Hypoxia and Cognitive Ability in Humans: A Systematic Review and Meta-Analysis

Abbreviated title: Hypoxia and Cognition

Daniel J. McKeown[1*], Douglas J. Angus[1], Ahmed A. Moustafa[1,2,3], Victor R. Schinazi[1,4]

[1]OXYNAV Laboratory, School of Psychology, Faculty of Society and Design, Bond University, Gold Coast, Queensland, 4229, Australia

[2] Department of Human Anatomy and Physiology, the Faculty of Health Sciences, University of Johannesburg, South Africa

[3] Centre for Data Analytics & School of Psychology, Bond University, Gold Coast, Queensland, Australia

[4]Future Health Technologies, Singapore-ETH Centre, Campus for Research Excellence and Technological Enterprise (CREATE), Singapore

* Corresponding author email address: dmckeown@bond.edu.au

ABSTRACT

This systematic and meta-analytical review examined how a reduction in oxygen availability to tissue (hypoxia) affects cognitive function. Hypoxia had a moderate-to-large detrimental effect on general cognitive ability and across domains, including memory, attention, executive function, processing speed, and psychomotor speed. Increased hypoxic severity was associated with greater declines in general cognitive ability and executive function, while longer duration of exposure was associated with greater declines in executive function and psychomotor speed. Participant age was a moderator for executive function and psychomotor speed, with older adults experiencing greater impairments. For executive function and psychomotor speed, the magnitude of these effects was less pronounced during intermittent and hypobaric exposures, potentially due to adaptive physiological mechanisms. While our models accounted for exposure characteristics and age of participants, substantial unexplained variance remained. These findings highlight hypoxia's impact on cognition and emphasize the need to investigate underlying neurophysiological mechanisms that may influence individual vulnerability.

INTRODUCTION

Overcoming reductions in oxygen availability to tissue (hypoxia) is a demanding physiological challenge requiring nervous, respiratory, renal, and cardiovascular adaptation. One of the earliest real-world accounts of cognitive impairment in a high-altitude setting comes from meteorologist James Glaisher and co-pilot Henry Coxwell during their 1862 hot air balloon ascent to 11,000 m above sea level. As altitude increased, Coxwell recounts losing motor control of his hands, while Glaisher lost visual acuity and eventually consciousness. Interestingly, the occurrence of these symptoms of "balloon sickness" intensified progressively during ascent. While anecdotal, this account suggests that hypoxia induces cognitive impairments that may depend on the duration and severity of exposure. Despite advances in our understanding of physiological adaptations to hypoxia, the impact of hypoxia on cognition remains unclear, particularly in relation to exposure severity and duration.

Cognition is a complex process, involving the integration of activity from multiple brain regions to interact and respond to a stimulus in the environment. Although valuable for understanding hypoxia's impact on cognition, real-world assessments in high-altitude environments cannot fully control for changes in barometric pressure and temperature. Considering this, laboratory-based assessments involving the manipulation of oxygen concentrations of inspired air (fraction of inspired oxygen; F_iO_2) allow for investigations of isolated hypoxic effects. In general, impairments in cognitive processes become evident from exposures equivalent to 2,500 to 3,000 m above sea level, but findings remain inconsistent, with hypoxia having a negligible-to-substantial impact on cognition [1, 2]. This inconsistency in findings may be due to the lack of standardization of tasks to assess cognitive processes, with multiple tasks assessing a cognitive process but with differing performance outcomes. Variation in the age of participants, severity, duration, and type (e.g., normobaric, hypobaric, intermittent) of exposure as well as intrinsic brain activity (e.g., periodic and aperiodic components of electroencephalographic [EEG] activity) may also contribute to inconsistent findings [3, 4].

Attentional tasks require the ability to sustain focus on a relevant stimulus while perceiving and appropriately ignoring irrelevant stimuli. Attention has been shown to be impaired in high-altitude exposures, with impairment increasing with the severity of exposure and attentional demand. Attentional performance is maintained at moderate altitudes (2,200 – 3,800 m above sea level) but reduced at extreme altitudes [5,800 m above sea level; 5, 6-8], in some cases persisting once returning to sea level [6]. Interestingly, attentional performance is more impacted by hypobaric exposures of hypoxia compared to normobaric exposures [8]. As attentional impairments are most evident during high-demand tasks [9], these findings suggest that attentional resources may be particularly vulnerable to hypoxia.

Tasks requiring retention and learning, such as those assessing memory-specific processes, have demonstrated similar results. Here, impairments in learning and retention of memory tasks can occur in high altitude and persist once returning to sea level [1 - 2 weeks post-summit; 10]. Studies on memory processing in pilots also demonstrate that increasing hypoxic severity and task complexity are associated with greater performance decrements [11, 12]. Alarmingly, individuals are

mostly unaware of such impairments while actively engaged in the task. Reductions in memory performance coincide with selective reductions in blood flow and delivery in the posterior cerebral artery, which is a primary supplier of oxygen to the hippocampus [8]. This suggests that the hippocampus may play an important role as the locus of hypoxia-induced memory impairment.

Complex tasks, such as performing two tasks simultaneously, place greater demands on executive functions to sustain goal-directed performance amid conflicting or distracting stimuli [13, 14]. In hypoxia, severe exposures have been shown to severely impact executive functions [15-18]. For example, when exposed to 30 – 60 minutes of acute hypoxia, participants exhibit greater reductions in performance during the naming (incongruent) component compared to the reading (congruent) component of the Stroop task [16-18]. Interestingly, increases in the Stroop interference effect are associated with decreasing peripheral blood oxygen saturation [S_pO_2; 18].

Task efficiency is also impacted by hypoxia and is related to the speed of processing and initiating a motor command [19], particularly when cognitive load is high [9, 20]. Hypoxia's impact on processing speed is also dependent on the type of stimulus being processed. Processing speed is reduced when presented with visual and auditory stimuli [9, 20, 21]. However, visual processing may be more impaired due to visual receptors inherently having a greater metabolic demand [22, 23] and involving a more diverse neural network [24]. Considering this, the type of cognitive task is an important factor when considering hypoxia's impact on speed of processing. Hypoxia also reduces an individual's ability to execute motor commands in response to a stimulus [25-28]. This is not surprising since exposure to hypoxia has regularly shown to reduce motor cortical and spinal output [29-32]. Psychomotor performance is reduced with increased reaction time during perceptual-motor tasks [25, 26, 33, 34]. However, this may be primarily due to reductions in processing speed [e.g., reduced visual sensitivity; 26, 33].

As human endeavours move into increasingly challenging environments, such as high altitude, a unique challenge is presented – how can complex cognitive processes operate with limited oxygen supply? Although 50 years of research [1, 2, 15, 35-39] has documented the impact hypoxia has on individual cognitive

processes, a comprehensive meta-analysis examining how factors such as severity, duration, type of exposure, participant age, and cognitive task type influence cognitive effects of hypoxia has not yet been conducted. In addition, previous systematic reviews have been limited by small study sizes [n = 7 - 18 included studies; 2, 35, 36], large variation in the time between onset of exposure and cognitive assessment [10 mins - 5 days; 35], the inclusion of concurrent exercise interventions [15], or a lack of meta-analytical approaches [1, 2, 36-39], all of which have hindered the standardization of hypoxic effects between studies [38].

Our systematic review and meta-analysis aims to provide a comprehensive synthesis of the impact hypoxia has on key cognitive domains including, memory, attention, executive function, processing speed, and psychomotor speed. In doing so, our review will: (1) provide an up-to-date assessment of the body of evidence, (2) address the question of how severity, duration, and type of exposure to hypoxia affects cognition compared to normal oxygen conditions in humans, and (3) what should be the focus of future research to better understand cognition in extreme environments.

RESULTS

Data Extraction

The two authors DJM and VRS agreed on 98.95% of the data items (*n* included= 38/857 data items, κ = .904) and disagreed on 1.05% (n= 9). From the disputed items, eight were ultimately included based on DJA's decision. In total, 46 data items were identified as being suitable for inclusion with measures consisting of author, year of publication, study design, intervention, sample characteristics, hypoxic duration and severity, outcomes, and main findings pertaining to cognitive processes (Table 1).

TABLE 1 HERE

Six data items were identified as having a substantial amount of risk of bias (Table 2). This risk was predominantly from Domain 1 (bias due to confounding effects), where studies either included a concomitant exercise protocol, did not include a control comparison condition (i.e., normoxia/sea level or low-lander

control), or had large participant age variability within the groups. Due to these potential risks of bias, these articles were not included in the meta-analyses.

TABLE 2 HERE

A Preferred Reporting Items for Systematic reviews and Meta-Analyses (PRISMA) flow diagram provides a summary of the inclusion process (Figure 1). A reference list of excluded data items is presented in Supplementary Material 1.

FIGURE 1 HERE

For the general cognitive ability and domain-specific models, significant coefficients are reported in text. For full model results including non-significant coefficients, see Supplementary Material 2. While funnel plots suggest the potential influence of publication bias in model results, the coefficients were similar between the random-effects and *trimfill*-corrected models (Supplementary Material 2). Considering this, the original random-effects and multivariate models are reported below.

General Cognitive Ability

A random-effects meta-analysis was conducted to examine the impact of hypoxia on general cognitive ability independent of moderators ($k = 151$). The average effect size was moderate and negative ($b = -.44$, 95% CI = -.55, -.33), and statistically significant ($z = -7.75$, $p < .0001$; Figure 2) indicating that hypoxia has a moderate adverse effect on general cognitive ability independent of severity or duration of exposure and other moderators. Substantial heterogeneity was observed in effect sizes across studies ($I^2 = 78.72\%$, $p < .0001$, $\tau^2 = 0.37$, $\tau = 0.61$, $H^2 = 4.70$), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

FIGURE 2 HERE

The multivariate model ($k = 141$) was a significant improvement over the reduced model (LRT = 40.70, $p < .0001$; full model: logLik = -172.28, AIC = 370.80, QE = 529.27; reduced model: logLik = -192.63, AIC = 389.34, QE = 661.83) indicating that the inclusion of the moderators explained more heterogeneity in the data, and a larger proportion of the variance (QM(7) = 29.19, $p < .001$).

Regarding the effects of each moderator, the severity of hypoxia was associated with greater impairment in general cognitive ability (b = 10.39, SE = 3.15, z = 3.30, p = .001, 95% CI = 4.22, 16.56).This result indicates that as the severity of hypoxia increases, so does the severity of impairment in general cognitive ability (Figure 3). The psychomotor speed domain was also a significant moderator with hypoxic impairment seen in general cognitive ability (b = -.48, SE = .12, z = -4.12, p < .0001, 95% CI = -.71, -.25), suggesting that psychomotor speed may be more adversely impacted by hypoxia compared to other cognitive domains. A moderate level of variance in effect sizes (σ^2 = 0.20, sqrt = .44) and a significant level of residual heterogeneity (QE(143) = 509.09, p < .0001) remained across studies, indicating that factors other than the included moderators contributed to the variance observed.

FIGURE 3 HERE

Attention

A random-effects meta-analysis was conducted to examine the impact of hypoxia on attention (k = 33). The average effect size was small and negative (b = -.25, 95% CI = -.43, -.06), and statistically significant (z = -2.61, p = .009; Figure 4) indicating that hypoxia has a small adverse effect on attention independent of severity or duration of exposure, and other moderators. Substantial heterogeneity was observed in effect sizes across studies (I^2 = 67.27%, p < .0001, τ^2 = 0.19, τ = 0.44, H^2 = 3.06), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

The inclusion of the multivariate model (k = 33) was not a significant improvement over the reduced model (LRT = 7.61, p = .18; full model: logLik = -23.28, AIC = 65.03, QE = 76.18; reduced model: logLik = -27.08, AIC = 58.56, QE = 98.19) indicating that the inclusion of the moderators did not explain more heterogeneity or a significant proportion of the variance in the data (QM(5) = 6.73, p = .24). A moderate level of variance in effect sizes (σ^2 = 0.14, sqrt = .37) and a significant level of residual heterogeneity (QE(27) = 76.18, p < .0001) remained across studies, indicating that factors other than the included moderators contributed to the variance observed.

