Which Exercise Interventions Can Most Effectively Improve Reactive Balance in Older Adults? A Systematic Review and Network Meta-Analysis

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Background: Reactive balance is the last line of defense to prevent a fall when the body loses stability, and beneficial effects of various exercise-based interventions on reactive balance in older adults have been reported. However, their pooled evidence on the relative effects has yet to be described.

Objective: To review and evaluate the comparative effectiveness of various exercise-based interventions on reactive balance in older adults.

Methods: Nine electronic databases and reference lists were searched from inception to August 2021. Eligibility criteria according to PICOS criteria were as follows: (1) population: older adults with the mean age of 65 years or above; (2) intervention and comparison: at least two distinct exercise interventions or one exercise intervention with a no-exercise controlled intervention (NE) compared in each trial; (3) outcome: at least one measure of reactive balance; (4) study: randomized controlled trial. The main network meta-analysis was performed on data from the entire older adult population, involving all clinical conditions as well as healthy older adults. Subgroup analyses stratified by characteristics of participants (healthy only) and reactive balance outcomes (simulated slip or trip while walking, simulated forward falls, being pushed or pulled, and movable platform) were also conducted.

Results: Thirty-nine RCTs (n = 1388) investigating 17 different types of exercise interventions were included in the network meta-analysis. Reactive balance training as a single intervention presented the highest probability (surface under the cumulative ranking (SUCRA) score) of being the best intervention for improving reactive balance and the greatest relative effects vs. NE in the entire sample involving all clinical conditions [SUCRA = 0.9; mean difference (95% Credible Interval): 2.7 (1.0 to 4.3)]. The results were not affected by characteristics of participants (i.e., healthy older adults only) or reactive balance outcomes.
INTRODUCTION

The World Health Organization reported that approximately 28–35% of people aged 65 or above experience at least one fall each year, and the frequency of falls increases with age and frailty level (World Health Organization, 2008). Among various intrinsic risk factors for falls, gait and balance problems have been considered the strongest risk factors (Deandrea et al., 2010; Ambrose et al., 2013). Balance can be mechanistically achieved and maintained by a complex set of sensorimotor control systems including the multisensory (visual, somatosensory, and vestibular system) integration into the central nervous system and the subsequent motor output of the musculoskeletal system (Shumway-Cook and Woollacott, 2017). However, older adults show age-related decline in sensorimotor systems, which in turn increases the risks of falls (Mahoney et al., 2019; Osoba et al., 2019). Given the inherent and inevitable age-related degeneration in sensorimotor systems, it is becoming increasingly clear that in order to prevent potential repercussions, such as aging-related disease, disabilities, injuries, and falls, there is an urgent need for effective interventions to decelerate or even reverse the retrogression in the balance and gait control systems (Kim et al., 2020; Sibley et al., 2021).

In daily life, reactive balance, referred to as the ability to control balance in response to mechanical disturbances, plays a critical role in avoiding and adapting to the complex environments that menace postural stability. The WHO Global Report on Falls Prevention in Older Age reported that factors related to the physical environment, for instance, uneven sidewalks, unmarked obstacles, and slippery surfaces, are some of the most common causes (30–50%) of falls in older adults (World Health Organization, 2008). Notably, slips and trips were the most prevalent causes of falls in regards to circumstances in older adults (Berg et al., 1997). Reactive balance strategies, such as swaying around the ankle or hip joints, taking a reactive step, or reaching to grasp a handhold (Shumway-Cook and Woollacott, 2017), need to be executed promptly so as to avoid falls following a postural perturbation. In the same vein, the balance recovery reactions have also shown age-related differences in older adults vs. young adults and in fallers vs. non-fallers (Alissa et al., 2020; Okubo et al., 2021).

There is a considerable amount of literature on the effects of a variety of interventions on reactive balance, including several systematic reviews and meta-analyses focusing on older adults (Bohm et al., 2015; Lesinski et al., 2015; McCrum et al., 2017; Moore et al., 2019). However, there remain some limitations in the prior syntheses. First, the exercise interventions were limited to only those interventions focused on balance or strength training despite the existence of many studies that use exercises that specifically train reactive balance. Consequently, to the best of our knowledge, none of the previous reviews or meta-analyses have considered comparative effects between different types of exercises and the efficacy of multifaceted exercise interventions with more than one type of exercise on reactive balance. Thus, there is a need for a more comprehensive and inclusive analysis utilizing precise coding of exercise types targeting specific biological systems and functional aspects for better prescriptive guidance (Sibley et al., 2021). Second, the systematic review by Moore et al. (2019) who examined the effectiveness of active physical training interventions on reactive balance did not perform a quantitative synthesis (Moore et al., 2019). Consequently, there remains a lack of pooled evidence on the relative effects of different exercise interventions on reactive balance. Moreover, a conventional pairwise meta-analysis is restricted to a head-to-head comparison of only two different interventions, and thus, RCTs with other types of exercise interventions, that are also effective, can potentially be excluded. To tackle this problem, a network meta-analysis (NMA) is well suited, because it facilitates comparisons of multiple pairs of interventions in one statistical model (Dias et al., 2018).

Therefore, the current study aimed to quantitatively synthesize the available evidence of RCTs in detail using a systematic review and NMA to: (1) combine information from all available randomized comparisons of a set of exercise interventions for reactive balance in older adults; (2) to appraise the relative effects of different exercise interventions on reactive balance; and (3) to determine the ranking of each to provide practical and clinical suggestions to design evidence-based exercise programs for reactive balance. The research question was as follows: “What type of exercise intervention is most effective in improving overall measures as well as each measure of reactive balance in older adults?”

METHODS

The protocol was prospectively registered in the PROSPERO database (CRD42021256638) and conducted in accordance with the PRISMA extension statement for network meta-analysis (Hutton et al., 2015).

