REVIEW

Extracellular Vesicles for the Diagnosis of Parkinson's Disease: Systematic Review and Meta-Analysis

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ABSTRACT: Parkinson's disease (PD) biomarkers are needed by both clinicians and researchers (for diagnosis, identifying study populations, and monitoring therapeutic response). Imaging, genetic, and biochemical biomarkers have been widely studied. In recent years, extracellular yesicles (EVs) have become a promising material for biomarker development. Proteins and molecular material from any organ, including the central nervous system, can be packed into EVs and transported to the periphery into easily obtainable biological specimens like blood, urine, and saliva. We performed a systematic review and meta-analysis of articles (published before November 15, 2022) reporting biomarker assessment in EVs in PD patients and healthy controls (HCs). Biomarkers were analyzed using random effects meta-analysis and the calculated standardized mean difference (Std.MD). Several proteins and ribonucleic acids have been identified in EVs in PD patients, but only α-synuclein (aSyn) and leucine-rich repeat kinase 2 (LRRK2) were reported in sufficient studies (n = 24 and 6, respectively) to

perform a meta-analysis. EV aSyn was significantly increased in neuronal L1 cell adhesion molecule (L1CAM)-positive blood EVs in PD patients compared to HCs (Std.MD = 1.84, 95% confidence interval = 0.76-2.93. P = 0.0009). Further analysis of the biological sample and EV isolation method indicated that L1CAM-IP (immunoprecipitation) directly from plasma was the best isolation method for assessing aSyn in PD patients. Upcoming neuroprotective clinical trials immediately need peripheral biomarkers for identifying individuals at risk of developing PD. Overall, the improved sensitivity of assays means they can identify biomarkers in blood that reflect changes in the brain. CNS-derived EVs in blood will likely play a major role in biomarker development in the coming years. © 2023 The Authors. Movement Disorders published by Wiley Periodicals LLC on behalf of International Parkinson and Movement Disorder Society.

Key Words: exosomes; biomarker; Parkinson's disease; L1 cell adhesion molecule; α-synuclein; plasma

Introduction

Timely and correct diagnosis of Parkinson's disease (PD) is challenging for clinicians and researchers. By

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the time motor symptoms appear, there has been an irreversible loss of the majority of central dopaminergic neurons.¹ Thus far, clinical trials of potentially neuroprotective agents in early motor PD have yielded

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negative results. Early and accurate diagnosis of patients with PD is needed for the development of diagnostic tools and for epidemiological and treatment studies. Such tools will also improve clinical care by providing a larger window for neuroprotective and therapeutic interventions-the prodromal phase of PD is up to 20 years.²

The definite diagnosis of PD can be confirmed only by neuropathological postmortem examination, and the accuracy of the clinical motor diagnosis is low, especially in the first 5 years of the disease.³⁻⁵ Diagnostic accuracy increases with disease duration, mainly due to the development in motor fluctuations.⁶

Research efforts have focused on developing objective tests and biomarkers to facilitate earlier and more accurate diagnosis of PD and to support clinical trials. Despite advances in imaging tools,^{7,8} and in genetic⁹⁻¹¹ and biochemical biomarkers,¹²⁻¹⁵ the diagnostic validity of PD has not improved. The recent advancement in using seed amplification assays (SAA) to detect pathologic α -synuclein (aSyn) seeds is currently well established in the cerebrospinal fluid (CSF)¹⁶ and broadly under development for peripheral biospecimens, but the assay is not yet optimized to reflect disease progression.¹⁷ Therefore, medical practitioners urgently need a biomarker that can be effectively used, enable early diagnosis, and reflect disease progression and response to interventions.

Several CSF biomarkers have been studied.^{12,13} The invasiveness of CSF collection makes it difficult to use for larger population screenings to identify, for example, those at risk of developing PD in the future. Peripheral, minimally invasive, or non-invasive biological specimens like blood, urine, and saliva have also been used for the development of biomarkers, but these have yet to be validated in clinical practice.¹⁸ In recent years, brain-derived extracellular vesicles (EVs) isolated from samples have garnered increasing attention as possible biomarkers: they are considered an attractive target with unique properties and a cargo rich in signaling molecules from the central nervous system (CNS).¹⁹⁻²¹

EVs are secreted by cells in both physiological and pathological conditions.²² They are single-membrane vesicles of endosomal origin, ~30 to 200 nm in diameter, with a characteristic cup-shaped morphology observed under the microscope. They can carry proteins, deoxyribonucleic acids, ribonucleic acids (RNA), lipids, and glycoconjugates.²³ EVs contain the majority of proteins from plasma membranes, endosomes and cytosol, and some proteins from the nucleus, mitochondria, and Golgi body from the cells of origin.²⁴ In addition to proteins, various RNA subtypes, namely mRNAs, microRNA (miRNAs), long non-coding RNAs, and circular RNAs, have been identified in EVs, uncovering new mechanisms in the genetic exchange between cells and intercellular transcriptional and

translational regulation.²⁵⁻²⁹ These vesicles are secreted by neural, muscle, and epithelial cells. They are vital in cell-to-cell communication and can be found in biofluids like CSF, blood, saliva, and urine.³⁰⁻³³ EVs can cross the blood-brain barrier (BBB) in both directions and can, therefore, act as mediators between the CNS and the periphery.^{34,35} These properties render EVs a very attractive biomarker; they can carry the contents of specific cell types from different organs in biological fluids that are easily collected and analyzed. Moreover, EV surface markers allow the isolation of EVs with specific cellular origin in the periphery, making it possible to assess e.g. brain derived neuronal or glial contents transported to the periphery via EVs.³⁶ Overall, these properties indicate that the contents of EVs or certain EV subgroups compared to total biofluid contents can provide more relevant information on the studied tissues and diseases.