Memory

A random-effects meta-analysis was conducted to examine the impact of hypoxia on the memory domain ($k = 44$). The average effect size was moderate and negative ($b = $ -.33, 95% CI = -.54, -.13), and statistically significant ($z = $ -3.17, $p = $ .002; Figure 4), indicating that hypoxia has a moderate adverse effect on memory independent of severity or duration of exposure, and other moderators. Substantial heterogeneity was observed in effect sizes across studies ($I^2 = 76.25\%$, $p < .0001$, $\tau^2 = 0.35$, $\tau = 0.59$, $H^2 = 4.21$), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

The inclusion of the multivariate model ($k = 44$) was a significant improvement over the reduced model (LRT = 38.79, $p < .001$; full model: logLik = -18.72, AIC = 88.86, QE = 57.45; reduced model: logLik = -38.11, AIC = 80.51, QE = 179.00) indicating that the inclusion of the moderators explained more heterogeneity and a significant proportion of the variance in the data (QM(14) = 30.16, $p = .007$).

Regarding the effects of each moderator, tasks that rely on episodic memory were significant ($b = $ -2.54, SE = .93, $z = $ -2.72, $p = .007$, 95% CI = -4.38, -.712), suggesting that episodic memory may be more adversely impacted by hypoxia compared to other memory tasks (e.g., working and associative memory). A moderate level of variance in effect sizes ($\sigma^2 = 0.27$, sqrt = .52) and a significant level of residual heterogeneity (QE(29) = 57.30, $p = .001$) remained across studies, indicating that factors other than the included moderators contributed to the variance observed.

Executive Function

A random-effects meta-analysis was conducted to examine the impact of hypoxia on the executive function domain ($k = 43$). The average effect size was large and negative ($b = $ -.67, 95% CI = -.91, -.43), and statistically significant ($z = $ -5.45, $p < .0001$; Figure 4) indicating that hypoxia has a large adverse effect on executive function independently of severity or duration of exposure, and other moderators. Substantial heterogeneity was observed in effect sizes across studies ($I^2 = 84.64\%$, $p < .0001$, $\tau^2 = .53$, $\tau = .73$, $H^2 = 6.51$), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

The inclusion of the multivariate model ($k = 43$) was a significant improvement over the reduced model (LRT = 109.26, $p < .0001$; full model: logLik = -8.55, AIC =

121.20, QE = 33.89; reduced model: logLik = -63.18, AIC = 130.66, QE = 236.87) indicating that the inclusion of the moderators explained more heterogeneity and a significant proportion of the variance in the data (QM(21) = 204.96, p < .0001).

Regarding the effects of each moderator, the severity of hypoxia was associated with greater impairment in executive function (b = 11.81, SE = 4.06, z = 2.91, p = .004, 95% CI = 3.86, 19.76; Figure 5), as well as longer durations of exposure (b = -.18, SE = .09, z = -2.07, p = .038, 95% CI = -.35, -.01), especially in normobaric exposures (b = -.57, SE = .29, z = -1.97, p = .048, 95% CI = -1.13, -.01). A range of executive function tasks were identified as being less impacted by hypoxia, including cognitive flexibility (b = 1.88, p < .0001), decision making (b = 1.81, p < .001), calculation error (b = 3.13, p < .0001), map orientation (b = 1.90, p < .0001), verbal fluency (b = 2.69, p < .0001), non-verbal fluency (b = 2.25, p < .001), proof reading (b = 1.42, p = .041), reasoning (b = 1.98, p < .0001), risk taking (b = 1.79, p < .0001), and spatial tracking (b = 1.45, p < .0001). Normobaric exposures of hypoxia had a greater impact on executive function compared to intermittent and hypobaric exposures (b = -.57, SE = .29, z = -1.97, p = .048, 95% CI = -1.13, -.01). Participant age was also a significant moderator (b = -.06, SE = .019, z = -2.86, p = .004, 95% CI = -.09, -.02), suggesting that hypoxia's effect on executive function increases with participant age. The estimated between-study variance was negligible (σ^2 < 0.01), and residual heterogeneity was not statistically significant (QE(21) = 31.91, p = .06), suggesting that effect sizes were highly consistent across studies, though unexplored factors may still contribute to some remaining variability.

Processing Speed

A random-effects meta-analysis was conducted to examine the impact of hypoxia on the processing speed domain (k = 18). The average effect size was moderate and negative (b = -.30, 95% CI = -.54, -.05), and statistically significant (z = -2.37, p = .018; Figure 4) indicating that hypoxia has a moderate adverse effect on processing speed independent of severity or duration of exposure, and other moderators. Substantial heterogeneity was observed in effect sizes across studies (I^2 = 64.43%, p < .0001, τ^2 = 0.17, τ = 0.42, H^2 = 2.81), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

The inclusion of the multivariate model (k = 18) was a significant improvement over the reduced model (LRT = 112.24, p < .0001; full model: logLik = -12.87, AIC = 111.19, QE = 35.85; reduced model: logLik = -63.17, AIC = 129.65, QE = 243.87), indicating that the inclusion of the moderators did explain more heterogeneity in the data. However, the inclusion of the moderators did not explain a significant proportion of the variance (QM(6) = 0.62, p = .99). There was still a moderate level of variance in effect sizes across studies (σ^2 = .46, sqrt = .68), indicating that the other factors outside of the included moderators contribute to the variance seen in effect sizes.

Psychomotor Speed

A random-effects meta-analysis was conducted to examine the impact of hypoxia on the psychomotor domain (k = 11). The average effect size was large and negative (b = -.72, 95% CI = -1.10, -.33), and statistically significant (z = -3.61, p < .001; Figure 4) indicating that hypoxia has a large negative effect on psychomotor speed independent of severity or duration of exposure, and other moderators. Substantial heterogeneity was observed in effect sizes across studies (I^2 = 76.59%, p < .0001, τ^2 = 0.31, τ = 0.56, H^2 = 4.27), suggesting that factors beyond the primary study variables contributed to the variability in effect sizes.

The inclusion of the multivariate model (k = 11) was not a significant improvement over the reduced model (LRT = .52, p = .97; full model: logLik = -8.13, AIC = 49.27, QE = 30.68; reduced model: logLik = -8.39, AIC = 22.29, QE = 45.86) indicating that the inclusion of the moderators did not explained more heterogeneity or a larger proportion of the variance in the data (QM(7) = 6.52, p = .48).

Regarding the effects of each moderator, duration of hypoxia was associated with greater impairment in psychomotor speed (b = -.17, SE = .07, z = -2.43, p = .015, 95% CI = -.30, -.03; Figure 5), indicating that the impact of hypoxia on psychomotor speed increases as the length of exposure increased. The type of hypoxic stimulus was also a significant moderator where effects were greater in hypobaric (b = -5.99, SE = 2.67, z = -2.25, p = .025, 95% CI = -11.22, -.76) and normobaric hypoxia (b = -8.51, SE = 3.79, z = -2.25, p = .025, 95% CI = -15.94, -1.09) compared to intermittent hypoxia. Participant age was a significant moderator (b = -.23, SE = .11, z = -2.02, p = .043, 95% CI = -.45, -.01), suggesting that

hypoxia's effect on psychomotor speed increases with participant age. A negligible level of variance in effect sizes across studies (σ^2 = .23, sqrt = .48) and a significant level of residual heterogeneity (QE(3) = 8.52, *p* = .036) remained, indicating that factors other than the included moderators contributed to the variance observed.

FIGURE 4 & 5 HERE

DISCUSSION

In this systematic and meta-analytical review, we provide robust evidence that hypoxia impairs general cognitive ability, alongside domain-specific functions including attention, memory, executive function, processing speed, and psychomotor speed. By including exposure severity, duration, type, participant age, and task characteristics, our models reveal that the impact of hypoxia on cognition is complex, not uniform, and dependent on the combination of moderating factors.

Hypoxia impairs both general and domain-specific cognitive function

We categorised general cognitive ability by pooling the performance of memory, attention, executive function, processing and psychomotor speed processes. We then constructed domains based on commonalities in cognitive tasks assessed in each study. Overall, hypoxia had a moderate negative effect on general cognitive ability, indicating that oxygen availability is integral for maintaining general cognitive function. However, not all cognitive domains had equal impairment. The psychomotor speed domain was more negatively impacted by hypoxia compared to other domains of the general cognitive ability model. This was also reflected in the domain-specific models, with psychomotor speed having the largest pooled negative effect of hypoxia.

Psychomotor speed reflects the efficiency of the sensorimotor system and is typically characterised by the speed at which one can efficiently elicit a motor response to a perceptual stimulus, resulting in voluntary movement [40]. Reductions in psychomotor processes typically begin from mild severities of hypoxia, primarily due to motor speed and accuracy reductions [37]. At moderate-to-severe altitudes (~4,000 m – 6,000 m above sea level), hypoxia has been shown to impair both perceptual and motor components of the sensorimotor system, including reduced

integration of sensory information [21, 41], decreased physiological tremor [42], diminished voluntary output from the motor cortex [31, 43], and lowered excitability of both cortical and spinal motoneurons [29, 30, 44]. However, whether reductions in psychomotor performance is due to reductions in processing speed or reaction time to initiate a motor response remains unclear. Given that processing speed was negatively affected in our model, but H-reflex and M-wave latencies (known markers of spinal and peripheral nerve conduction) typically remain unchanged under hypoxic conditions [45, 46], it is plausible that hypoxia has a stronger impact on speed of processing rather than speed of motor output.

More complex cognitive domains were impaired by hypoxia, such as attention, memory, and executive function. However, not all tasks within these domains were impaired to the same extent. Of all cognitive domains, attention was the least affected by hypoxia, even in models accounting for various moderators. This effect may be due to the limited characteristics of tasks used to assess attention. Here, most tasks (e.g., Stroop, go-no-go, and flanker task) focus on response inhibition, which may reflect only one aspect of attention.

In the memory domain, episodic memory was more impaired by hypoxia compared to working and associative memory. Episodic memory typically requires individuals to recall details about previous events such as recalling a set of words, events, or pictures [47]. Episodic memory has long been recognised to be dependent on the integrity of the hippocampus [48, 49], with hippocampal lesion studies demonstrating impairments in episodic memory but not visual pattern matching or schematic spatial knowledge [50-52]. The hippocampus is known to be particularly susceptible to hypoxia due to its high metabolic activity and dense synaptic connectivity [53-58]. Here, investigations into the impact hypoxia has on animal models provide insight into the potential underlying mechanisms responsible for the decrements seen in hippocampal function when exposed to hypoxia. Specifically, when exposed to transient continuous oxygen deprivation, the CA3-CA1 synapses and NMDA receptors of CA1 neurons of the rodent's hippocampus exhibit long-term potentiation indicative of pathological hyperexcitability [59-63], disruptions to modulatory neurotransmitter activity [e.g., reduced ACh production; 64, 65], and altered firing of excitatory and inhibitory neurons [66]. These intrinsic neuronal adaptations have a detrimental impact on key cognitive processes. Indeed, exposure

to hypoxia is known to impair memory in mice when performing maze tasks [64, 67, 68] and passive avoidance tasks [69], which may recover after adaptation has occurred [70]. As episodic memory is reliant on CA1 and CA3 hippocampal neurons, it is unsurprising that memory, and particularly episodic memory, is susceptible to hypoxia-induced impairments.

Executive function was also negatively impacted by hypoxia, but more pronounced in some tasks than others. This likely reflects the heterogeneous nature of the tasks used to assess executive functions. For example, verbal fluency tasks involve language-based processes while non-verbal fluency relies on visuospatial skills, both of which are categorised as executive functions. Therefore, accounting for heterogeneity in tasks may help explain some of the variance remaining in the moderator model. Executive function tasks are strongly reliant on prefrontal cortical activity, which is sensitive to hypoxia [71]. However, tasks that were less affected included cognitive flexibility, verbal and non-verbal fluency, decision making, reasoning, risk taking, and map orientation performance. These less effected tasks may rely on more distributed neural networks outside of the prefrontal cortex, such as the parietal and temporal cortices [13, 14, 72]. Considering this, performance preservation in these tasks may reflect compensation from other cortical regions (less of a dependence on prefrontal cortical activity), rather than the resilience of executive function.