Eligibility Criteria

The population of interest included older adults with a sample mean age of 65 years or above with no restriction on the injury or disorder type and with no history of falls studied in various research settings (e.g., community, clinics, and long-term care facilities). Studies were included, if at least two experimental groups participated in each of the different exercise interventions or if there was at least one exercise intervention group...
TABLE 1 | Definitions of exercise types.

| Exercise type                              | Code  | Definitions                                                                 |
|--------------------------------------------|-------|-----------------------------------------------------------------------------|
| Single balance exercise including reactive balance component | SBR   | An intervention including a balance exercise with one or more mechanical postural perturbations given during the exercise |
| Single balance exercise not including reactive balance component | SBNR  | An intervention including a balance exercise without any mechanical postural perturbations |
| Multiple balance exercises including reactive balance component | MBR   | An intervention including more than one type of balance exercise with one or more mechanical postural perturbations given during one of the exercises |
| Multiple balance exercises not including reactive balance component | MBNR  | An intervention including more than one type of balance exercise without any mechanical postural perturbations |
| Unspecified balance exercise              | balUS | Balance exercise without any details given in the original article          |
| Gait training including reactive balance component | gaitR | An intervention including gait training with one or more mechanical postural perturbations given during the exercise |
| Gait training not including reactive balance component | gaitNR | An intervention including gait training without any mechanical postural perturbations |
| Whole body vibration                       | WBV   | Any activity performed on a machine with a vibrating platform               |
| Strength                                   | str   | Exercise that uses the external resistance load (e.g., body weight, resistance bands, machines) to force skeletal muscles contract. |
| Power                                      | pw    | Exercise that applies the maximum amount of force (muscle contraction against a resistance) in the shortest period of time. |
| 3D exercise                                | 3d    | Exercise that requires multi-dimensional movements with a specific name of the exercise (e.g., Yoga, dance, Tai Chi) |
| Flexibility                                | flex  | Exercise that intends to restore or maintain the optimal range of motion (ROM) available to a joint or joints. |
| Functional training                        | FT    | Exercise that utilizes functional activities as the training stimulus that is based on the theoretical concept of task specificity |
| Aerobic                                    | aer   | Exercise aimed at cardiovascular conditioning. It is aerobic in nature and simultaneously increases the heart rate and the return of blood to the heart. |
| No exercise                                | NE    | A group received none of the exercise interventions listed above             |

with a no-exercise control group. Studies involving any non-exercise interventions (e.g., medication, electrical stimulation, or nutritional supplement) were excluded. Details regarding the exercise interventions must have been provided in the reports. The studies must have included at least one reactive balance assessment, which is defined in this study as an assessment using mechanical postural perturbation during a static or dynamic steady-state task. The studies included in this review were restricted to randomized controlled trials (RCTs) and written in the English language.

Search Strategy

The following electronic databases were initially searched by one reviewer (Y.K.) from the inception to February 2021: MEDLINE, EBSCO, CINAHL, SPORTDiscus, PsycINFO, PubMed, WorldCat.org, OpenGrey.eu, and PROQUEST were additionally searched for unpublished trials. To keep this search up to date, an updated search followed in August 2021 by two reviewers (YK and MV). Earlier reviews and bibliographies of included studies were reviewed for additional potentially relevant trials. The combination of the following keywords was employed for the database searches: (aged OR aging OR old* OR elder* OR senior*) AND (exercise OR train* OR activit* OR rehabilitation OR therap* OR physiotherapy OR hydrotherapy OR conditioning OR exertion OR recreation* OR aerobic* OR stretch* OR strengthen* OR walk* OR jog* OR run* OR cycl* OR pilates OR yoga OR tai chi OR ai chi OR dance OR swim*) AND (reactive postural response OR stepping response OR perturbation OR slip perturbation OR reactive balance OR reactive stepping OR protective stepping OR compensatory stepping OR anticipatory postural adjustment* OR compensatory postural adjustment* OR anticipatory postural response* OR compensatory postural response* OR anticipatory adjustment* OR compensatory adjustment* OR postural adaptation* OR postural stabilization OR automatic postural response* OR postural stepping response*) AND (random*).

Study Selection

After exporting the references and removing duplicates, titles and abstracts of records were screened independently by two reviewers (YK and MV) according to the eligibility criteria. Full texts of all potentially relevant trials were subsequently retrieved and reviewed to confirm the final eligible trials. Any disagreements were resolved via consensus, and when any disagreement was elusive, a third reviewer (EB) acted as an arbiter.

Data Extraction and Coding

A total of 46 eligible studies were reviewed and coded in REDCap (https://www.projectredcap.org/) by one reviewer (YK) and confirmed by a second reviewer (M.V.). Any disagreements were resolved via consultation with a third reviewer (EB). The
extracted data included: (1) study characteristics; (2) baseline demographics of participants; (3) exercise interventions; (4) reactive balance outcome measures; and (5) results. Exercise categorizations developed by Howe et al. and Sibley et al. were modified in consideration of the purpose of the current research and applied to the coding (Table 1) (Howe et al., 2011; Sibley et al., 2021).

Means (M) and standard deviations (SD) for all eligible outcomes of reactive balance measures at baseline and post-intervention were extracted for the analysis. Missing data related to eligibility and study outcomes (i.e., data not reported either in a text or on publicly accessible data repositories) were requested to the corresponding authors via email. In the case of no response after one month, a second request was sent, if another month elapsed without response, the data were considered irretrievable. If the requested, but not retrieved data were presented in a graphical format rather than numeric data (e.g., tabular format), Engauge Digitizer 12.1 (http://markummitchell.github.io/engauge-digitizer/) was applied for data digitization and extraction.

Risk of Bias
To ascertain an overall and study-level risk of bias for each trial, a pair of reviewers (YK and MV) independently determined the
### TABLE 2 | Characteristics of participants.

