Приложение 1
К отчету по проекту “Влияние когнитивных искажений на восприятие (дез)информации, связанной со здоровьем, и их поведенческие и нейрональные корреляты”, 2023-2024 год
(1 год выполнения проекта)

СПИСОК ЛИТЕРАТУРЫ

(a) Феномен когнитивных искажений предвзятости подтверждения (confirmation bias) и его влияние на доверие к информации о здоровье

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(в) Нейрофизиологические корреляты имплицитных процессов

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