Type, severity, and duration modulate hypoxia's impact on cognition

The meta-analyses included studies performed in normobaric, hypobaric, and intermittent hypoxia. Variations in physiological adaptations across these types of exposure may account for some of the unexplained variance. For instance, when hypoxia is encountered in hypobaric environments, reductions in atmospheric pressure and temperature typically occur. This can alter respiratory, haematological, circulatory, and metabolic adaptations [73]. Similarly, intermittent exposure to hypoxia can produce a hormetic response [74]. This enhances pulmonary function, lowers blood pressure, mitigates inflammation, and increases cardiac vagal activity, leading to improved cognition, particularly in older populations [36, 74]. Further consideration needs to be given to other environmental factors outside of isolated hypoxic effects that could contribute to alterations in cognitive performance. For

example, cold stress can impact executive functioning and memory performance but is dependent on task complexity and attentional demand [2]. It is likely that adaptive processes that occur with hypobaric and intermittent exposures also mitigated hypoxia's impact on some cognitive domains.

Severity of hypoxia exposure was associated with lower performance in general cognitive ability and the executive function domain, whereby performance was reduced to a greater extent at more severe exposures. As physiological adaptations to hypoxia have an upper limit of mitigation [75], it is likely that at extreme altitudes impairment in cognitive function is unavoidable. However, as prolonged exposures to hypoxia result in more long-term adaptations [73, 76], paradoxically, remaining at a moderate level of altitude over a prolonged duration may also lessen the impact of hypoxia on some cognitive processes. This may explain why duration of exposure was associated with worse performance in the executive function, and psychomotor domains but not the general cognitive ability, nor attention, memory, and processing speed domains. Why severity and duration moderator effects were seen in some domains but not others remains unclear.

Limitations

While a substantial amount of variance was explained by the inclusion of domain or task, participant age, and severity, duration, and type of hypoxia, a moderate amount of unexplained variance remained for all models. At the task level, previous reviews have identified the challenge of comparing between tasks to assess a common cognitive process [38]. We aimed to account for this variance by assessing standardised effects while controlling for the weighting of each effect across the studies included. It is well known that aging is associated with cognitive decline in a range of cognitive processes [77, 78]. This was reflected in our meta-analyses as increasing participant age was associated with greater hypoxic impairment in executive function and psychomotor speed domains. Greater cognitive impairment in older individuals during hypoxia is likely due to reduced ability to effectively compensate for cognitive deficits, which are already associated with age-related cognitive decline [79]. For example, oxygen availability to active neural tissue is typically reduced by greater cerebrovascular resistance associated with aging, which reduces oxygen supply-demand and increases relative cognitive load [80].

Individual alpha peak frequency of the EEG is reflective of changes in cognitive load [81] and has been linked to variation in processing speed [81], memory [82], and attention [83]. Aperiodic activity, also referred to the level of 'neural noise' in the brain, has been linked to better cognitive performance in young and older populations [77, 84, 85]. Individuals who have greater exponents (reflective of reduced excitation:inhibition balance) and offsets (greater neural spiking) also exhibit better cognitive performance during processing speed [84], memory [86-88], and executive function [89] tasks. EEG research in hypoxia has, however, primarily focused on periodic activity [for review see 4]. Hypoxia has shown to left shift peak alpha frequency, indicated by a reduction in alpha power and an increase in theta power in canonical frequency bands [90] but see [91]. Hypobaric exposure to hypoxia reduces the exponent, reflective of a 'noisier' brain akin with non-pathological alterations in brain activity with age [3]. Considering this, future EEG research needs to focus on uncovering how hypoxia impacts aperiodic EEG activity and how this is associated with variation in hypoxia-induced reductions in cognitive performance.

Conclusion

We provide meta-analytical evidence that exposure to hypoxia has a moderate-to-large negative effect on general and domain-specific cognitive function. These effects are moderated by the severity or duration of exposure, depending on the cognitive task. Intermittent and hypobaric exposures of hypoxia may have a lower impact cognitive function due to protective adaptative physiological processes leading to increased hypoxia tolerability. Aging was associated with greater impairment in hypoxia in executive function and psychomotor speed. Although these findings align with the evidence of previous reviews indicating that hypoxia has a net negative effect on cognitive performance, a moderate level of unexplained variance remained due to moderating factors outside of those included in the meta-analysis. This variance may be explained by differences in underlying intrinsic brain activity that has previously been associated with performance differences in cognitive tasks. Considering this, future research should aim to uncover how these neural correlates change with hypoxic exposure.

METHOD

This systematic and meta-analytical review aligned with the 2020 PRISMA guidelines [92].

Eligibility Criteria

Eligibility criteria conformed to the population, intervention, comparator, outcome, setting, types (PICOST) framework. Data items included all relevant randomised-controlled trials, observational studies, longitudinal studies, cohort studies, case-control studies, before-after studies, and cross-sectional studies or surveys involving healthy humans exposed to either environmental (high altitude) or simulated (laboratory-based) hypoxia. Studies were included if they used normal oxygen, sea-level conditions, or low-landers as a comparator and measured outcomes related to memory, attention, executive function, processing speed, or psychomotor speed. Data items with hypoxic exposure due to pathology (e.g., chronic obstructive pulmonary disease), with interventions that were pharmaceutical or exercise-based, which did not have suitable comparators, or conducted in a hospital-based setting, were excluded from the review.

Data Sources & Search Strategy

The Evidence Research Accelerator [TERA; 93] suite was used to aid in the data source and search strategy, and data screening and selection process. The SearchRefiner tool [94] of TERA was used to aid in data sourcing and developing the search strategy. A literature search querying SCOPUS, PubMed, Medline, and Cochrane Library from inception to 6th January 2025 was performed. Search queries consisted of the terms "Hypoxia" or "Altitude", or "Hypobaric", and "Cognition" or "Attention" or "Memory" or "Executive Function" or "Processing Speed" or "Psychomotor Speed" with each of their combinations searched in each respective database. Thirteen seed PMIDs were used as "relevant citations" to aid in refining the literature search (Supplementary Material 1). In total, 858 data items were identified in the literature search. Using the Deduplicator tool [95] of TERA, one data item was identified as 'likely being a duplicate' and was removed [96], bringing the total number of data items to 857.

Selection & Data Collection Process

Authors DJM and VRS independently screened the citation search, title/abstract, full-texts, and trial registries for inclusion of data items against the inclusion and exclusion criteria. Any disagreements regarding data item inclusion were resolved by referring to the third author, DJA.

Risk of Bias

The ROBINS-I V2 assessment tool for non-randomised interventional studies was used to assess risk of bias within and between data items. Bias was assessed against seven domains: bias due to confounding (Domain 1), bias in the classification of interventions (Domain 2), bias in the selection of participants (Domain 3), bias due to deviation from intended intervention (Domain 4), bias due to missing data (Domain 5), bias arising from measurement of the outcome (Domain 6), bias in the selection of the reported result (Domain 7). For Domain 1, confounding factors consisted of participant-related, task-related, environment-related, and methodology-related factors (Table 3). Risk of bias assessments were performed by authors DJM and VRS. Any discrepancies between studies were evaluated by the third author, DJA. A list of data items identified as being of moderate–severe risk during the risk of bias assessment is presented in Table 2 (n = 6).

TABLE 3 HERE

Meta-analysis

Meta-analyses were performed in *R* (v 4.3.1) using the *metafor* package (v 4.8-0). A general cognitive ability construct was created, including all cognitive tasks, measured in the data items. Cognitive domain constructs for attention, memory, executive function, processing speed, and psychomotor speed were created from data items categorized in specified cognitive tasks (Table 4).

TABLE 4 HERE

Firstly, for the general cognitive ability and individual cognitive domains, a reduced model was calculated, which consisted of including study ID as a random-effect without including moderators. Following this, a full model was calculated by performing a multivariate mixed-effects model, which included moderators. Previous literature suggests that the impact of hypoxia on cognition may be dependent on the severity (F_iO_2 level) and duration (minutes of exposure) of the hypoxic stimulus.

Accordingly, severity and duration of hypoxia were included as moderators in the multivariate model. For studies comparing highlanders to lowlanders, we treated the highlanders' exposure duration as chronic and used the mean age of participants (in minutes) as a proxy for duration. Due to large variability in duration of exposure and skewness between studies, duration of exposure was log-transformed. Additionally, the type (normobaric hypoxia, hypoxia in barometric pressure equivalent to sea level; hypobaric hypoxia, hypoxia in barometric pressure lower than sea level; intermittent hypoxia, cyclic exposures to normobaric hypoxia typically in one minute intervals) of stimulus, cognitive domain for the cognitive ability model, cognitive task for the individual domain models, and participant age, were added as additional moderators to identify if the effects of hypoxia are type-, domain/task-, and age-specific. The reduced and full models were compared against each other using an ANOVA. In both models, Hedge's g (standardised mean difference) was calculated to quantify the effect size of each cognitive outcome measure, and 95% confidence intervals were used for variance in effect size.

Forest plots were used to visualise the relationships between effect sizes and the overall random effect. Funnel plots were used to inspect publication bias (Figure 6 and Supplementary Material 2). The *trimfill* function of the *metafor R* package was used to correct effect sizes for potential confounding of publication bias (Supplementary Material 2). Positive Hedge's g scores indicate better cognitive performance in hypoxic conditions, while negative Hedge's g scores indicate worse cognitive performance in hypoxic conditions. To visualise the relationship between hypoxia severity and cognitive ability decline, a meta-regression based on the multivariate mixed effects model was fit to a predicted dataset where severity was varied across its observed range derived from the 152 effect sizes. The duration of exposure and participant age were held constant as the average means in the general cognitive ability model (duration = 11 hours, participant age = 30 years).

FIGURE 6 HERE

DATA AVAILABILITY

Collated data and analysis code used in this manuscript can be found at

https://github.com/MindSpaceLab/Hypoxia_Cognition_Review

COMPETING INTERESTS AND FUNDING

The authors declare that no competing interests exist, and no funding was obtained for this work.

AUTHOR CONTRIBUTIONS

Conceptualization: D.J.M., D.J.A, and V.R.S.; Data curation: D.J.M., D.J.A, and V.R.S.; Formal analysis: D.J.M., D.J.A, A.A.M., and V.R.S; Project administration: D.J.M.; Visualization: D.J.M.; Writing - original draft: D.J.M., D.J.A, A.A.M., and V.R.S.; Writing - review & editing: D.J.M., D.J.A, A.A.M., and V.R.S.

Table 1. Characteristics of included data items (n = 46).

Author & Year	Design	Intervention	Sample characteristics	Duration	Severity	Outcome measures	Main cognitive findings
Aebi et al., 2020 [97]	Non-randomised cross-over	NN, HH, NH, HN	16 healthy pilot trainees, 26 ± 4 yr	30 min	NN, HN = 440m HH, NH = 5500m	KLT KSS	↑ KLT error in NH & HH ↓Calculations in HH
Ando et al., 2023 [8]	Non-randomised cross-over	NN, HH	12 males, 23 ± 2 yr	7 hr	NN = 0.21 F_iO_2 HH = 0.12 F_iO_2	Go/No-Go Memory task accuracy	↓ Memory task accuracy in HH ↔ Go/No-Go RT & accuracy
Asmaro et al., 2013 [11]	Repeated measures within subject	NN, HH	35 aviation industry employees, 19-69 yr	30 min	NN = SL HH1 = 5335m HH2 = 7620m	ST DST Forward & Back TMT A & B	↓ Performance during ST and DST Forward & Backward, & TMT A&B in HH
Bailey et al., 2019 [98]	Cross-sectional observational	-	23 male highlander 17 male lowlanders	-	Highlanders = 3640m Lowlanders = 80m	RAVL RDF & RDB TMT A & B DSST GPDT	↓ RAVLT, RDB, RDF, TMT B, DSST, ST in highlanders compared to lowlanders
Bakharev et al., 1981 [99]	Repeated measures within subject	LA, HA	7 young men	8 days	HA = 3400-3600m	Visual, short term, long term, operative, figurative, associative, involuntary, and auditory memory (RAVLT-like)	↓ Memory at day 2 Those of central Asia (highlanders) had less severe reductions compared to Pskov Region (lowlanders).
Blogg et al., 2006 [100]	RCT	NN, HH	13 subjects, 29.5 ± 6.4	10 min	21kPa, 15kPa, 10kPa P_iO_2	FCRT Mannequin test Finger-tapping test Medium-tapping test	↔ Psychomotor performance at 15kPa ↑ FCRT and medium-tapping test error at 10kPa. Although participants perceived being more focused
Bloomfield et al., 2023 [101]	RCT	NN, PCH, ICH	12 adults, 26 ± 6.2 yr	30 min	10% oxygen gas (5500m)	CNSVS test battery	↓ Neurocognitive index score, cognitive flexibility, complex attention, composite memory, and visual memory in PCH ↓ Neurocognitive index score, composite memory, psychomotor