| Study                        | Disease category                                      | Sample size (post-intervention) | Attrition rate (%) | Age (years) |
|------------------------------|-------------------------------------------------------|---------------------------------|--------------------|-------------|
| Allin et al. (2020)          | Healthy                                               | 34 (29)                         | 15                 | 70.4        |
| Arampatzis et al. (2011)     | Healthy                                               | 55 (38)                         | 31                 | 67.7        |
| Arghavani et al. (2020)      | Healthy (fallers: 6 months)                           | 60 (49)                         | 18                 | 69.6        |
| Beiling and Roller (2009)    | Healthy                                               | 23 (19)                         | 17                 | 80.0        |
| Bieryla et al. (2007)        | Healthy                                               | 12 (11)                         | 8                  | 73.3        |
| Bogaerts et al. (2007)       | Healthy                                               | 220 (161)                       | 27                 | 67.1        |
| Cabrera-Martos et al. (2020) | Parkinson’s                                           | 44 (44)                         | 0                  | 76.5        |
| Cherup et al. (2019)         | Parkinson’s                                           | 42 (35)                         | 17                 | 71.2        |
| Chyu et al. (2010)           | Postmenopausal women with osteopenia                  | 61 (53)                         | 13                 | 71.9        |
| Donath et al. (2016)         | Healthy                                               | 59 (48)                         | 19                 | 69.7        |
| Gatts and Woollacott (2007)  | Healthy (balance deficiency without any neurological disorder); Arthritis, back, knee, or hip surgery not excluded. | 22 (19) | 14 | 77.6 |
| Granacher et al. (2006)      | Healthy                                               | 60 (60)                         | 0                  | 66.5        |
| Granacher et al. (2009)      | Healthy                                               | 40 (40)                         | 0                  | 67.0        |
| Hamed et al. (2018)          | Healthy                                               | 63 (47)                         | 25                 | 71.2        |
| Hatzitaki et al. (2009)      | Healthy                                               | 56 (56)                         | 0                  | 70.9        |
| Hu and Woollacott (1994)     | Healthy                                               | 24 (24)                         | 0                  | 75.2        |
| Inacio et al. (2018)         | Healthy                                               | 18 (18)                         | 0                  | 71.9        |
| Jagdhane et al. (2016)       | Healthy                                               | 6 (6)                           | 0                  | 73.3        |
| Kim and Lockhart (2010)      | Healthy                                               | 18 (18)                         | 0                  | NS          |
| Klamroth et al. (2019)       | Parkinson’s                                           | 43 (37)                         | 14                 | 65.3        |
| Lacroix et al. (2016)        | Healthy                                               | 66 (60)                         | 9                  | 72.8        |
| Li et al. (2009)             | Healthy                                               | 50 (40)                         | 20                 | 65.3        |
| Ma et al. (2019)             | Healthy                                               | 33 (24)                         | 27                 | 69.8        |
| Mansfield et al. (2010)      | Healthy (fallers: 5 years)                            | 34 (30)                         | 12                 | 69.7        |
| Marigold et al. (2005)       | chronic stroke                                        | 59 (48)                         | 19                 | 67.8        |
| Morat et al. (2019)          | Healthy                                               | 51 (45)                         | 12                 | 69.4        |
| Ni et al. (2014)             | Healthy                                               | 48 (39)                         | 19                 | 74.2        |
| Ochi et al. (2015)           | Healthy                                               | 20 (20)                         | 0                  | 80.6        |
| Okubo et al. (2019)          | Healthy                                               | 44 (41)                         | 7                  | 72.1        |
| Pamukoff et al. (2014)       | Healthy (some lower extremity mobility dysfunction)    | 20 (15)                         | 25                 | 70.8        |
| Parijat and Lockhart (2012)  | Healthy                                               | 24 (24)                         | 0                  | 72.7        |
| Parijat et al. (2015a)       | Healthy                                               | 24 (24)                         | 0                  | 72.4        |
| Parijat et al. (2015b)       | Healthy                                               | 24 (24)                         | 0                  | 72.4        |
| Pluchino et al. (2012)       | Healthy                                               | 40 (27)                         | 33                 | 72.1        |
| Qutubuddin et al. (2007)     | Parkinson’s                                           | 22 (15)                         | 32                 | 72.8        |
| Reger et al. (2020)          | Healthy                                               | 30 (30)                         | 0                  | 71.0        |
| Rossi et al. (2014)          | Healthy                                               | 46 (46)                         | 0                  | 67.5        |
| Santos et al. (2017)         | Parkinson’s                                           | 40 (40)                         | 0                  | 67.8        |
| Schienstedt et al. (2015)    | Parkinson’s                                           | 40 (32)                         | 20                 | 75.7        |
| Shimada et al. (2003)        | Healthy                                               | 34 (32)                         | 6                  | 80.9        |
| Sohn and Kim (2015)          | Healthy                                               | 18 (18)                         | 0                  | 73.7        |
| Thomas and Kacinski (2016)   | Healthy                                               | 24 (24)                         | 0                  | 67.1        |
| Wang et al. (2019)           | Healthy                                               | 146 (146)                       | 0                  | 72.7        |
| Wolf et al. (1997)           | Healthy                                               | 72 (54)                         | 25                 | 76.9        |
| Wooten et al. (2018)         | Healthy (fallers: 1 year)                             | 30 (16)                         | 47                 | 72.6        |
# TABLE 3 | Summary of exercise interventions.

| Study                  | Dosage | Total duration (week) | Exercise interventions       |
|------------------------|--------|-----------------------|------------------------------|
|                        | Min/session | Time/week | Group1 | Group2 | Group3 |
| Allin et al. (2020)    | 30-60  | 2 | 2 | SBR + gaitR | MBR + gaitNR + str |
| Arampatzis et al. (2011) | 90  | 2 | 14 | MBR | SBR + gaitNR + str |
| Arghavani et al. (2020) | 60  | 3 | 8 | SBR | SBR + gaitNR + str |
| Beling and Roller (2009) | 60  | 3 | 12 | MBR + gaitNR + flex + str | NE |
| Bierly et al. (2007)   | 15     | 1 | 1 | gaitNR | gaitNR |
| Bogaerts et al. (2007) | 40-90  | 3 | 1 year | MBNR + WBV | SBRN + str + flex + aer |
| Cabrera-Martos et al. (2020) | 45  | 3 | 8 | FT | FT + flex |
| Cherup et al. (2019)   | 60     | 2 | 12 | pw | Str |
| Chyu et al. (2010)     | 60     | 3 | 24 | 3d | NE |
| Donath et al. (2016)   | 66     | 2 | 8 | 3d | MBNR |
| Gatts and Woollacott (2007) | 90  | 5 | 3 | 3d | SBRN + flex |
| Gatts (2008)           | 90     | 5 | 3 | 3d | SBRN + flex |
| Granacher et al. (2006) | 60  | 3 | 13 | str | SBRN |
| Granacher et al. (2009) | 60  | 3 | 13 | str | NE |
| Hamed et al. (2018)    | 90     | 2 | 14 | str | SBR |
| Hatzitaki et al. (2009) | 30   | 3 | 4 | SBNR | SBRN |
| Hu and Woollacott (1994) | 60  | 10 sessions (total) | 15 days (total) | SBRN | NE |
| Inacio et al. (2018)   | 15     | 3 | 8 | pw | str |
| Jagdhane et al. (2016) | 60     | 3 | 4 | SBR | NE |
| Kim and Lockhart (2010) | NR   | NR | 8 | str | MBNR |
| Klamroth et al. (2019) | 40     | 2 | 8 | gaitR | gaitNR |
| Lacroix et al. (2016)  | 45     | 3 | 12 | MBNR + str + pw | MBNR + str + pw |
| Li et al. (2009)       | 60     | 4 for 6 weeks, 7 for 10 weeks | 16 | 3d | NE |
| Ma et al. (2019)       | 60     | 2 | 12 | 3d | NE |
| Mansfeld et al. (2010) | 30     | 3 | 6 | SBR | SBRN + flex |
| Marigold et al. (2005) | 60     | 3 | 10 | MBR + gaitNR | SBRN + flex |
| Morat et al. (2019)    | 40     | 3 | 8 | SBR | SBNR |
| Ni et al. (2014)       | 60     | 2 | 12 | 3d | MBNR |
| Ochi et al. (2015)     | 30     | 3 | 12 | MBNR + WBV | SBRN + str |
| Okubo et al. (2019)    | 40     | 3 | 1 | gaitR | gaitNR |
| Pamukoff et al. (2014) | 60     | 3 | 6 | pw | str |
| Parjat and Lockhart (2012) | 40  | 1 | 1 | gaitR | gaitNR |
| Parjat et al. (2015a)  | 35–55  | 1 | 1 | gaitR | gaitNR |
| Parjat et al. (2015b)  | 35–55  | 1 | 1 | gaitR | gaitNR |
| Piuchino et al. (2012) | 60     | 2 | 8 | MBNR + gaitNR | 3d | MBNR |
| Qutubuddin et al. (2007) | 30   | 2 | 4 | balUS | MBNR + gaitNR |
| Rieger et al. (2020)   | NS     | 1 | 1 | gaitR | gaitNR |
| Rossi et al. (2014)    | 40     | 3 | 6 | SBNR | NE |
| Santos et al. (2017)   | 60     | 2 | 8 | str + flex | MBR + gaitNR |
| Schienstedt et al. (2015) | 60  | 2 | 7 | str | MBR |
| Shimada et al. (2003)  | 40     | 2–3 | 12 | MBNR | gaitNR |
| Sohn and Kim (2015)    | 60     | 3 | 8 | str | balUS |
| Thomas and Kalkinski (2016) | 70  | 2 | 6 | MBNR | NE |
| Wang et al. (2019)     | 30     | 1 | 1 | gaitR | gaitNR |
| Wolf et al. (1997)     | 60     | 1–2 | 15 | MBR | NE |
| Wooten et al. (2018)   | 60     | 3 | 6 | MBNR | 3d |

