

Fall Risk Reduction in Cognitively Impaired Elderly using Brain-Computer Interface: A Review

Ping-Chen Tsai

*Department of Electrical Engineering and Electronics
University of Liverpool
Liverpool, L66DH
Email: PingChen@liverpool.ac.uk*

Abstract—Decades of research and experimental studies have investigated strategies to reduce older people fall risk. The electrophysiological signal, electroencephalogram (EEG), is a strong basis of brain-computer interface (BCI) for clinical interpretation [1, 2]. EEG correlates with human intention [1].

Keywords— Brain Computer Interface, Cognitive function, fall risk assessment, motor rehabilitation

1. Introduction

Most BCI applications involves assistive technology that help the disabled people with motor movement or communication to improve their well beings[3]

1.1. Risk of fall assessment

Falls in ageing population is a rising public health problem globally. People aged over 65 years are entitled to higher risk of unintentional fall than younger adults and children, leading to fall-related injuries or even as severe as death [4]. Among fall-experienced elderly, those who residing at home, geriatric patients and inpatients adults are the majority [5]. In addition to fall-induced immobility that limits their daily activity, falls pose negative psychosocial effects on elderly such as dependence and social isolation [6]. Falling injuries also requires clinical interventions, bringing high financial burden to their family. To prevent fall of older people, World Health Organization (WHO) effective fall prevention strategies start by exploring variable risk factors and thus develop related healthcare training as well as safer environment to reduce the possibilities for falls [4].

Risk factors for falling varies, hence fall risks assessment tools are widely studied and used for more efficient diagnosis for falls. The accuracy and validity of risk assessment tools should be either tested or supported by evidence before putting in use by patients. As early as 1981, one of the first fall risk assessment tool was established based on incident report data [7]. Later, various assessment tool with quantitative indication of fall risk

were developed such as Downton scale, the St. Thomas Risk Assessment Tool in falling elderly inpatients (STRATIFY, effective for acute hospital [8])(lists of tools reference see [9]). In the last decade, most the emerging fall risk assessment tools were developed by retrospective case-control study to identify fall risk factors. These tools diverse across settings such as in the low-level care facility[10], geriatric care units[11] and acute care hospital [12, 13]. Other assessment tools are introduced by multi-factorial statistical approach that incorporate elders' gait performance and cognitive impairment to rate fall risks [14, 15]. The environment and human physical as well as mental status when fall prone to happen offers clues to specific factors for fall. Hence appropriate selection of risk assessment tool can lead to effective fall prevention intervention. **XXX number of the reviewed articles validated the tools, or supply literature evidence for the tools' reliability [16].**

Popular interventions for fall prevention strategies for older people includes living environmental modification, muscle strengthening training programme such as Tai-Chi exercises, and hiring assistive devices to treat impaired sensory functions [4]. (Do we need this section?)

1.2. Correlation between cognitive decline and risk fall (strong or loose)

Poorer gait in older people is associated with falls and limited mobility in daily life. Ageing process causes brain shrinking that damages brain structure resulting in cognitive impairments. Cognitively impaired older people have been shown to have lower capability in gait control, which leads to abnormalities of balance and inability to recover from loss of balance [19] hence falling [18, 20]. Cognitive declines cause neurocognitive disorders in the brain central nervous system with diverse clinical diagnoses such as traumatic brain injury, frontotemporal degeneration and Alzheimer disease, to name a few. The commonly shared characteristics of neurocognitive disorders on older people are declines in execution function (working memory, sensory integration, motor planning), social and cognition, perceptual-motor coordination, and other cognitive factors such as complex attention [21]. Execution functions, complex attention and perceptual-motor function are the three most relevant principal domains of cognitive function to fall risks due to their significant involvement in neural pathways of motor system [17].

Execution function:

Standing and walking is attention-demanding, high-level, goal-directed controlled task. Clinical and research study in the recent years has established appreciated evidence of correlation between execution functions and limb movements that produce gait [22]. The area of the prefrontal lobe and, in particular, the dorsolateral prefrontal cortex (DLPFC, Brodmann's area 9) and the cingulate cortex (e.g., the anterior cingulate: ACC, Brodmann's areas 24, 32) have been related to the cognitive aspects of EF [22]

Complex attention:

detail goes here....