							speed and ↑ RT in ICH
Bonnon et al., 1995 [102]	Comparative	NN, HH	6 subject experimental and unspecified control group	60 hr	4383m	Cognitive-motor task	↑ Cognitive-motor performance in control group ↔ Performance in experimental group comparable once at 60 hrs.
Bonnon et al., 1999 [103]	Comparative	NN, HH	7 subject experimental and unspecified control group	17 days	9 days at 3500-4400m, 8 days climbing 500m/day to 5400m	Cognitive motor task	↔ Difference between experimental and control group.
Bouquet et al., 1999 [28]	Comparative	NN, HH	8 male climbers, 24 – 37 yr 10 controls, 22 – 40 yr	31 days	Simulated climb to 8848m	Purdue pegboard, recognition task	↓ Purdue pegboard performance > 8000m ↓Recognition performance in from 5500m ↓Performance in complex task vs simpler task
Champod et al., 2013 [104]	Non-randomised cross-over	NN, IH	17 healthy males	6 hrs of 1min on/1 min off	-	*n*-back task	↑RT during spatial but not verbal n-back task in IH across all levels of difficulty
Chroboczek et al., 2023 [105]	Non-randomised cross-over	NN, NH	17 healthy males	15 min	NH = 0.13 F_iO_2	ST	↔ ST reading and naming interference
Crow et al., 1971 [106]	Repeated measures within subject	HH	86 medical students, 19 – 30 years. Organised into groups of 4 or 6	30 min	HH = 609m, 2438m, or 3656m	Required to memorize sequences of 6 digits at 1Hz (n-back-like task) to assess short-term memory	↔ error rate between conditions
Davis et al., 2015 [27]	Comparative	HH	132 residents into three groups based on altitude of residence (0-1500m, 1501-3000m, >3000m)	-	Testing occurred at 4860m	Go/no-go test score VFT Hand movement test	↔ VFT ↓ Go/no-go in >3000m group ↓ Correct hand movements in >3000m and 1500-3000m group compared to 1500m group
Falla et al.,	RCT	NN, HH	48 pre-hospital emergency service	30 min	NN = 200m HH = 3000m or 5000m	PVT BART	↓ PVT at 5000m compared to 200m, no difference in other test Participants

2024a *[107]*			personnel			DSST VAS	did not detect ↓ PVT on VAS
Falla et al., 2024b [108]	RCT	HH	36 pre-hospital emergency service personnel	30 min	Two trials at 4000m, one with supplement oxygen and one without (control)	PVT DSST *n*-back GPDT	↑ PVT with oxygen supplementation ↑ 0.37ms in PVT each % reduction in S_pO_2.
Gerard et al., 2000 [109]	Randomised cross-over	NH	23 healthy adults	2 hr	NH = 5000m with and without 6% O_2 enrichment	Story Memory TMT A&B DSST Finger Tapping Digit Vigilance GPDT PASA Matrix Reasoning KSS Probed Memory Recall PVT CPT CTT PERS VAS	↓ RT in all tasks with O_2 enrichment ↑ Hand-eye coordination, and mood with O_2 enrichment.
Guo et al., 2017 [110]	Observation	NN, HH	163 healthy adults	24 hr	HH = 3700m	RT Attention DSST DST Pursuit aiming	↔ Cognitive measures between those with headache and those without at HA
Hill et al., 2014 [111]	Comparative	LA, HA	229 in total: children (4-10 yr, LA = 44, HA = 52), adolescents (13-16 yr, LA = 23, HA = 25), young adults (25-40yr, LA = 20, HA = 20), older adults (>50 years, LA = 22, HA = 23)	-	LA = 500m HA = 3700m	Cognitive Drug Research battery	↓ Processing speed, attention in HA participants compared to LA.
Hohenauer et al., 2022 [112]	Randomised cross-over	NN, NH, HH	20 healthy participants, 27.6 yr	NH = 15 min HH1 = 1hr HH2 =	NN = 550m NH = .144 F_iO_2 HH = 2980m	ANAM	↔ RT, logical relations, math processing, switching, and *n* back tests in any condition

Study	Design	Conditions	Sample	Duration	Altitude/Condition	Tasks	Results
				24 hr			
Keeler et al., 2023 [113]	Repeated measures within subject	NN, NH	21 healthy adults, 27 ± 5 yr	90 min	NN = 0.206 F_iO_2 NH = 0.14 F_iO_2	Gridshot Capacity Multitracker	↔ no difference in any gamified task.
Koller et al., 1993 [114]	Comparative	NH	10 acclimatised mountaineers and 11 non-acclimatised controls	-	NH = 6000m	Attentiveness	↓ Attentiveness in lowlanders compared to highlanders >4000m
Latshang et al., 2013 [115]	Randomised cross-over	NN, HH	51 healthy men, 24 yr	2 days at each HH level	NN = 490m HH1 = 1630m HH2 = 2590m	PVT, DASS, *n*-back tasks	↔ All cognitive tasks across conditions.
Legg et al., 2016 [116]	Randomised cross-over	NN, HH	36 air force personnel	30 min	HH1 = 8000ft HH2 = 12000ft	ANAM	↔ All cognitive tasks across conditions
Malle et al., 2013 [12]	Comparative	NN, HH	28 pilots in HH and 29 control pilots	90 min	NN = SL HH = 10000m	PASA	↓ Working memory in HH group compared to NN
Merz et al., 2013 [117]	Comparative	NN, HH	32 mountaineers, separated into two groups summitting at different rates	-	NN = SL HH1 = 4497m HH2 = 5533m HH3 = 6265m	LB VFT RFFT CT	↔ All cognitive tasks across conditions
Niedermeier et al., 2017 [118]	RCT	NN, IH, HH	52 participants into a sham or IH pre-acclimatisation group	NN, IH = 1hr HH = 12 hr	NN = 0.2 F_iO_2 IH = 0.13 F_iO_2 HH = 0.13 F_iO_2	Game of Dice Task	↓ Risky decision making in HH for IH group compared to sham
Noble et al., 1993 [119]	Comparative	NN, NH	2 groups of 12 male subjects	30 min	NN = SL NH = 78% S_pO_2	TMT DSST LRT RT	↓ Performance of TMT and no learning effect in NH group compared to sham. ↑ RT
Parker et al., 2017 [120]	Comparative	NN, NH	10 orthopaedic surgeons into either NN or NH group	45 min	NN = SL NH = 0.14 F_iO_2	VWMC	↔ VWMC tasks across conditions
Pighin et al., 2014 [121]	Randomised cross-over	NN, NH	26 university students, 23 ± 6.8 yr	60 min	NN = 0.21 F_iO_2 NH = 0.14 F_iO_2	Decision-making task	↓ Loss aversion in NH compared to NN.

Study	Study type	Conditions	Participants	Duration	Altitude details	Cognitive test	Results
Quackenbush et al., 2016 [122]	Comparative	NN, HH	1 group performed VIHT 1 group performed sham	30 min	HH1 = 3048m HH2 = 3658m	ST SDMT DST	↑ ST and SDMT performance for VIHT group compared to sham
Ramirez-delaCruz et al., 2025 [123]	Randomised cross-over	NN, NH	19 participants, 23.7 ± 3.9 yr	45 min	NN = SL NH1=0.15 F_iO_2 NH2=0.13 F_iO_2 NH3=0.11 F_iO_2	Memory task, Go/No-go task (sustained attention), Ericksen Flanker test, BAT (anticipation), RT	↑ sustained attention at NH1, ↑ anticipation at NH1 & NH3, ↑ RT at NH3 ↔ no difference across conditions
Rice et al., 2019 [124]	Repeated measures within subject	NN, NH	60 aviation students	5 min/level	NN = SL NH1 = 2438m NH2 = 3658m NH3 = 4877m NH4 = 6096m	Visual Sequence Comparison X-Plane	↓ Visual Sequency Comparison & X-Plane performance in NH compared to NN
Schega et al., 2016 [125]	RCT	NN, IH	34 participants, 60 – 75 yr	90 min	IH = 90-85% S_pO_2	ST	↑ ST performance in IH group
Schega et al., 2013 [126]	RCT	NN, IH	34 participants, 60 – 75 yr	90 min	IH = 90-85% S_pO_2	ZVT	↑ ZVT performance in IH group
Schlaepfer et al., 1992 [127]	Repeated measures within subject	NN, HH, NH	10 healthy subjects	-	HH = 3450m NH = 0.145 F_iO_2	Read briefly displayed letters	↓ time needed to read letters in HH and NH
Shukitt-Hale et al., 1998 [128]	Repeated measures within subject	NN, HH	23 army personnel, 18-29 yr	4.5 hr	NN = 550m HH1 = 4200m HH2 = 4700m	Battery of cognitive tasks	↓ Performance in battery of cognitive tasks only after a few hours of exposure in a grade dependent fashion
Tesler et al., 2015 [129]	Randomised cross-over	NN, HH	26 participants, 26.9 yr	2 days	NN = 490m HH1 = 1630m HH2 = 2590m	Visio-motor learning task	↓ Sleep dependent memory performance in HH conditions
Thomas et al., 2007	Prospective	HH	11 subjects, 27 ± 1.5 yr	9 hr	HH1 = 7700ft HH2 = 10000ft HH3 = 13000ft	PVT *n*-back	↔ No difference in PVT & *n*-back

[130]							
Turner et al., 2015a [131]	Randomised match-paired	NN, NH	22 adults, 23 yr	50 min	NN = SL NH = 0.1 F_iO_2	CNS Vital Signs	↓ Performance in CNS Vital Signs
Turner et al., 2015b [132]	Randomised match-paired	NN, NH	15 adults, 21-55 yr	90 min	NN = SL NH = 0.1 F_iO_2	CNS Vital Signs	↓ Performance in CNS Vital Signs
van der Post et al., 2002 [133]	Randomised cross-over	NN, NH	12 healthy subjects	2 hr	NN = 97% S_pO_2 HH1 = 90% S_pO_2 HH2 = 80% S_pO_2	Memory task RT	↑ RT in HH2 compared to HH1 and NN. ↑ Cognitive performance in HH1 compared to NN and HH2
Van Dorp et al., 2007 [134]	RCT	NH	22 healthy subjects	45 min	Sham hypoxia = 103 mmHg PetO2 without inspired CO_2 Isocapnic hypoxia = 40mmHg with 0.4 mmHg clamped CO_2 Poikilocapnic hypoxia = 40mmHg without CO_2	PVT and MAT-bat	↓ Performance in PVT and MAT-bat
Wu et al., 2023 [135]	RCT	NN HH	82 males 20.76 ± 1.6, into control or a RIPC group	6 days	HH = 3680m	WHO-NCTB	↑ Spatial memory and digit symbol performance in RIPC group compared to controls ↑ Spatial memory performance in RIPC group compared to RIPC at baseline
Yan et al., 2011a [136]	Comparative	NN HH	2 groups: 1 born at SL, the other in HA	-	-	2-back task	↔ Working memory performance
Yan et al., 2011b [137]	Comparative	NN HH	2 groups: 1 born at SL, the other in HA	-	-	Verbal working memory	↑RT in HA group ↓ Accuracy in HA group

Table 2. ROBINS-I-V2 risk of bias assessment for data items.

Data item	Domain 1	Domain 2	Domain 3	Domain 4	Domain 5	Domain 6	Domain 7	Overall
Chroboczek et al., 2023	Serious	Low	Low	Low	Moderate	Low	Moderate	Serious
Crow et al., 1971	Moderate	Moderate	Moderate	Low	Low	Low	Moderate	Moderate
Gerard et al., 2000	Serious	Low	Low	Low	Serious	Low	Moderate	Serious
Guo et al., 2017	Serious	Low	Moderate	Low	Low	Serious	Moderate	Serious
Quackenbush et al., 2016	Moderate	Serious	Moderate	Serious	Low	Low	Moderate	Serious
Van Dorp et al., 2007	Serious	Low	Low	Low	Low	Low	Moderate	Serious

Domain 1, Bias due to confounding; Domain 2, Bias in the selection of participants; Domain 3, Bias in the classification of interventions; Domain 4, Bias due to deviations from intended interventions; Domain 5, Bias due to missing data; Domain 6, Bias in the measurement of outcomes; Domain 7, Bias in the selection of the reported result.