SBR, Single balance exercise including reactive balance component; SBRN, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; balUS, Unspecified balance exercise; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; FT, Functional training; flex, Flexibility; aer, Aerobic; NE, No exercise.
FIGURE 2 | Summary of the distributions of the reviewers’ judgements across the studies for each risk of bias domain.

FIGURE 3 | Network geometry of the included exercise programs. Each line indicates a direct comparison of two different exercise programs. The thickness of the edge is proportional to the number of direct comparisons in the network. Different exercise types combined in one program are connected via underscores. The blue triangles refer to multi-arm trials comprised of three exercise programs in the nodes. SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; balUS, Unspecified balance exercise; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; flex, Flexibility; aer, Aerobic; NE, No exercise.
bias arising from the following domains using the Cochrane risk of bias tool (RoB 2): (1) randomization process; (2) deviations from the intended interventions; (3) missing outcome data; (4) measurement of the outcome; and (5) selection of the reported result (Sterne et al., 2019). Each domain was assigned a judgement of “low risk,” “some concerns,” or “high risk.” Disagreements were resolved through discussion or referral to a third reviewer (EB).

**Data Synthesis and Statistical Analysis**

Considering indeterminate baseline similarities of reactive balance measures in several studies, change values from baseline to post-intervention were calculated or directly extracted from the published data. If there were more than one post-intervention measure (e.g., post-intervention and follow-up), only the data immediately following the termination of the intervention phase was used. SDs for changes from baseline (pre) to post-intervention (post) were calculated using the following formula (Higgins et al., 2019):

\[
SD_{\text{change}} = \sqrt{SD_{\text{pre}}^2 + SD_{\text{post}}^2 - 2 \times \text{Corr} \times SD_{\text{pre}} \times SD_{\text{post}}}
\]

Corr in the SD\text{change} equation is the correlation coefficient describing how similar the pre and post-interventions were across participants. When the correlation coefficient was not reported, it was set as 0.5 (Fu et al., 2008; Bruderer-Hofstetter et al., 2018; Lai et al., 2018; Wu et al., 2021). In the case of a lower score signifying better performance in reactive balance measures (e.g., reaction time), scale directions were adjusted by multiplying \(-1\) to the M\text{change} data, which led to a greater effect size indicating an improvement. Missing SDs were imputed from standard errors (SE), 90 or 95% confidence intervals (CI). Using the M\text{change} and SD\text{change} data, standardized mean differences (SMD) and standard errors (SE) were calculated.

To include multi-arm trials, two approaches were adopted to avoid a unit-of-analysis error (Rücker et al., 2017; Higgins et al., 2019):
et al., 2019). First, all relevant experimental intervention groups composed of the same categories of exercises were combined into a single group. This step enabled a single pairwise comparison between a combined group and a comparison group in each study. Second, in the case of heterogeneous exercise types across all intervention groups, we included all relevant comparisons as a series of two-arm comparisons and reflect the fact that comparisons within multi-arm studies are correlated (Schwarzer et al., 2015). Accordingly, adjusted SEs of the two-arm meta-analysis and indirect comparisons (i.e., from network meta-analysis) due to data not being reported.

The analyses utilized a burn-in period (50,000 iterations) and a follow-up period (100,000 iterations) to minimize bias of initial values when the chain reached its target distribution (Brooks and Gelman, 1998). The convergence was assessed using the trace plot, density plot, and Brooks-Gelman-Rubin diagnostic statistics (Brooks and Gelman, 1998).

The overall geometry of the network was presented in a network graph. Based on Bayesian posterior rank probabilities, the ranking of exercise interventions was estimated using a hierarchical tool, the surface under the cumulative ranking curve (SUCRA) score, measured on a scale from 0 (theoretically the worst) to 1 (the best). In addition, a network forest plot was produced with the “no exercise (NE)” as a reference intervention. The posterior distribution of the SMDs was reported using the mean differences for the reference intervention with 95% credible intervals (CrI), which indicate that there is a 95% probability that the unobserved (unknown) effect estimates would fall within the intervals (Hespanhol et al., 2019). If a 95% CrI contains zero (i.e., null effect representing the null hypothesis), the effect can be considered statistically insignificant (Hespanhol et al., 2019). The relative effects with 95% CrI of all pairs of exercise interventions were reported in a matrix. Consistency, which is the most important assumption underlying a NMA and indicates agreement between direct and indirect estimates in the network (Salanti et al., 2014), was checked using the node-splitting analysis. The original intention of the first subgroup analysis was to conduct a network meta-analysis stratified by characteristics of participants (i.e., healthy and disease-specific). However, due to the insufficient number of exercise interventions to establish a network in each disease category (e.g., Parkinson’s disease), the first subgroup analysis was performed by the inclusion of studies with healthy older adults only (78% of all studies) who did not have any disease, injury, or disability at the time of the studies. The second subgroup analysis was conducted by grouping the outcome measures by the types of reactive balance outcomes: (1) simulated slip or trip while walking; (2) simulated forward falls; (3) being pushed or pulled; (4) movable platform; and (5) balance test battery. A sensitivity analysis was carried out using a frequentist framework NMA to appraise the robustness of the results. Sources of statistical heterogeneity and small study bias were not explored due to the insufficient number of trials (k ≤ 5) for each comparison. All data syntheses and statistical analyses were conducted using “Gemtc” (version 1.0-1), “rjags” (version 4-10), and “netmeta” (version 1.4-0) packages in R (Version 4.1.0, R Foundation for Statistical Computing, Vienna, Austria).

