binary m-sequences non-binary m-sequence m-sequence non-binary m-sequence, m-sequence, APA sequence, Gold codes, Golay sequence, de Bruijn sequence m-sequence, Gold codes, Golay sequence, de Bruijn sequence, modulated codes modulated Gold codes m-sequence m-sequence | CCA CNN TRCA, EEG2Code, CNN CCA CCA Combined EEGNet FBTRCA, LSTM CCA CCA CCA EEG-inception reconvolution, CCA reconvolution, CCA CCA TRCA | Bayesian dynamic stopping Eyestrain spatial frequency Background music Biometrics fatigue Eyestrain, fatigue Dynamic stopping Games Simulated EEG Empirical EEG Gender VEPdgets, Dry EEG c-VEP versus SSVEP | |

2022

| 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 | article | quasi steady-state | CLAD | |
| Martínez-Cagigal et al. | article | | | 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 |

| 2020 | | | | | |
|---|--------------|--|-------------------|---|--|
| 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 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, Benda, and Volosyak | proceedings | m-sequence | FBCCA | Asynchronous, multi-session | |
| Z. Huang, Zheng, Wu, and Wang | article | m-sequence | transfer-learning | Subject-to-subject transfer | |
| Volosyak, Rezeika, Benda, Gembler, and Stawicki | article | m-sequence | CCA | SSVEP, SSMVEP, c-VEP, BCI illiteracy | |
| 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 | | |

| 2019 | | | | |
|--|--------------|----------------------------|----------------------------------|---|
| Author(s) | Type | Stimulus | Decoding | Notes |
| Ahmadi | dataset | modulated Gold codes | | |
| Ahmadi | dataset | modulated Gold codes | | |
| Ahmadi, Borhanazad, Tump, Farquhar, and Desain | proceedings | modulated Gold codes | CCA | Number of electrodes, montage |
| Ahmadi, Borhanazad, Tump, Farquhar, and Desain | article | modulated Gold codes | CCA | Number of electrodes, montage |
| 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) |
| Kadioğlu, Yıldız, Closas, Fried-Oken, and Erdoğmuş | article | m-sequence | Maximum likelihood | Color (green-red), fusion of c-VEP and eye 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 Mito | proceedings | | | frequency-hopping VEP |
| Nagel and Spüler | article | optimized random sequences | EEG2Code | Asynchronous, non-control state |
| Nagel and Spüler | article | random sequences | EEG2Code | c-VEP |
| Nagel | dissertation | | | For psychological experiments (button presses without behavior) |
| Peng and Huang | proceedings | m-sequence | sLDA | Fatigue |
| Shirzhiyan et al. | article | m-sequence, chaotic codes | CCA, spatiotemporal beam-forming | |
| Turi and Clerc | article | m-sequence | | 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 |

2018

| 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, Higer, and Erdoganmus | 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 | chapter | m-sequence | CCA, SVM | Color (green-blue), 40 Hz |
| Isaksen, Mohebbi, and Puthusserypady | article | m-sequence, Gold code, Barker code | correlation | |
| 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 | |

2016

| 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 Erdoganmus | 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 | |
| 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 | |

2015

| 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. | proceedings | Gold code | | correlation | Wheelchair |
| Nezamfar, Salehi, and Erdogan | proceedings | m-sequence | | maximum likelihood | Color (red-green, blue-yellow, black-white), presentation rate (30, 60, 110 Hz) |
| Sato and Washizawa | proceedings | m-sequence | | correlation | Automatic repeat request |
| Spüler | proceedings | m-sequence | | CCA, SVM | Windows applications |
| Thielen, van den Broek, Farquhar, and Desain | article | modulated Gold codes | | reconvolution, CCA | |
| Waytowich and Krusienski | 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 |

2013

| 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 | |
| Riechmann, Finke, and Ritter | proceedings | hierarchical codebook | SVM | |
| Spüler, Rosenstiel, and Bogdan | proceedings | m-sequence | OCSVM, | Robot, SSVEP versus c-VEP |
| Spüler, Rosenstiel, and Bogdan | proceedings | m-sequence | OCSVM | Color (red-green, black-white) |
| Spüler, Walter, Rosenstiel, and Bogdan | article | m-sequence | CCA, OCSVM | Unsupervised online calibration |
| Spüler, Walter, Rosenstiel, and Bogdan | | | | Unsupervised online calibration |
| | | | | c-VEP, ERN, P300, TMSEP, CCEP |

2012

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

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 |

2008

| 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 |

2007

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

2006

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

2002

| Author(s) | Type | Stimulus | Decoding | Notes |
|---------------------|-------------|-----------------|-----------------|--------------|
| Hanagata and Momose | proceedings | | | |

1992

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

1984

| Author(s) | Type | Stimulus | Decoding | Notes |
|------------------|-------------|-----------------|-----------------|--------------|
| Sutter | proceedings | | | |

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