Table 3. List of confounding factors taken into consideration of data items assessed against Domain 1 of the ROBINS-I V2.

Participant factors	Task factors	Environmental factors	Methodological factors
- Differences in baseline cognitive performance between conditions - Age variability - Physical fitness variability - Health conditions - Acclimatisation - Genetic factors	- Complexity of task - Familiarity to task	- Severity and duration of hypoxia - Simulation or environmental hypoxia - Temperature and pressure	- Sensitivity and reliability of tools used - Timing of assessments - Control groups used

Table 4. List of each cognitive domain construct and the associated cognitive tasks.

Cognitive Domain	Cognitive task	
Memory	• Working memory (score) • Learning and memory (score) • Short-term memory (score) • Associative memory (score) • Episodic memory (score)	• Long-term memory (score) • Involuntary memory (score) • Visual memory (score) • Verbal memory (score)
Attention	• Response inhibition (score)	
Processing speed	• Processing speed (milliseconds)	
Psychomotor speed	• Psychomotor speed (score) • Motor speed (score)	• Accuracy (score) • Fine-motor control (score)
Executive function	• Executive function (score) • Verbal fluency (score) • Non-verbal fluency (score) • Tower task (score) • Cognitive flexibility (score) • Coding (score) • Decision-making (score) • Visual acuity (score) • Reasoning (score)	• Risk-taking (score) • Addition (score) • Incorrect answers (score) • Neurocognitive index (score) • Pattern recognition (score) • Number comparison (score) • Proof-reading (score) • Spatial tracking (score) • Map compass (score)

Figure 1. PRISMA flow diagram detailing the systematic inclusion and exclusion of data items. A total of 858 items were identified during the initial phase. One article was removed as a duplicate. The 857 articles were then screened based on titles and abstracts with 811 being excluded. From the forty-six data items assessed for eligibility, six were excluded due to risk of bias consisting of (e.g., concomitant exercise protocol, lack of control/ normoxia condition, and no control/normoxia statistical comparisons). Ultimately, forty data items were included for meta-analyses.

General Cognitive Ability

Study	SMD [95% CI]
Parker et al., 2017	-0.71 [-1.66, 0.25]
Shukitt-Hale et al., 1998	-2.04 [-2.78, -1.31]
Shukitt-Hale et al., 1998	-1.33 [-1.98, -0.67]
Ramirez-delaCruz et al., 2025	-1.44 [-2.12, -0.77]
Ramirez-delaCruz et al., 2025	-0.83 [-1.43, -0.23]
Ramirez-delaCruz et al., 2025	-0.40 [-0.97, 0.16]
Bakharev et al., 1981	0.46 [-0.69, 1.61]
Ando et al., 2023	-0.89 [-1.78, 0.01]
Asmaro et al., 2013	-0.80 [-1.29, -0.30]
Bloomfield et al., 2023	-0.13 [-0.97, 0.70]
Davis et al., 2015	0.00 [-0.42, 0.42]
Davis et al., 2015	-1.25 [-1.90, -0.60]
Falla et al., 2024a	0.00 [-0.40, 0.40]
Falla et al., 2024b	0.00 [-0.47, 0.47]
Hill et al., 2014	0.49 [-0.16, 1.14]
Hohenaur et al., 2022	0.04 [-0.60, 0.67]
Hohenaur et al., 2022	-0.36 [-1.00, 0.28]
Latshang et al., 2013	0.00 [-0.84, 0.84]
Latshang et al., 2013	-0.23 [-1.07, 0.61]
Legg et al., 2016	0.21 [-0.26, 0.68]
Legg et al., 2016	-0.08 [-0.55, 0.39]
Legg et al., 2016	0.46 [-0.01, 0.94]
Legg et al., 2016	-0.05 [-0.52, 0.42]
Ramirez-delaCruz et al., 2025	-0.63 [-1.20, -0.06]
Ramirez-delaCruz et al., 2025	-0.35 [-0.91, 0.21]
Schega et al., 2013	0.46 [-0.24, 1.16]
Schega et al., 2016	-0.31 [-0.99, 0.36]
Schega et al., 2016	-0.67 [-1.36, 0.03]
Thomas et al., 2007	0.07 [-0.80, 0.95]
Turner et al., 2015a	-0.38 [-0.99, 0.23]
Turner et al., 2015a	-0.43 [-1.05, 0.18]
Turner et al., 2015b	-0.32 [-1.06, 0.43]
Wu et al., 2023	1.03 [0.56, 1.49]
Shukitt-Hale et al., 1998	-2.22 [-2.97, -1.47]
Shukitt-Hale et al., 1998	-1.15 [-1.79, -0.51]
Bloomfield et al., 2023	-0.91 [-1.80, -0.02]
Hohenaur et al., 2022	0.02 [-0.62, 0.65]
Hohenaur et al., 2022	0.02 [-0.62, 0.65]
Legg et al., 2016	-0.01 [-0.48, 0.46]
Legg et al., 2016	-0.39 [-0.86, 0.09]
Turner et al., 2015a	-0.44 [-1.06, 0.17]
Turner et al., 2015b	-0.37 [-1.12, 0.38]
Bloomfield et al., 2023	-1.56 [-2.55, -0.57]
Turner et al., 2015b	-0.58 [-1.35, 0.18]
Aebi et al., 2020	-0.23 [-0.95, 0.49]
Niedermeier et al., 2017	-0.60 [-1.05, -0.16]
Hill et al., 2014	-1.86 [-2.66, -1.06]
Asmaro et al., 2013	-1.43 [-2.00, -0.85]
Asmaro et al., 2013	-1.93 [-2.53, -1.33]
Bloomfield et al., 2023	-0.83 [-1.71, 0.05]
Keeler et al., 2023	-0.36 [-0.98, 0.27]
Keeler et al., 2023	0.08 [-0.54, 0.70]
Turner et al., 2015a	-0.44 [-1.05, 0.17]
Parker et al., 2017	0.06 [-0.86, 0.98]
Aebi et al., 2020	0.89 [0.13, 1.65]
Bakharev et al., 1981	-0.37 [-1.51, 0.78]
Bailey et al., 2019	-1.61 [-2.35, -0.88]
Bakharev et al., 1981	0.37 [-0.78, 1.51]
Shukitt-Hale et al., 1998	0.93 [0.31, 1.55]
Shukitt-Hale et al., 1998	-0.36 [-0.95, 0.24]
Bloomfield et al., 2023	-0.33 [-1.17, 0.51]
Bloomfield et al., 2023	-1.41 [-2.35, -0.48]
Merz et al., 2013	-0.95 [-1.47, -0.42]
Shukitt-Hale et al., 1998	-2.74 [-3.57, -1.91]
Shukitt-Hale et al., 1998	-2.00 [-2.72, -1.28]
Bakharev et al., 1981	-2.37 [-3.95, -0.79]
Shukitt-Hale et al., 1998	-1.13 [-1.77, -0.50]
Shukitt-Hale et al., 1998	-1.47 [-2.14, -0.81]
Bailey et al., 2019	-0.24 [-0.89, 0.42]
Bloomfield et al., 2023	0.12 [-0.72, 0.95]
Bloomfield et al., 2023	0.46 [-0.39, 1.31]
Falla et al., 2024a	0.22 [-0.19, 0.62]
Hill et al., 2014	-0.22 [-0.85, 0.42]
Hohenaur et al., 2022	0.25 [-0.39, 0.89]
Hohenaur et al., 2022	-0.01 [-0.64, 0.63]
Latshang et al., 2013	0.09 [-0.75, 0.93]

Study	SMD [95% CI]
Noble et al., 1993	-1.71 [-2.76, -0.66]
Noble et al., 1993	-2.46 [-3.57, -1.35]
Ramirez-delaCruz et al., 2025	-0.63 [-1.20, -0.06]
Ramirez-delaCruz et al., 2025	-0.22 [-0.77, 0.34]
Schega et al., 2013	-0.09 [-0.79, 0.60]
Turner et al., 2015a	0.28 [-0.33, 0.89]
Wu et al., 2023	-0.41 [-0.85, 0.04]
Wu et al., 2023	-0.73 [-1.19, -0.27]
Yan et al., 2011a	-0.56 [-1.10, -0.01]
Yan et al., 2011b	-0.61 [-1.15, -0.07]
Bakharev et al., 1981	-0.67 [-1.83, 0.50]
Bailey et al., 2019	-0.61 [-1.29, 0.07]
Bloomfield et al., 2023	-0.49 [-1.34, 0.36]
Davis et al., 2015	-1.44 [-1.94, -0.94]
Davis et al., 2015	-1.97 [-2.64, -1.30]
Falla et al., 2024a	-0.22 [-0.63, 0.19]
Falla et al., 2024b	0.17 [-0.30, 0.64]
Falla et al., 2024b	-0.30 [-0.77, 0.17]
Hill et al., 2014	-1.08 [-1.77, -0.40]
Noble et al., 1993	-1.09 [-1.99, -0.19]
Hohenaur et al., 2022	-0.16 [-0.80, 0.48]
Hohenaur et al., 2022	-0.31 [-0.95, 0.33]
Legg et al., 2016	0.08 [-0.39, 0.55]
Legg et al., 2016	0.05 [-0.42, 0.52]
Falla et al., 2024a	-0.10 [-0.51, 0.30]
Pighin et al., 2014	-0.25 [-0.81, 0.31]
Bakharev et al., 1981	-1.84 [-3.21, -0.46]
Bonnon et al., 1999	-0.01 [-1.14, 1.12]
Bloomfield et al., 2023	-0.07 [-0.90, 0.77]
Aebi et al., 2020	1.28 [0.48, 2.08]
Keeler et al., 2023	0.05 [-0.57, 0.67]
Keeler et al., 2023	-0.05 [-0.67, 0.57]
Shukitt-Hale et al., 1998	-1.33 [-1.99, -0.68]
Shukitt-Hale et al., 1998	-0.72 [-1.33, -0.11]
Davis et al., 2015	0.09 [-0.35, 0.53]
Davis et al., 2015	-0.16 [-0.72, 0.39]
Merz et al., 2013	-0.67 [-1.19, -0.16]
Bloomfield et al., 2023	-0.40 [-1.24, 0.45]
Schlaepfer et al., 1992	-1.32 [-2.39, -0.26]
Schlaepfer et al., 1992	-2.28 [-3.53, -1.04]
Bloomfield et al., 2023	-1.32 [-2.28, -0.36]
Asmaro et al., 2013	-1.78 [-2.38, -1.18]
Asmaro et al., 2013	-1.81 [-2.38, -1.25]
Bailey et al., 2019	-1.24 [-2.00, -0.48]
Champod et al., 2013	0.60 [-0.11, 1.30]
Champod et al., 2013	0.51 [-0.20, 1.22]
Falla et al., 2024b	0.18 [-0.29, 0.65]
Falla et al., 2024b	0.12 [-0.35, 0.59]
Hill et al., 2014	-0.26 [-0.90, 0.37]
Hohenaur et al., 2022	-0.25 [-0.89, 0.39]
Hohenaur et al., 2022	0.08 [-0.55, 0.72]
Hohenaur et al., 2022	0.11 [-0.52, 0.75]
Hohenaur et al., 2022	-0.01 [-0.65, 0.62]
Hohenaur et al., 2022	0.01 [-0.63, 0.64]
Hohenaur et al., 2022	-0.32 [-0.96, 0.32]
Hohenaur et al., 2022	-0.02 [-0.65, 0.62]
Hohenaur et al., 2022	-0.19 [-0.83, 0.45]
Keeler et al., 2023	0.29 [-0.33, 0.92]
Keeler et al., 2023	-0.28 [-0.90, 0.35]
Latshang et al., 2013	0.17 [-0.66, 1.01]
Legg et al., 2016	0.05 [-0.41, 0.52]
Legg et al., 2016	-0.18 [-0.65, 0.29]
Legg et al., 2016	-0.10 [-0.57, 0.37]
Legg et al., 2016	-0.06 [-0.53, 0.40]
Noble et al., 1993	-0.40 [-1.24, 0.45]
Parker et al., 2017	0.00 [-0.92, 0.92]
Ramirez-delaCruz et al., 2025	-0.10 [-0.66, 0.45]
Ramirez-delaCruz et al., 2025	0.26 [-0.30, 0.82]
Ramirez-delaCruz et al., 2025	-0.02 [-0.58, 0.53]
Thomas et al., 2007	0.60 [-0.30, 1.49]
Turner et al., 2015a	-1.41 [-2.11, -0.71]
Turner et al., 2015a	-1.05 [-1.72, -0.38]
Turner et al., 2015a	-1.32 [-1.99, -0.64]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Wu et al., 2023	0.39 [-0.05, 0.83]
Random-Effects Model	-0.43 [-0.54, -0.32]

Standardized Mean Difference

Figure 2. Forest plot for all 151 effect sizes with the grouped effect for the random-effects model categorising cognitive ability in hypoxia.