### RESULTS

#### Study Selection

A total of 7,394 records were retrieved from electronic databases and two additional records were obtained from other sources, of which 384 studies remained after removing duplicates and screening titles and abstracts. Based on the full-text screening, 46 records fulfilled the eligibility criteria and thus were included for qualitative analysis (i.e., systematic review), whereas seven studies were additionally excluded from the quantitative analysis (i.e., network meta-analysis) due to data not being reported.

#### Table 4: Ranking of Exercise Interventions

| SMC | Exercise Description | SUCRA Score | SMC | Exercise Description | SUCRA Score |
|-----|----------------------|-------------|-----|----------------------|-------------|
| 1   | SBR                  | 0.90        | 1   | SBR                  | 0.94        |
| 2   | pw                   | 0.67        | 2   | pw                   | 0.70        |
| 3   | gaitR                | 0.62        | 3   | gaitR                | 0.64        |
| 4   | SBNR + flex          | 0.58        | 4   | SBNR + flex          | 0.61        |
| 5   | MBR + gaitNR         | 0.58        | 5   | MBR + gaitNR         | 0.60        |
| 6   | str + flex           | 0.55        | 6   | str + flex           | 0.57        |
| 7   | balUS                | 0.49        | 7   | balUS                | 0.49        |
| 8   | str                  | 0.49        | 8   | str                  | 0.49        |
| 9   | SBNR                 | 0.46        | 9   | SBNR                 | 0.46        |
| 10  | MBNR                 | 0.46        | 10  | MBNR                 | 0.45        |
| 11  | MBR                  | 0.45        | 11  | MBR                  | 0.44        |
| 12  | MBNR + gaitNR        | 0.44        | 12  | MBNR + gaitNR        | 0.43        |
| 13  | MBNR + WBV           | 0.40        | 13  | MBNR + WBV           | 0.38        |
| 14  | SBNR + str           | 0.40        | 14  | SBNR + str           | 0.37        |
| 15  | gaitNR               | 0.39        | 15  | gaitNR               | 0.37        |
| 16  | 3d                   | 0.35        | 16  | 3d                   | 0.33        |
| 17  | NE                   | 0.27        | 17  | NE                   | 0.23        |

SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; balUS, Unspecified balance exercise; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; flex, Flexibility; aer, Aerobic; NE, No exercise.
and not irretrievable (Kim and Lockhart, 2010; Okubo et al., 2019; Wang et al., 2019), exercise types not included in the network (Allin et al., 2020; Cabrera-Martos et al., 2020), exercise intervention included in only one trial (Lacroix et al., 2016), and no continuous data reported (Beling and Roller, 2009), resulting in a total of 39 studies for NMA. The schematic flow chart for the selection process is presented in Figure 1, and all included studies are listed in Supplementary Table 1.

Characteristics of Included Studies
The eligible studies represented a total of 1,745 older adults, included in both pre and post-intervention analyses, with the mean age of 71.9 ± 3.9 years (ranged from 65.3–80.9 years). The majority of the studies exclusively included community-dwelling healthy older adults (k = 36). Ten studies reported on older adults selected for a specific disease or medical condition, such as Parkinson’s disease (k = 6), post-surgical interventions for knees, hips, or backs (k = 2), postmenopausal women with osteopenia (k = 1), and chronic stroke (k = 1).

The duration and frequency of the exercise interventions ranged from 1 week to 1 year, 1–5 sessions/week, and 15–90 min/session. Of the 46 studies, 16 executed multicomponent (i.e., multifaceted) exercise interventions in at least one group. Reactive balance was assessed before and after the exercise interventions by use of laboratory-induced slip, trip, and falls, external impacts (e.g., pulling or pushing a body part), platform translation, and treadmill perturbation (e.g., rapid change of the speed) while participants were performing a steady-state task, such as standing or walking. Twenty studies provided training with a postural perturbation while standing or walking, and 11 of which implemented task-specific training (i.e., comparable reactive balance task included in the assessment and training) (Wolf et al., 1997; Bieryla et al., 2007; Beling and Roller, 2009; Mansfield et al., 2010; Parijat and Lockhart, 2012; Jagdhane et al., 2016; Morat et al., 2019; Okubo et al., 2019; Wang et al., 2019; Arghavani et al., 2020; Rieger et al., 2020). The characteristics of the participants and exercise interventions are presented in Tables 2, 3, respectively. Outcome measures and main findings are summarized in Supplementary Table 2.

Risk of Bias
The summary of the risk of bias assessment across all included studies is presented in Figure 2. Detailed results of the assessment are reported in Supplementary Table 3. Overall, the majority of

**FIGURE 5** | Forest plot of the relative effects of exercise interventions with a no-exercise as a reference group. SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; balUS, Unspecified balance exercise; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; flex, Flexibility; aer, Aerobic; NE, No exercise.
outcomes were at some concerns (50%) and high risk (48%), and only one study was rated as at low risk. Missing outcome data (46%) was the most influential source of high risk of bias. Selection of the reported result (83%), randomization process (76%), and deviations from intended interventions (61%) were also common sources of bias.