Perceptual motor function:

detail goes here....

1.3. Using BCI to reduce cognitive decline:

Motor BCI translates neural signals from the motor areas of the cerebral cortex. Decoding algorithm is developed to interpret the motor-associated cortical activity.(to be justified!!!)

Investigate effective modalities to simulate the organization of brain functional networks and manipulate neural activities by signal processing. Bandpassing neural signals allows to keep a combinations of frequency which give access to certain brain activity components, the action potentials in cortical area. The manipulated activities by BCI could be modulated volitionally which is subject to neuralfeedback training strategy.

1.4. Using BCI to reduce risk of fall

2. Method

3. Search Strategy

4. Study selection

Literature selection criteria should prevent bias and errors as much as possible. In this section, state how you exclude bias.

5. Analysis

6. Quality assessment for the Selected Studies

7. Conclusion

The conclusion goes here.

Acknowledgments

The authors would like to thank...

References

- [1] K. J. Panoulas, L. J. Hadjileontiadis, and S. M. Panas, "Brain-computer interface (BCI): Types, processing perspectives and applications," in *Multimedia Services in Intelligent Environments*. Springer Berlin Heidelberg, Sep. 8, 2010, pp. 299–321.
- [2] R. Mane, T. Chouhan, and C. Guan, "BCI for stroke rehabilitation: Motor and beyond," *Journal of Neural Engineering*, vol. 17, no. 4, p. 041 001, Aug. 2020.
- [3] Kubler, Mushahwar, Hochberg, and Donoghue, "BCI meeting 2005—workshop on clinical issues and applications," *IEEE Transactions on Neural Systems and Rehabilitation Engineering*, vol. 14, no. 2, pp. 131–134, Jun. 2006.
- [4] "Fact sheet 344: Falls," World Health Organization, Tech. Rep., Jan. 2012.
- [5] D. Jung, S. Shin, and H. Kim, "A fall prevention guideline for older adults living in long-term care facilities," *International Nursing Review*, vol. 61, no. 4, pp. 525–533, Sep. 2014.
- [6] A. C. Scheffer, M. J. Schuurmans, N. van Dijk, T. van der Hooft, and S. E. de Rooij, "Fear of falling: Measurement strategy, prevalence, risk factors and consequences among older persons," *Age and Ageing*, vol. 37, no. 1, pp. 19–24, Jan. 2008.
- [7] R. Oulton, "Use of incidence report data in a system-wide quality assurance/risk management program," *QRB Qual Rev Bull*, vol. 7, no. 6, pp. 2–7, Jun. 1981, PMID: 6789277.
- [8] K. L. Perell, A. Nelson, R. L. Goldman, S. L. Luther, N. Prieto-Lewis, and L. Z. Rubenstein, "Fall risk assessment measures: An analytic review," *The Journals of Gerontology Series A: Biological Sciences and Medical Sciences*, vol. 56, no. 12, pp. M761–M766, Dec. 2001.
- [9] M. Aranda-Gallardo, J. M. Morales-Asencio, J. C. Canca-Sanchez, S. Barrero-Sojo, C. Perez-Jimenez, A. Morales-Fernandez, M. E. de Luna-Rodriguez, A. B. Moya-Suarez, and A. M. Mora-Banderas, "Instruments for assessing the risk of falls in acute hospitalized patients: A systematic review and meta-analysis," *BMC Health Services Research*, vol. 13, no. 1, Apr. 2013.
- [10] K. S. van Schooten, M. E. Taylor, J. C. Close, J. C. Davis, S. S. Paul, C. G. Canning, M. D. Latt, P. Hoang, N. A. Kochan, P. S. Sachdev, H. Brodaty, C. M. Dean, F. Hulzinga, S. R. Lord, and K. Delbaere, "Sensorimotor, cognitive, and affective functions contribute to the prediction of falls in old age and neurologic disorders: An observational study," *Archives of Physical Medicine and Rehabilitation*, 2020.