Figure 3. Predicted cognitive ability decline in hypoxia as a function of severity.

Memory

Study	SMD [95% CI]
Bakharev et al., 1981	0.46 [-0.69, 1.61]
Hill et al., 2014	-1.86 [-2.66, -1.06]
Bakharev et al., 1981	-0.37 [-1.51, 0.78]
Bailey et al., 2019	-1.61 [-2.35, -0.88]
Bakharev et al., 1981	0.37 [-0.78, 1.51]
Bakharev et al., 1981	-1.84 [-3.21, -0.46]
Bonnon et al., 1999	-0.01 [-1.14, 1.12]
Bloomfield et al., 2023	-0.40 [-1.24, 0.45]
Bloomfield et al., 2023	-1.32 [-2.28, -0.36]
Bakharev et al., 1981	-2.37 [-3.95, -0.79]
Asmaro et al., 2013	-1.78 [-2.38, -1.18]
Asmaro et al., 2013	-1.81 [-2.38, -1.25]
Bailey et al., 2019	-1.24 [-2.00, -0.48]
Champod et al., 2013	0.60 [-0.11, 1.30]
Champod et al., 2013	0.51 [-0.20, 1.22]
Falla et al., 2024b	0.18 [-0.29, 0.65]
Falla et al., 2024b	0.12 [-0.35, 0.59]
Hill et al., 2014	-0.26 [-0.90, 0.37]
Hohenaur et al., 2022	-0.25 [-0.89, 0.39]
Hohenaur et al., 2022	0.08 [-0.55, 0.72]
Hohenaur et al., 2022	0.11 [-0.52, 0.75]
Hohenaur et al., 2022	-0.01 [-0.65, 0.62]
Hohenaur et al., 2022	0.01 [-0.63, 0.64]
Hohenaur et al., 2022	-0.32 [-0.96, 0.32]
Hohenaur et al., 2022	-0.02 [-0.65, 0.62]
Hohenaur et al., 2022	-0.19 [-0.83, 0.45]
Keeler et al., 2023	0.29 [-0.33, 0.92]
Keeler et al., 2023	-0.28 [-0.90, 0.35]
Latshang et al., 2013	0.17 [-0.66, 1.01]
Legg et al., 2016	0.05 [-0.41, 0.52]
Legg et al., 2016	-0.18 [-0.65, 0.29]
Legg et al., 2016	-0.10 [-0.57, 0.37]
Legg et al., 2016	-0.06 [-0.53, 0.40]
Noble et al., 1993	-0.40 [-1.24, 0.45]
Parker et al., 2017	0.00 [-0.92, 0.92]
Ramirez-delaCruz et al., 2025	-0.10 [-0.66, 0.45]
Ramirez-delaCruz et al., 2025	0.26 [-0.30, 0.82]
Ramirez-delaCruz et al., 2025	-0.02 [-0.58, 0.53]
Thomas et al., 2007	0.60 [-0.30, 1.49]
Turner et al., 2015a	-1.41 [-2.11, -0.71]
Turner et al., 2015a	-1.05 [-1.72, -0.38]
Turner et al., 2015a	-1.32 [-1.99, -0.64]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Wu et al., 2023	0.39 [-0.05, 0.83]
Random-Effects Model	-0.35 [-0.56, -0.15]

Standardized Mean Difference

Memory

Study	SMD [95% CI]
Bakharev et al., 1981	0.46 [-0.69, 1.61]
Hill et al., 2014	-1.86 [-2.66, -1.06]
Bakharev et al., 1981	-0.37 [-1.51, 0.78]
Bailey et al., 2019	-1.61 [-2.35, -0.88]
Bakharev et al., 1981	0.37 [-0.78, 1.51]
Bakharev et al., 1981	-1.84 [-3.21, -0.46]
Bonnon et al., 1999	-0.01 [-1.14, 1.12]
Bloomfield et al., 2023	-0.40 [-1.24, 0.45]
Bloomfield et al., 2023	-1.32 [-2.28, -0.36]
Bakharev et al., 1981	-2.37 [-3.95, -0.79]
Asmaro et al., 2013	-1.78 [-2.38, -1.18]
Asmaro et al., 2013	-1.81 [-2.38, -1.25]
Bailey et al., 2019	-1.24 [-2.00, -0.48]
Champod et al., 2013	0.60 [-0.11, 1.30]
Champod et al., 2013	0.51 [-0.20, 1.22]
Falla et al., 2024b	0.18 [-0.29, 0.65]
Falla et al., 2024b	0.12 [-0.35, 0.59]
Hill et al., 2014	-0.26 [-0.90, 0.37]
Hohenaur et al., 2022	-0.25 [-0.89, 0.39]
Hohenaur et al., 2022	0.08 [-0.55, 0.72]
Hohenaur et al., 2022	0.11 [-0.52, 0.75]
Hohenaur et al., 2022	-0.01 [-0.65, 0.62]
Hohenaur et al., 2022	0.01 [-0.63, 0.64]
Hohenaur et al., 2022	-0.32 [-0.96, 0.32]
Hohenaur et al., 2022	-0.02 [-0.65, 0.62]
Hohenaur et al., 2022	-0.19 [-0.83, 0.45]
Keeler et al., 2023	0.29 [-0.33, 0.92]
Keeler et al., 2023	-0.28 [-0.90, 0.35]
Latshang et al., 2013	0.17 [-0.66, 1.01]
Legg et al., 2016	0.05 [-0.41, 0.52]
Legg et al., 2016	-0.18 [-0.65, 0.29]
Legg et al., 2016	-0.10 [-0.57, 0.37]
Legg et al., 2016	-0.06 [-0.53, 0.40]
Noble et al., 1993	-0.40 [-1.24, 0.45]
Parker et al., 2017	0.00 [-0.92, 0.92]
Ramirez-delaCruz et al., 2025	-0.10 [-0.66, 0.45]
Ramirez-delaCruz et al., 2025	0.26 [-0.30, 0.82]
Ramirez-delaCruz et al., 2025	-0.02 [-0.58, 0.53]
Thomas et al., 2007	0.60 [-0.30, 1.49]
Turner et al., 2015a	-1.41 [-2.11, -0.71]
Turner et al., 2015a	-1.05 [-1.72, -0.38]
Turner et al., 2015a	-1.32 [-1.99, -0.64]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Turner et al., 2015b	-0.64 [-1.40, 0.12]
Wu et al., 2023	0.39 [-0.05, 0.83]
Random-Effects Model	-0.35 [-0.56, -0.15]

Standardized Mean Difference

Processing Speed

Study	SMD [95% CI]
Bailey et al., 2019	-0.24 [-0.89, 0.42]
Bloomfield et al., 2023	0.12 [-0.72, 0.95]
Bloomfield et al., 2023	0.46 [-0.39, 1.31]
Falla et al., 2024a	0.22 [-0.19, 0.62]
Hill et al., 2014	-0.22 [-0.85, 0.42]
Hohenaur et al., 2022	0.25 [-0.39, 0.89]
Hohenaur et al., 2022	-0.01 [-0.64, 0.63]
Latshang et al., 2013	0.09 [-0.75, 0.93]
Noble et al., 1993	-1.71 [-2.76, -0.66]
Noble et al., 1993	-2.46 [-3.57, -1.35]
Ramirez-delaCruz et al., 2025	-0.63 [-1.20, -0.06]
Ramirez-delaCruz et al., 2025	-0.22 [-0.77, 0.34]
Schega et al., 2013	-0.09 [-0.79, 0.60]
Turner et al., 2015a	0.28 [-0.33, 0.89]
Wu et al., 2023	-0.41 [-0.85, 0.04]
Wu et al., 2023	-0.73 [-1.19, -0.27]
Yan et al., 2011a	-0.56 [-1.10, -0.01]
Yan et al., 2011b	-0.61 [-1.15, -0.07]
Random-Effects Model	-0.30 [-0.54, -0.05]

Standardized Mean Difference

Psychomotor Speed

Study		SMD [95% CI]
Parker et al., 2017		-0.71 [-1.66, 0.25]
Bloomfield et al., 2023		-0.33 [-1.17, 0.51]
Bailey et al., 2019		-0.61 [-1.29, 0.07]
Bloomfield et al., 2023		-0.49 [-1.34, 0.36]
Davis et al., 2015		-1.44 [-1.94, -0.94]
Davis et al., 2015		-1.97 [-2.64, -1.30]
Falla et al., 2024a		-0.22 [-0.63, 0.19]
Falla et al., 2024b		0.17 [-0.30, 0.64]
Falla et al., 2024b		-0.30 [-0.77, 0.17]
Hill et al., 2014		-1.08 [-1.77, -0.40]
Noble et al., 1993		-1.09 [-1.99, -0.19]
Random-Effects Model		-0.72 [-1.10, -0.33]

Standardized Mean Difference

Figure 4. Forest plot for each cognitive domain for the random-effects model categorising hypoxia's domain-specific impact.

Figure 5. Bubble plot visualising the 151 effect sizes across all domains in relation to the severity and duration of hypoxia exposure. Notably, as fraction of inspired oxygen decreases effect of hypoxia increases for most cognitive domains.

Figure 6. Funnel plot visualising the distribution of effect sizes included in the reduced model of the General Cognitive Ability Domain.

REFERENCES

1. Martin, K., et al., *The Impact of Environmental Stress on Cognitive Performance: A Systematic Review.* Human Factors, 2019. **61**(8): p. 1205-1246.
2. Taylor, L., et al., *The Impact of Different Environmental Conditions on Cognitive Function: A Focused Review.* Front Physiol, 2015. **6**: p. 372.
3. Coronel-Oliveros, C., et al., *Elevating Understanding: Linking High-Altitude Hypoxia to Brain Aging Through EEG Functional Connectivity and Spectral Analyses.* Network Neuroscience, 2023: p. 1-39.
4. Otto, S.R., et al., *EEG as a neural measure of hypoxia-related impairment.* Frontiers in Cognition, 2025. **Volume 4 - 2025**.
5. Limmer, M. and P. Platen, *The influence of hypoxia and prolonged exercise on attentional performance at high and extreme altitudes: A pilot study.* PLoS One, 2018. **13**(10): p. e0205285.
6. Bonnon, M., M.C. Noel-Jorand, and P. Therme, *Effects of different stay durations on attentional performance during two mountain expeditions.* Aviat Space Environ Med, 2000. **71**(7): p. 678-84.
7. Ando, S., Y. Yamada, and M. Kokubu, *Reaction time to peripheral visual stimuli during exercise under hypoxia.* J Appl Physiol (1985), 2010. **108**(5): p. 1210-6.
8. Ando, S., et al., *Acute hypoxia impairs posterior cerebral bioenergetics and memory in man.* Exp Physiol, 2023. **108**(12): p. 1516-1530.
9. Wang, Y., et al., *Long-Term Exposure to High Altitude Affects Voluntary Spatial Attention at Early and Late Processing Stages.* Scientific Reports, 2014. **4**(1): p. 4443.
10. Kramer, A.F., J.T. Coyne, and D.L. Strayer, *Cognitive function at high altitude.* Hum Factors, 1993. **35**(2): p. 329-44.
11. Asmaro, D., J. Mayall, and S. Ferguson, *Cognition at altitude: impairment in executive and memory processes under hypoxic conditions.* Aviat Space Environ Med, 2013. **84**(11): p. 1159-65.
12. Malle, C., et al., *Working memory impairment in pilots exposed to acute hypobaric hypoxia.* Aviat Space Environ Med, 2013. **84**(8): p. 773-9.
13. Szameitat, A.J., et al., *Localization of executive functions in dual-task performance with fMRI.* Journal of Cognitive Neuroscience, 2002. **14**(8): p. 1184-99.
14. Ghanavati, E., et al., *Differential role of prefrontal, temporal and parietal cortices in verbal and figural fluency: Implications for the supramodal contribution of executive functions.* Scientific Reports, 2019. **9**(1): p. 3700.
15. Ando, S., et al., *The interactive effects of acute exercise and hypoxia on cognitive performance: A narrative review.* Scandinavian Journal of Medicine & Science in Sports, 2020. **30**(3): p. 384-398.
16. Ochi, G., et al., *Neural basis for reduced executive performance with hypoxic exercise.* Neuroimage, 2018. **171**: p. 75-83.
17. Chroboczek, M., et al., *Acute Normobaric Hypoxia Lowers Executive Functions among Young Men despite Increase of BDNF Concentration.* International Journal of Environmental Research and Public Health, 2022. **19**(17): p. 10802.
18. Ochi, G., et al., *Hypoxia-induced lowered executive function depends on arterial oxygen desaturation.* The Journal of Physiological Sciences, 2018. **68**(6): p. 847-853.
19. Silva-Fernandes, A., et al., *Processing speed mediates the association between physical activity and executive functioning in elderly adults.* Front Psychol, 2022. **13**: p. 958535.