**Network Meta-Analysis**

Data from a total of 39 studies ($n = 1388$, age = 71.5 ± 3.9 years) were included in the NMA. Of the 15 exercise types reported in Table 1, 14 types were included in the NMA as functional training was implemented in only one study and consequently included in a disconnected network (Cabrera-Martos et al., 2020). There were 11 multi-arm trials, and three of which consisted of two groups sharing the same exercise type and the third group with another type (Hatzitaki et al., 2009; Ni et al., 2014; Lacroix et al., 2016); thus, data in these two groups were combined into a single group. Two exercise groups in studies by Gatts and Woollacott (2007) and Gatts (2008), str and NE groups in studies by Granacher et al. (2006, 2009), and two exercise groups in studies by Parijat et al. (2015a,b) shared the same participants, respectively. Thus, each of the aforementioned pairs of studies was combined as a single study in NMA. Overall, 17 exercise interventions with either single or multiple exercise components were included in the NMA. The geometric distribution of the network is depicted in Figure 3. When a study involves a trial arm with a combination of the pre-categorized exercise types, the combination was considered as another distinct exercise intervention.

Estimates of all exercise programs against all others in NMA were reported in a matrix (Figure 4). In the 17 exercise programs, SBR displayed the highest probability of being the most effective exercise intervention (SUCRA score = 0.90) for improving reactive balance, followed by pw (SUCRA score = 0.67) and gaitR (SUCRA score = 0.62) (Table 4).

**FIGURE 6** Network geometry of the included exercise programs in healthy older adults: Each line indicates a direct comparison of two different exercise programs. The thickness of the edge is proportional to the number of direct comparisons in the network. Different exercise types combined in one program are connected via underscores. The blue triangles refer to multi-arm trials comprised of three exercise programs in the nodes. SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; NE, No exercise.
The relative treatment effect estimates of each exercise program with the no-exercise program being the mutual contrast for comparison are presented in a forest plot (Figure 5). SBR, pw, and gaitR demonstrated the largest mean difference vs. NE; however SBR only demonstrated a statistically significant difference when compared to the no-exercise program (mean difference = 2.7, 95% CrI = 1.0 to 4.3). The trace plot, density plot, and Brooks-Gelman-Rubin diagnostic statistics showed good convergence, which signifies our data has converged to a reasonable distribution. Relatively reliable evidence was derived from the statistical consistency between direct and indirect evidence demonstrated by the node-splitting model (p > 0.05). According to the sensitivity analysis using a Frequentist framework of NMA, the ranking based on the P-scores showed identical results (Table 4). The results suggest that our main findings regarding the relative effectiveness of each exercise intervention are robust for future decisions.

Subgroup Analyses
In the subgroup analysis for healthy older adults (k = 29, n = 1120, age = 71.5 ± 3.7 years), effects of 12 exercise programs were compared (Figure 6).

According to the SUCRA scores, SBR was the highest-ranked exercise program (0.90), followed by pw (0.71), which was consistent with the ranking in the complete sample (Table 5). The other exercise programs ranked slightly differently from the NMA for the complete sample; however, the rankings based on the SUCRA scores were consistent with those estimated by P-scores in the frequentist framework (Table 5). The relative effects of all exercise interventions compared to NE were presented in Figure 7. A relative effect matrix was additionally created for all comparisons in the healthy older adults (Figure 8). Too few trials in other disease categories (Parkinson’s disease: 6 trials, arthritis: 2 trials, osteopenia: 1 trial) and types of exercise interventions were available to establish a network in each category and conduct further disease-specific subgroup analyses.

For the second subgroup analysis regarding the types of reactive balance outcomes, the first type (simulated slip or trip while walking) was analyzed for gaitR vs. gaitNR using a multilevel MA due to insufficient trials in other treatment comparisons (i.e., only one direct comparison). The second (simulated forward falls), third (being pushed or pulled), and fourth (movable platform) types were analyzed using NMA. The fifth type (balance test battery) was not analyzed due to the insufficient number of exercise interventions and direct comparisons to establish a network. When a slip or trip was simulated while walking, participants showed greater improvements in measures of balance recoveries after gaitR training vs. gaitNR training (SMD = 0.60; 95% CI, 0.33 to 0.88). In other types, SBR presented the first or second highest probability of being the best intervention for improving each reactive balance task. The ranking and relative effects of each exercise vs. NE are reported in Table 6 and Figure 9, respectively.

**TABLE 5** Ranking of exercise interventions in healthy older adults.

| Bayesian framework | SUCRA score | Frequentist framework | P-score |
|-------------------|-------------|-----------------------|---------|
| Ranking | Exercise | | Ranking | Exercise | |
| 1 | SBR | 0.90 | 1 | SBR | 0.95 |
| 2 | pw | 0.71 | 2 | pw | 0.76 |
| 3 | str | 0.52 | 3 | str | 0.53 |
| 4 | gaitR | 0.52 | 4 | gaitR | 0.52 |
| 5 | SBNR | 0.50 | 5 | SBNR | 0.52 |
| 6 | MBR | 0.47 | 6 | MBR | 0.47 |
| 7 | MBNR | 0.46 | 7 | MBNR | 0.46 |
| 8 | MBNR + WBV | 0.43 | 8 | MBNR + WBV | 0.41 |
| 9 | SBNR + str | 0.42 | 9 | SBNR + str | 0.41 |
| 10 | gaitNR | 0.40 | 10 | gaitNR | 0.37 |
| 11 | 3d | 0.35 | 11 | 3d | 0.32 |
| 12 | NE | 0.32 | 12 | NE | 0.28 |

SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; gaitR, Gait training including reactive balance component; gaitNR, Gait training not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; NE, No exercise.

**DISCUSSION**
To our knowledge, this study is the first NMA to determine which type of exercise intervention is most effective in improving reactive balance in older adults. In this study, we compared the effects of commonly used exercise interventions on reactive balance in older adults. The NMA was used to analyze the data of 39 RCTs including 1,388 participants, which revealed that older adults receiving a balance exercise with a reactive balance component showed the most improvements in reactive balance, followed by power training (second) and gait training with a reactive balance component (third) among 17 different exercise interventions.

The results of this current study highlight the importance of applying the principle of specificity to training interventions designed to improve reactive balance. This is consistent with the hypothesis put forth by Grabiner et al. (2014) suggesting that task-specific perturbation training is superior to conventional exercise approaches in improving reactive balance capacity and thus preventing falls (Grabiner et al., 2014). Of the 46 trials in the current study, there were 20 trials including at least one exercise intervention with a reactive balance component, and ten of which utilized the same parameters of postural perturbations during the training and assessment (i.e., task-specific reactive balance training) (Wolf et al., 1997; Bieryla et al., 2007; Beling and Roller, 2009; Mansfield et al., 2010; Parijat and Lockhart, 2012; Morat et al., 2019; Okubo et al., 2019; Allin et al., 2020; Arghavani et al., 2020; Rieger et al., 2020). This latter point is especially important given that a specific type of reactive balance exercise has no, or at most a limited transfer effect on non-trained reactive balance tasks (Kümmel et al., 2016; Harper et al., 2021). The cognitive processes, muscle synergies, and succeeding kinematic strategies to counteract the perturbation...
are entirely determined by the parameters of the perturbations, such as type, magnitude, direction, and the point of application (Winter et al., 1990; Grabiner et al., 2014; Chen et al., 2017), and reactive balance improves in the tasks that are specifically trained with the same parameters. That is one of the reasons SBR showed greater improvements than other types of exercises. It seems reasonable to speculate that if all 20 trials used the exact same training and assessment tasks, the performance gains in reactive balance would be even greater. However, the estimates in the second subgroup analyses (i.e., types of reactive balance outcomes) regarding high SUCRA scores of SBR should be interpreted with caution given the small number of trials and several wide credible intervals.