- [11] J. Michalcova, K. Vasut, M. Airaksinen, and K. Bielakova, "Inclusion of medication-related fall risk in fall risk assessment tool in geriatric care units," *BMC Geriatrics*, vol. 20, no. 1, Nov. 2020.
- [12] M.-H. Chiu, H.-D. Lee, H.-F. Hwang, S.-C. Wang, and M.-R. Lin, "Medication use and fall-risk assessment for falls in an acute care hospital," *Geriatrics & Gerontology International*, vol. 15, no. 7, pp. 856–863, Sep. 2014.
- [13] E. Aryee, S. L. James, G. M. Hunt, and H. F. Ryder, "Identifying protective and risk factors for injurious falls in patients hospitalized for acute care: A retrospective case-control study," *BMC Geriatrics*, vol. 17, no. 1, Nov. 2017.
- [14] L. H. J. Kikkert, M. H. de Groot, J. P. van Campen, J. H. Beijnen, T. Hortobágyi, N. Vuillerme, and C. C. J. Lamoth, "Gait dynamics to optimize fall risk assessment in geriatric patients admitted to an outpatient diagnostic clinic," *PLOS ONE*, vol. 12, no. 6, M. S. Kellermayer, Ed., e0178615, Jun. 2017.
- [15] M. E. Taylor, A. A. Butler, S. R. Lord, K. Delbaere, S. E. Kurrle, A. S. Mikolaizak, and J. C. Close, "Inaccurate judgement of reach is associated with slow reaction time, poor balance, impaired executive function and predicts prospective falls in older people with cognitive impairment," *Experimental Gerontology*, vol. 114, pp. 50–56, Dec. 2018.
- [16] H. Myers, "Hospital fall risk assessment tools: A critique of the literature," *International Journal of Nursing Practice*, vol. 9, no. 4, pp. 223–235, Aug. 2003.
- [17] W. Zhang, L.-F. Low, M. Schwenk, N. Mills, J. D. Gwynn, and L. Clemson, "Review of gait, cognition, and fall risks with implications for fall prevention in older adults with dementia," *Dementia and Geriatric Cognitive Disorders*, vol. 48, no. 1-2, pp. 17–29, 2019.
- [18] C. P. Carty, N. J. Cronin, D. Nicholson, G. A. Lichtwark, P. M. Mills, G. Kerr, A. G. Cresswell, and R. S. Barrett, "Reactive stepping behaviour in response to forward loss of balance predicts future falls in community-dwelling older adults," *Age and Ageing*, vol. 44, no. 1, pp. 109–115, Jun. 2014.
- [19] D. G. Thelen, L. A. Wojcik, A. B. Schultz, J. A. Ashton-Miller, and N. B. Alexander, "Age differences in using a rapid step to regain balance during a forward fall," *The Journals of Gerontology Series A: Biological Sciences and Medical Sciences*, vol. 52A, no. 1, pp. M8–M13, Jan. 1997.
- [20] K. L. Martin, L. Blizzard, A. G. Wood, V. Srikanth, R. Thomson, L. M. Sanders, and M. L. Callisaya, "Cognitive function, gait, and gait variability in older people: A population-based study," *The Journals of Gerontology: Series A*, vol. 68, no. 6, pp. 726–732, Oct. 2012.
- [21] P. S. Sachdev, D. Blacker, D. G. Blazer, M. Ganguli, D. V. Jeste, J. S. Paulsen, and R. C. Petersen, "Classifying neurocognitive disorders: The DSM-5 approach," *Nature Reviews Neurology*, vol. 10, no. 11, pp. 634–642, Sep. 2014.
- [22] G. Yogev-Seligmann, J. M. Hausdorff, and N. Giladi, "The role of executive function and attention in gait," *Movement Disorders*, vol. 23, no. 3, pp. 329–342, Dec. 2007.