20. Seech, T.R., et al., *Impaired Sensory Processing During Low-Oxygen Exposure: A Noninvasive Approach to Detecting Changes in Cognitive States.* Front Psychiatry, 2020. **11**: p. 12.

21. Blacker, K.J. and D.G. McHail, *Effects of Acute Hypoxia on Early Visual and Auditory Evoked Potentials.* Front Neurosci, 2022. **16**: p. 846001.

22. Barbur, J.L. and D.M. and Connolly, *Effects of hypoxia on color vision with emphasis on the mesopic range.* Expert Review of Ophthalmology, 2011. **6**(4): p. 409-420.

23. Wong-Riley, M.T., *Energy metabolism of the visual system.* Eye Brain, 2010. **2**: p. 99-116.

24. Kumar, U., et al., *Mapping the unique neural engagement in deaf individuals during picture, word, and sign language processing: fMRI study.* Brain Imaging and Behavior, 2024. **18**(4): p. 835-851.

25. Bloomfield, P.M., et al., *Carbon dioxide protects simulated driving performance during severe hypoxia.* Eur J Appl Physiol, 2023. **123**(7): p. 1583-1593.

26. Fowler, B., et al., *The threshold for hypoxia effects on perceptual-motor performance.* Hum Factors, 1987. **29**(1): p. 61-6.

27. Davis, J.E., et al., *Cognitive and psychomotor responses to high-altitude exposure in sea level and high-altitude residents of Ecuador.* J Physiol Anthropol, 2015. **34**(1): p. 2.

28. Bouquet, C.A., et al., *Psychomotor skills learning under chronic hypoxia.* Neuroreport, 1999. **10**(14): p. 3093-9.

29. McKeown, D.J., M.J. Simmonds, and J.J. Kavanagh, *Reduced blood oxygen levels induce changes in low-threshold motor unit firing that align with the individual's tolerance to hypoxia.* J Neurophysiol, 2019. **121**(5): p. 1664-1671.

30. McKeown, D.J., G.M. Stewart, and J.J. Kavanagh, *The severity of acute hypoxaemia determines distinct changes in intracortical and spinal neural circuits.* Exp Physiol, 2023.

31. McKeown, D.J., et al., *Severe acute hypoxia impairs recovery of voluntary muscle activation after sustained submaximal elbow flexion.* J Physiol, 2021.

32. Goodall, S., E.Z. Ross, and L.M. Romer, *Effect of graded hypoxia on supraspinal contributions to fatigue with unilateral knee-extensor contractions.* J Appl Physiol (1985), 2010. **109**(6): p. 1842-51.

33. Fowler, B., M. Taylor, and G. Porlier, *The effects of hypoxia on reaction time and movement time components of a perceptual-motor task.* Ergonomics, 1987. **30**(10): p. 1475-85.

34. Pun, M., et al., *Effect of Acute, Subacute, and Repeated Exposure to High Altitude (5050 m) on Psychomotor Vigilance.* Front Physiol, 2018. **9**: p. 677.

35. McMorris, T., et al., *Effect of acute hypoxia on cognition: A systematic review and meta-regression analysis.* Neurosci Biobehav Rev, 2017. **74**(Pt A): p. 225-232.

36. Boulares, A., et al., *Effects of Intermittent Hypoxia Protocols on Cognitive Performance and Brain Health in Older Adults Across Cognitive States: A Systematic Literature Review.* J Alzheimers Dis, 2024. **101**(1): p. 13-30.

37. Virués-Ortega, J., et al., *Neuropsychological Functioning Associated with High-Altitude Exposure.* Neuropsychology Review, 2004. **14**(4): p. 197-224.

38. Post, T.E., et al., *Sensitivity of cognitive function tests to acute hypoxia in healthy subjects: a systematic literature review.* Frontiers in Physiology, 2023. **14**.

39. Aboouf, M.A., et al., *The Brain at High Altitude: From Molecular Signaling to Cognitive Performance.* Int J Mol Sci, 2023. **24**(12).

40. Craighero, L., *The Role of the Sensorimotor System in Cognitive Functions.* Brain Sci, 2022. **12**(5).

41.   Blacker, K.J., et al., *Deficits in Visual Processing During Hypoxia as Evidenced by Visual Mismatch Negativity.* Aerospace Medicine and Human Performance, 2021. **92**(5): p. 326-332.

42.   Krause, W.L., et al., *Acute hypoxia activates human 8–12 Hz physiological tremor.* Respiration Physiology, 2000. **123**(1): p. 131-141.

43.   Goodall, S., et al., *Supraspinal fatigue after normoxic and hypoxic exercise in humans.* J Physiol, 2012. **590**(11): p. 2767-82.

44.   Goodall, S., G. Howatson, and K. Thomas, *Modulation of specific inhibitory networks in fatigued locomotor muscles of healthy males.* Exp Brain Res, 2018. **236**(2): p. 463-473.

45.   Kayser, B., R. Bökenkamp, and T. Binzoni, *Alpha-motoneuron excitability at high altitude.* European journal of applied physiology and occupational physiology, 1993. **66**: p. 1-4.

46.   Willer, J.C., G. Miserocchi, and H. Gautier, *Hypoxia and monosynaptic reflexes in humans.* Journal of Applied Physiology, 1987. **63**(2): p. 639-645.

47.   Mayer, V. and A.J. Yates, *Intellectual changes following temporal lobectomy for psychomotor epilepsy; preliminary communication.* J Neurol Neurosurg Psychiatry, 1955. **18**(1): p. 44-52.

48.   Mayes, A.R. and N. Roberts, *Theories of episodic memory.* Philos Trans R Soc Lond B Biol Sci, 2001. **356**(1413): p. 1395-408.

49.   Ekstrom, A.D. and C. Ranganath, *Space, time, and episodic memory: The hippocampus is all over the cognitive map.* Hippocampus, 2018. **28**(9): p. 680-687.

50.   Spiers, H.J., et al., *Bilateral hippocampal pathology impairs topographical and episodic memory but not visual pattern matching.* Hippocampus, 2001. **11**(6): p. 715-25.

51.   Rosenbaum, R.S., et al., *Remote spatial memory in an amnesic person with extensive bilateral hippocampal lesions.* Nature Neuroscience, 2000. **3**(10): p. 1044-1048.

52.   Bird, C.M. and N. Burgess, *The hippocampus and memory: insights from spatial processing.* Nature Reviews Neuroscience, 2008. **9**(3): p. 182-194.

53.   Vetrovoy, O., et al., *Prenatal Hypoxia Induces Premature Aging Accompanied by Impaired Function of the Glutamatergic System in Rat Hippocampus.* Neurochem Res, 2021. **46**(3): p. 550-563.

54.   Zhang, X.Y., et al., *Crocin attenuates acute hypobaric hypoxia-induced cognitive deficits of rats.* Eur J Pharmacol, 2018. **818**: p. 300-305.

55.   Zhu, M., et al., *Effect of acute exposure to hypobaric hypoxia on learning and memory in adult Sprague-Dawley rats.* Behav Brain Res, 2019. **367**: p. 82-90.

56.   Shushanyan, R., et al., *Histological and cytochemical analysis of the brain under conditions of hypobaric hypoxia-induced oxygen deficiency in albino rats.* Acta Histochem, 2023. **125**(8): p. 152114.

57.   Lu, G., G. Rili, and M. Shuang, *Impact of hypoxia on the hippocampus: A review.* Medicine (Baltimore), 2025. **104**(12): p. e41479.

58.   Biswal, S., et al., *Global hypoxia induced impairment in learning and spatial memory is associated with precocious hippocampal aging.* Neurobiology of Learning and Memory, 2016. **133**: p. 157-170.

59.   Ai, J. and A. Baker, *Long-term potentiation of evoked presynaptic response at CA3–CA1 synapses by transient oxygen-glucose deprivation in rat brain slices.* Experimental Brain Research, 2006. **169**(1): p. 126-129.

60.   Crepel, V., et al., *Anoxia-induced LTP of isolated NMDA receptor-mediated synaptic responses.* Journal of Neurophysiology, 1993. **69**(5): p. 1774-1778.

61. Quintana, P., et al., *Acid-sensing ion channel 1a drives AMPA receptor plasticity following ischaemia and acidosis in hippocampal CA1 neurons.* J Physiol, 2015. **593**(19): p. 4373-86.

62. Kumral, A., et al., *Erythropoietin improves long-term spatial memory deficits and brain injury following neonatal hypoxia-ischemia in rats.* Behav Brain Res, 2004. **153**(1): p. 77-86.

63. Ji, W., et al., *NMDA Receptor-Mediated Excitotoxicity Is Involved in Neuronal Apoptosis and Cognitive Impairment Induced by Chronic Hypobaric Hypoxia Exposure at High Altitude.* High Alt Med Biol, 2021. **22**(1): p. 45-57.

64. Bekker, A., et al., *Isoflurane preserves spatial working memory in adult mice after moderate hypoxia.* Anesth Analg, 2006. **102**(4): p. 1134-8.

65. Gibson, G.E. and T.E. Duffy, *Impaired synthesis of acetylcholine by mild hypoxic hypoxia or nitrous oxide.* J Neurochem, 1981. **36**(1): p. 28-33.

66. Hencz, A., et al., *Mild hypoxia-induced structural and functional changes of the hippocampal network.* Front Cell Neurosci, 2023. **17**: p. 1277375.

67. McCullough, M.L., *The behavioral effects of prenatal hypoxia in the rat.* Dev Psychobiol, 1976. **9**(4): p. 335-42.

68. Sun, M.K., H. Xu, and D.L. Alkon, *Pharmacological protection of synaptic function, spatial learning, and memory from transient hypoxia in rats.* J Pharmacol Exp Ther, 2002. **300**(2): p. 408-16.

69. Lin, H., et al., *Attenuating brain edema, hippocampal oxidative stress, and cognitive dysfunction in rats using hyperbaric oxygen preconditioning during simulated high-altitude exposure.* J Trauma Acute Care Surg, 2012. **72**(5): p. 1220-7.

70. Shao, G., et al., *Hypoxic Preconditioning Improves Spatial Cognitive Ability in Mice.* Neurosignals, 2008. **15**(6): p. 314-321.

71. Zani, A., Y. Dishi, and A.M. Proverbio, *From oxygen shortage to neurocognitive challenges: behavioral patterns and imaging insights.* Frontiers in Cognition, 2024. **Volume 3 - 2024**.

72. Iaria, G., et al., *Retrosplenial and hippocampal brain regions in human navigation: complementary functional contributions to the formation and use of cognitive maps.* European Journal of Neuroscience, 2007. **25**(3): p. 890-899.

73. Hoiland, R.L., et al., *Ventilatory and cerebrovascular regulation and integration at high-altitude.* Clin Auton Res, 2018. **28**(4): p. 423-435.

74. Lee, D., et al., *Prefrontal executive function enhanced by prior acute inhalation of low-dose hypoxic gas: Modulation via cardiac vagal activity.* Neuroimage, 2025. **310**: p. 121139.