Repeated exposure to specific learning environments, therefore, leads to specific motor adaptation and learning. Motor adaptation is a learning process in which the nervous system learns how to predict and cancel impacts of a novel environment (e.g., perturbation), and ultimately maximize performance in that environment (Izawa et al., 2008). The cerebral cortex plays a key role in the acquisition and facilitation of balance recovery skills (Beck et al., 2007; Bolton, 2015). Through repeated exposure to a postural perturbation, our sensorimotor system learns (e.g., procedural learning) internal models for the sensorial prediction and motor commands and uses the learned models for an efficient and optimized movement plan (Izawa et al., 2008), ultimately improving compensatory reactions in older adults (Bohm et al., 2015; König et al., 2019). If mechanical perturbations transpire in consistent patterns with regards to the timing, magnitude, type, and direction, those who have undergone training using the same perturbation system could employ proactive (anticipatory) postural adjustments (i.e., feedforward control in anticipation of or before a postural perturbation) (Bhatt et al., 2006; Aruin et al., 2017; Curuk et al., 2020). Utilization of proactive postural adjustments, facilitated by the repeated exposure to the perturbation, significantly reduces the need for compensatory adjustments after a perturbation (Kanekar and Aruin, 2014); thus, the predictability regarding the perturbation and reactive balance task ultimately impacts greater adaptability and controllability. Such motor training is capable of altering corticospinal excitations and reorganizing motor maps and synaptic changes in the cerebral cortex, which ultimately facilitates the acquisition of a specific balance recovery skill (Beck et al., 2007; Grabiner et al., 2014), and the neuroplastic changes after training offer revealing clinical insights. However, when the patterns of a perturbation unpredictably change, the proactive postural adjustments, that are strictly relying on prior experience, can be deteriorated, which may compromise application to real-world falls where people rarely know in advance how and when they will get perturbed. Thus, a perturbation during training needs to be offered in various patterns to maximize the unpredictability and prepare older adults for the unpredictable nature of real-world falls (Harper et al., 2021). Further, to promote motor adaptation and learning, the elements of the training regimen should be properly determined first, and the challenge should be increased by adjusting the parameters of the perturbation, complexity of the context, and cognitive processing demands (Harper et al., 2021).
The greatest effect of SBR and relatively less effective multicomponent exercise interventions can be further scrutinized via several critical principles of exercise training including volume, intensity, and frequency. Training volume is largely determined by the time commitment (duration) of the training. However, the total duration and frequency of the interventions are broadly ranged across the included studies as previously reported, and the average duration of each training session was 52.2 ± 19.7 min. If an intervention included multiple types of exercises in a single session, the intervention may lack the critical time needed to focus on reactive balance training. According to Burgomaster et al. (Burgomaster et al., 2008), low-volume, high-intensity training and high-volume, low-intensity training induce comparable changes in selected whole-body and skeletal muscle adaptations when the frequencies and the total durations are identical (Burgomaster et al., 2008; Hawley, 2008). Thus, if lack of time is a barrier to satisfying the need for reactive balance training, the intensity aspect of the training should be considered as a way to compensate for the deficit and induce targeted changes in reactive balance. Further, it is encouraging that Bhatt and Pai have demonstrated significant improvements in reactive balance performance after a single high-intensity training with task-specific postural perturbations (Bhatt and Pai, 2009). This is particularly noteworthy given that such minimal training effects were retained for several months when properly selecting the intensity and specificity of the training despite the relatively small total volume. Thus, future trials may wish to take account of the aforementioned factors, including specificity, volume, and intensity of the training to maximize the time-effective transfer to real-world scenarios.

Lastly, given the high ranking of power training, the probable inter-relation with reactive balance is clinically notable. In situations where a mechanical perturbation is applied and a fall begins, the rate of torque development in the lower or upper extremity joints with intersegment coordination has
TABLE 6 | Ranking of exercise interventions in each reactive balance outcome category.

| Ranking | Exercise | SUCRA score | Ranking | Exercise | SUCRA score | Ranking | Exercise | SUCRA score |
|---------|----------|-------------|---------|----------|-------------|---------|----------|-------------|
| 1       | MBNR + WBV | 0.77        | 1       | SBR      | 0.73        | 1       | SBR      | 0.79        |
| 2       | SBR       | 0.65        | 2       | SBNR + flex | 0.63     | 2       | MBR      | 0.75        |
| 3       | Str       | 0.64        | 3       | 3d       | 0.35        | 3       | pw       | 0.72        |
| 4       | SBNR + str | 0.52       | 4       | NE       | 0.30        | 4       | baUS     | 0.60        |
| 5       | pw        | 0.39        | 5       | str      | 0.58        | 5       | MBR      | 0.54        |
| 6       | MBR       | 0.39        | 6       | MBR + gaitNR | 0.48    | 6       | MBR      | 0.48        |
| 7       | NE        | 0.14        | 7       | MBNR     | 0.48        | 7       | MBNR     | 0.43        |
| 8       | MBR + gaitNR | 0.43    | 8       | SBNR     | 0.43        | 8       | MBR      | 0.43        |
| 9       | SBNR + flex | 0.43     | 9       | SBR      | 0.43        | 9       | SBNR     | 0.43        |
| 10      | 3d        | 0.30        | 10      | NE       | 0.14        | 10      | 3d       | 0.30        |
| 11      | MBNR + WBV | 0.25       | 11      | NE       | 0.14        | 11      | MBNR     | 0.25        |
| 12      | NE        | 0.14        | 12      | NE       | 0.14        | 12      | NE       | 0.14        |

A. Simulated forward falls, B. Being pushed or pulled, C. Movable platform. SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; WBV, Whole body vibration; Str, Strength; pw, Power; 3d, 3D exercise; NE, No exercise.