75. West, J.B., *High-Altitude Medicine.* American Journal of Respiratory and Critical Care Medicine, 2012. **186**(12): p. 1229-1237.

76. Caldwell, H.G., et al., *Stability in neurovascular function at 3800m.* Physiol Behav, 2017. **182**: p. 62-68.

77. Finley, A.J., et al., *Resting EEG Periodic and Aperiodic Components Predict Cognitive Decline Over 10 Years.* The Journal of Neuroscience, 2024. **44**(13): p. e1332232024.

78. Salthouse, T.A., *Shared and unique influences on age-related cognitive change.* Neuropsychology, 2017. **31**(1): p. 11-19.

79. Snyder, B., et al., *Cerebral Hypoxia: Its Role in Age-Related Chronic and Acute Cognitive Dysfunction.* Anesth Analg, 2021. **132**(6): p. 1502-1513.

80. McKetton, L., et al., *Cerebrovascular Resistance in Healthy Aging and Mild Cognitive Impairment.* Front Aging Neurosci, 2019. **11**: p. 79.

81. Haegens, S., et al., *Inter- and intra-individual variability in alpha peak frequency.* Neuroimage, 2014. **92**(100): p. 46-55.

82. Angelakis, E., et al., *Peak alpha frequency: an electroencephalographic measure of cognitive preparedness.* Clin Neurophysiol, 2004. **115**(4): p. 887-97.

83. Campbell, A.J., et al., *Resting-state EEG correlates of sustained attention in healthy ageing: Cross-sectional findings from the LEISURE study.* Neurobiol Aging, 2024. **144**: p. 68-77.

84. Ouyang, G., et al., *Decomposing alpha and 1/f brain activities reveals their differential associations with cognitive processing speed.* Neuroimage, 2020. **205**: p. 116304.

85. McKeown, D.J., et al., *Lower aperiodic EEG activity is associated with reduced verbal fluency performance across adulthood.* Neurobiology of Aging, 2025. **151**: p. 29-41.

86. Virtue-Griffiths, S., et al., *Task-related changes in aperiodic activity are related to visual working memory capacity independent of event-related potentials and alpha oscillations.* bioRxiv, 2022: p. 2022.01.18.476852.

87. Preston, M., N. Schaworonkow, and B. Voytek, *Oscillations and aperiodic activity: Evidence for dynamic changes in both during memory encoding.* bioRxiv, 2022: p. 2022.10.04.509632.

88. Preston, M., Jr., N. Schaworonkow, and B. Voytek, *Time-resolved aperiodic and oscillatory dynamics during human visual memory encoding.* J Neurosci, 2025.

89. Dave, S., et al., *Electrophysiological evidence for preserved primacy of lexical prediction in aging.* Neuropsychologia, 2018. **117**: p. 135-147.

90. Ozaki, H., S. Watanabe, and H. Suzuki, *Topographic EEG changes due to hypobaric hypoxia at simulated high altitude.* Electroencephalogr Clin Neurophysiol, 1995. **94**(5): p. 349-56.

91. Zhao, J.P., et al., *Characteristics of EEG activity during high altitude hypoxia and lowland reoxygenation.* Brain Res, 2016. **1648**(Pt A): p. 243-249.

92. Page, M.J., et al., *The PRISMA 2020 statement: an updated guideline for reporting systematic reviews.* BMJ, 2021. **372**: p. n71.

93. Clark, J., et al., *A full systematic review was completed in 2 weeks using automation tools: a case study.* J Clin Epidemiol, 2020. **121**: p. 81-90.

94. Scells, H. and G. Zuccon, *searchrefiner: A Query Visualisation and Understanding Tool for Systematic Reviews*, in *27th ACM International Conference on Information and Knowledge Management*. 2018, ACM: Torino, Italy. p. 1939 - 1942.

95. Forbes, C., et al., *Automation of duplicate record detection for systematic reviews: Deduplicator.* Systematic Reviews, 2024. **13**(1): p. 206.

96. Dopwell, F., et al., *"More than meets the eye": when the neonatal course may impact several years out.* J Dev Behav Pediatr, 2014. **35**(7): p. 467-9.

97. Aebi, M.R., et al., *Cognitive Impairment During Combined Normobaric vs. Hypobaric and Normoxic vs. Hypoxic Acute Exposure.* Aerosp Med Hum Perform, 2020. **91**(11): p. 845-851.

98. Bailey, D.M., et al., *Exaggerated systemic oxidative-inflammatory-nitrosative stress in chronic mountain sickness is associated with cognitive decline and depression.* J Physiol, 2019. **597**(2): p. 611-629.

99. Bakharev, V.D., *Investigation of memory during adaptation to high mountain conditions.* Hum Physiol, 1981. **7**(6): p. 409-14.

100. Blogg, S.L. and M. Gennser, *Cerebral blood flow velocity and psychomotor performance during acute hypoxia.* Aviat Space Environ Med, 2006. **77**(2): p. 107-13.

101. Bloomfield, P.M., et al., *Cocoa flavanols protect cognitive function, cerebral oxygenation, and mental fatigue during severe hypoxia.* J Appl Physiol (1985), 2023. **135**(2): p. 475-484.

102. Bonnon, M., M.C. Noël-Jorand, and P. Therme, *Psychological changes during altitude hypoxia.* Aviat Space Environ Med, 1995. **66**(4): p. 330-5.

103. Bonnon, M., M.C. Noël-Jorand, and P. Therme, *Criteria for psychological adaptation to high-altitude hypoxia.* Percept Mot Skills, 1999. **89**(1): p. 3-18.

104. Champod, A.S., et al., *Effects of acute intermittent hypoxia on working memory in young healthy adults.* Am J Respir Crit Care Med, 2013. **187**(10): p. 1148-50.

105. Chroboczek, M., et al., *Exercise-Induced Elevated BDNF Concentration Seems to Prevent Cognitive Impairment after Acute Exposure to Moderate Normobaric Hypoxia among Young Men.* Int J Environ Res Public Health, 2023. **20**(4).

106. Crow, T.J. and G.R. Kelman, *Effect of mild acute hypoxia on human short-term memory.* Br J Anaesth, 1971. **43**(6): p. 548-52.

107. Falla, M., et al., *Simulated Acute Hypobaric Hypoxia Effects on Cognition in Helicopter Emergency Medical Service Personnel - A Randomized, Controlled, Single-Blind, Crossover Trial.* Hum Factors, 2024. **66**(2): p. 404-423.

108. Falla, M., et al., *Effect of oxygen supplementation on cognitive performance among HEMS providers after acute exposure to altitude: the HEMS II randomized clinical trial.* Scand J Trauma Resusc Emerg Med, 2024. **32**(1): p. 65.

109. Gerard, A.B., et al., *Six percent oxygen enrichment of room air at simulated 5,000 m altitude improves neuropsychological function.* High Alt Med Biol, 2000. **1**(1): p. 51-61.

110. Guo, W.Y., et al., *Physiological and psychological factors associated with onset of high-altitude headache in Chinese men upon acute high-altitude exposure at 3700 m.* Cephalalgia, 2017. **37**(4): p. 336-347.

111. Hill, C.M., et al., *Cognitive performance in high-altitude Andean residents compared with low-altitude populations: from childhood to older age.* Neuropsychology, 2014. **28**(5): p. 752-60.

112. Hohenauer, E., et al., *The effects of normobaric and hypobaric hypoxia on cognitive performance and physiological responses: A crossover study.* PLoS One, 2022. **17**(11): p. e0277364.

113. Keeler, J.M., et al., *Gamified assessment of cognitive performance during moderate hypoxia.* PLoS One, 2023. **18**(7): p. e0288201.

114. Koller, E.A., et al., *The effects of acute altitude exposure in Swiss highlanders and lowlanders.* Eur J Appl Physiol Occup Physiol, 1993. **66**(2): p. 146-54.

115. Latshang, T.D., et al., *Are nocturnal breathing, sleep, and cognitive performance impaired at moderate altitude (1,630-2,590 m)?* Sleep, 2013. **36**(12): p. 1969-76.

116. Legg, S.J., et al., *Effects of mild hypoxia in aviation on mood and complex cognition.* Appl Ergon, 2016. **53 Pt B**: p. 357-63.

117. Merz, T.M., et al., *Cognitive performance in high-altitude climbers: a comparative study of saccadic eye movements and neuropsychological tests.* Eur J Appl Physiol, 2013. **113**(8): p. 2025-37.

118. Niedermeier, M., et al., *Is decision making in hypoxia affected by pre-acclimatisation? A randomized controlled trial.* Physiol Behav, 2017. **173**: p. 236-242.

119. Noble, J., J.G. Jones, and E.J. Davis, *Cognitive function during moderate hypoxaemia.* Anaesth Intensive Care, 1993. **21**(2): p. 180-4.

120. Parker, P.J., et al., *Working Memory Capacity and Surgical Performance While Exposed to Mild Hypoxic Hypoxemia.* Aerosp Med Hum Perform, 2017. **88**(10): p. 918-923.
121. Pighin, S., et al., *Loss aversion and hypoxia: less loss aversion in oxygen-depleted environment.* Stress, 2014. **17**(2): p. 204-10.
122. Quackenbush, J., et al., *Respiratory Muscle Training and Cognitive Function Exercising at Altitude.* Aerosp Med Hum Perform, 2016. **87**(1): p. 18-25.
123. Ramírez-delaCruz, M., et al., *Effects of different exposures to normobaric hypoxia on cognitive performance in healthy young adults.: Normobaric hypoxia and cognitive performance.* Physiol Behav, 2025. **288**: p. 114747.
124. Rice, G.M., et al., *Dry-EEG Manifestations of Acute and Insidious Hypoxia During Simulated Flight.* Aerosp Med Hum Perform, 2019. **90**(2): p. 92-100.
125. Schega, L., et al., *Effect of intermittent normobaric hypoxia on aerobic capacity and cognitive function in older people.* J Sci Med Sport, 2016. **19**(11): p. 941-945.
126. Schega, L., et al., *Effects of intermittent hypoxia on cognitive performance and quality of life in elderly adults: a pilot study.* Gerontology, 2013. **59**(4): p. 316-23.
127. Schlaepfer, T.E., P. Bärtsch, and H.U. Fisch, *Paradoxical effects of mild hypoxia and moderate altitude on human visual perception.* Clin Sci (Lond), 1992. **83**(5): p. 633-6.
128. Shukitt-Hale, B., L.E. Banderet, and H.R. Lieberman, *Elevation-dependent symptom, mood, and performance changes produced by exposure to hypobaric hypoxia.* Int J Aviat Psychol, 1998. **8**(4): p. 319-34.
129. Tesler, N., et al., *Ascent to moderate altitude impairs overnight memory improvements.* Physiol Behav, 2015. **139**: p. 121-6.
130. Thomas, R.J., et al., *Nocturnal hypoxia exposure with simulated altitude for 14 days does not significantly alter working memory or vigilance in humans.* Sleep, 2007. **30**(9): p. 1195-203.
131. Turner, C.E., et al., *Acute hypoxic gas breathing severely impairs cognition and task learning in humans.* Physiol Behav, 2015. **142**: p. 104-10.
132. Turner, C.E., W.D. Byblow, and N. Gant, *Creatine supplementation enhances corticomotor excitability and cognitive performance during oxygen deprivation.* J Neurosci, 2015. **35**(4): p. 1773-80.
133. van der Post, J., et al., *Evaluation of tests of central nervous system performance after hypoxemia for a model for cognitive impairment.* J Psychopharmacol, 2002. **16**(4): p. 337-43.
134. Van Dorp, E., et al., *Inspired carbon dioxide during hypoxia: effects on task performance and cerebral oxygen saturation.* Aviat Space Environ Med, 2007. **78**(7): p. 666-72.
135. Wu, Y., et al., *Remote ischemic preconditioning improves spatial memory and sleep of young males during acute high-altitude exposure.* Travel Med Infect Dis, 2023. **53**: p. 102576.
136. Yan, X., et al., *Adaptive influence of long term high altitude residence on spatial working memory: an fMRI study.* Brain Cogn, 2011. **77**(1): p. 53-9.
137. Yan, X., et al., *Prolonged high-altitude residence impacts verbal working memory: an fMRI study.* Experimental Brain Research, 2011. **208**(3): p. 437-445.