been considered as a critical determinant of balance recovery by taking a step or reach to grasp (Madigan, 2006). Aging inherently brings a loss of motor neurons, associated with apoptosis, and reduction and denervation of muscle fibers, specifically related to type II muscle fibers. These changes lead to a decrease in the muscles’ capacity to produce maximum muscle strength, power, and rate of force development (Aagaard et al., 2010). In general, fallers generate less muscle power than non-fallers, and older adults generate less power than young adults (Madigan, 2006; Perry et al., 2007). By utilizing the comparability between muscle power and reactive balance, such as forceful and controlled movements with high velocity, all power training groups in the current analysis demonstrated improvements in measures of reactive balance. There are a handful of studies investigating the correlations between muscle power and reactive balance performances (Muehlbauer et al., 2015); however, the effectiveness of power training on reactive balance has been explored only in a few, recent trials (Pamukoff et al., 2014; Inacio et al., 2018; Cherup et al., 2019). The results of this study may have implications for future directions in assessing the relationship between muscle power and reactive balance.

Given that the vast majority of falls occur while walking (Tinetti et al., 1988; Berg et al., 1997; Li et al., 2006; Kelsey et al., 2012; Robinovitch et al., 2013), training to counter postural perturbations while walking is imperative. However, the ranking of gaitR was relatively lower than SBR. Because gaitR was only compared with gaitNR in the network, the ranking was dominantly determined by indirect evidence. Standard errors from indirect evidence are greater than those from direct evidence, which represents the lower accuracy of an estimate (Higgins et al., 2019). Thus, the indirect evidence should be interpreted with caution, and more RCTs with direct comparisons between gaitR and other exercises may guarantee more accurate posterior distributions and the ranking of gaitR.

Clinical Implications

Considering the findings of this study, it would be advisable for clinicians to preferentially include reactive balance training in line with specifically targeted context, direction, and type of postural perturbations, and power training as a secondarily or complementary approach to improve reactive balance in older adults irrespective of their clinical classifications. Multicomponent exercise interventions not including a reactive balance component may not bring as marked changes in reactive balance as a single reactive balance training does, whereas they still have benefits regarding general health and physical functioning. The possibility of task-specific training adaptations with balance training using external mechanical perturbations has far-reaching clinical and research implications. In fact, beyond simply training one specific type of balance reaction (e.g., a slip), future trials may wish to include multiple types of reactive...
FIGURE 9 | Forest plots of the relative effects of exercise interventions with no-exercise as a reference group in each reactive balance outcome category. (A) Simulated forward falls, (B) Being pushed or pulled, and (C) Movable platform. SBR, Single balance exercise including reactive balance component; SBNR, Single balance exercise not including reactive balance component; MBR, Multiple balance exercises including reactive balance component; MBNR, Multiple balance exercises not including reactive balance component; WBV, Whole body vibration; str, Strength; pw, Power; 3d, 3D exercise; NE, No exercise.
balance tasks in various simulated contexts that are likely to occur in daily life and appraise the generalizability and ecological validity of the trained tasks from a long-term perspective. Moreover, the addition of power training may synergize the effects on functional reflex activities as well as general functional capabilities needed for daily tasks and reducing falls in older adults.

Strengths and Limitations
One of the major strengths of the current study is the use of a NMA. The notable advantage of a NMA over a conventional pairwise meta-analysis is the ability to allow for indirect comparisons, accounting for the effects of multiple interventions in a single statistical model (Schwarzer et al., 2015). Thus, a NMA concurrently summarizes both direct and indirect comparisons between multifarious interventions and enables more complex statistical models and broader interpretation. Random-effects models attempt to generalize the results beyond the trials included in the NMA with an assumption that the selected trials are random samples from a larger population (Cheung et al., 2012). Accordingly, the use of a NMA with a random-effects model in this study enhances the applicability and generalizability of study findings. It should be noted that in general, the indirect estimates tend to have greater variance than direct estimates, and the reliability of the indirect estimates are influenced by the number of direct estimates in the network (Dias et al., 2018). Future meta-analyses may wish to assess publication bias and heterogeneity with a greater number of trials in each direct comparison.

The interpretations of the results in the current study are limited due to small sample sizes and the existence of the probable risk of bias in the included studies. For example, only two trials included more than 100 total participants (Bogaerts et al., 2007; Wang et al., 2019). Furthermore, there was heterogeneity in participants and exercise interventions. For example, there were several distinct disease groups, and the frequency and duration were set differently for various exercise interventions pooled together. With further trials, future reviews may wish to break down the analyses on the basis of hypothetical effect modifiers, such as detailed age and disease groups, baseline functional capacities, or dosage of intervention, for more specific clinical decisions. Also, the low number of trials per comparison precluded investigating sources of publication bias and heterogeneity, and the overall risk of bias was appraised as some concern or high-risk level. Thus, a comprehensive search of published and unpublished works of literature with a paired screening process was conducted to guarantee all available literature was identified to reduce the potential risk of publication bias. Considering the number of trials per each direct comparison, sample sizes, and overall risks of bias, the results of our analyses may as such guide future research. Despite the aforementioned limitations, we believe that this systematic review with a NMA shed light on better understanding effective interventions for reactive balance in older adults via more comprehensive and inclusive analyses of available literature.

CONCLUSIONS
In conclusion, our NMA indicates that SBR, which simulates a real-life fall scenario and induces a specific balance recovery, is generally more efficacious in improving reactive balance than any other exercise intervention in older adults. Importantly, power training also appears to have greater impacts on reactive balance than other exercise interventions. Our results highlight the importance of task-specific exercise interventions with respect to the targeted postural perturbation and reactions. More trials with high methodological quality, low risk of bias, larger samples, and older adults with a specific disease or disability need to be conducted to construct a comprehensive literature basis, which would facilitate a more thorough NMA. The findings of this study could be used to design exercise-based interventions for improving reactive balance in older adults.

DATA AVAILABILITY STATEMENT
The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author/s.

AUTHOR CONTRIBUTIONS
YK, MV, and EB performed the search, selection, data extraction, and risk of bias assessments. YK performed all data analyses, drafted the manuscript, and attests that all listed authors meet the authorship criteria and that no others meeting the criteria have been omitted. All authors participated in the conception and design of the study, provided feedback on early and advanced drafts of the manuscript, critically revised for important intellectual content, approved the final version of the manuscript, and agreed to be accountable for all aspects of the work.

SUPPLEMENTARY MATERIAL
The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fnagi.2021.764826/full#supplementary-material

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