In this systematic review and meta-analysis, we aimed to identify EV biomarkers for the diagnosis of PD and assess their clinical significance. In addition, we assessed the importance of the biological specimen and the EV isolation method.

Patients and Methods

This systematic review was performed in accordance with the transparent reporting guidelines of the Preferred Reporting Items for Systematic Reviews and Meta-Analyses 2020 (PRISMA 2020).³⁷

Search Strategy

The literature search was performed using PubMed, Web of Science, and Google Scholar databases from their inception through November 15, 2022. We did not impose any restrictions on publication language. The keywords "Parkinson's disease," "exosomes," "biomarker," and "diagnosis" were used in different combinations. The search flow chart is shown in Figure 1.

Selection Process and Eligibility Criteria

All articles retrieved from the electronic databases were imported to Mendeley Reference Manager. Duplicates, reviews, posters, bulletins, news, and conference abstracts were removed manually. The titles and abstracts of the retrieved articles were screened by two authors (M.X. and A.C.), and studies on cellular and animal models or studies irrelevant to PD were excluded (n = 79). After this step, 82 studies were selected for fulltext reading.

The inclusion and exclusion criteria were developed based on the research question. Inclusion criteria considered cohort and case–control studies consisting of patients diagnosed with PD. For PD, we excluded possible co-pathologies with other neurodegenerative



FIG. 1. Flow chart of study selection.

diseases, and for control groups of healthy individuals, we excluded those with signs of neurological or related diseases. Only studies presenting EV isolation from biological specimens and dysregulation of EV composition and contents were included.

Data Collection

Data collected from the selected studies include first author, year of publication, sample matrix, EV isolation method, EV characterization method, biomarker, biomarker analysis method, study design, number of PD patients and HCs, male-to-female ratio, age of HCs, age of PD patients during sample collection, age of PD patients at disease onset, disease duration, Movement Disorder Society-Unified Parkinson's Disease Rating Scale (MDS-UPDRS) score, Hoehn & Yahr stage, Mini-Mental Status Examination (MMSE), Montreal Cognitive Assessment (MoCa), and information on medications. Biomarker outcomes of the studies were collected as means and standard deviations (SD). When data were provided in other forms, the appropriate conversions and calculations were performed according to Hozo et al and Wan et al.^{38,39} Biomarker outcomes of the studies were collected as means and SD. When data were provided only in graphs, we used Web plot digitizer to extract the values.⁴⁰

Risk of Bias/Quality Assessment

Risk of bias assessment was performed separately by two reviewers for each study using the QUADAS-2 tool (Quality Assessment of Diagnostic Accuracy Studies).⁴¹ QUADAS-2 assesses four domains: patient selection, index test, reference standard, and flow and timing. The risk of bias is determined in each domain by answering "yes" or "no" to each of the signaling questions. If the answer is "yes" to all signaling questions, the bias for this domain is considered low. Domains including "no" answers are generally considered as high risk for bias but should be determined individually to assess if the negative answer to the question can render the whole domain problematic.

Data Analysis

The meta-analysis was performed using Review Manager.⁴² We calculated the standardized mean difference (Std.MD) to estimate the average change in values of each biomarker in PD patients compared to HCs. The included studies use different methods to measure the same biomarker, and the absolute values vary significantly. An Std.MD above 1 indicates the biomarker is higher in the PD group and that below 1 indicates the biomarker is higher is higher in the HC group. Random effects meta-analysis was performed to incorporate heterogeneity. Heterogeneity was estimated using the I^2 test. For biomarkers with more than three studies, subgroup analyses were also performed.

Results

Characteristics of Selected Studies

The first literature search of the electronic databases yielded 300 studies; 247 studies remained after duplicate removal. We then screened the studies by abstract and title and excluded 86 reviews, 61 irrelevant studies, and 18 studies on cellular and animal models; 82 studies underwent full-text review. After full-text assessment, 20 studies were excluded due to inappropriate data format or reporting, resulting in 62 studies undergoing quantitative (n = 35) or qualitative analyses (n = 27) (Fig. 1; Tables S1 and S3).⁴³⁻¹⁰⁴

Studies were published between 2014 and 2022, with increased publications in the last 2 years. Participants for the included studies were recruited from Asia (34 of 62 studies), Europe (19 of 62 studies), and North America (10 of 62 studies), indicating that certain populations might be not included or underrepresented.

Twenty-three were case-control studies, and 39 were cohorts. Participant numbers ranged from 7 to 290 in the PD groups and 7 to 215 in the control groups; the male-to-female ratio was reported in all but four studies. All but 13 studies properly reported the age of participants and included the age of PD diagnosis and/or disease duration. The age at sample collection was 64.74 ± 7.24 years for healthy individuals and 67.16 ± 7.90 years for PD patients. Age at disease onset was 61.26 ± 5.97 years, and disease duration was 5.58 ± 3.38 years. This is important in the case of PD diagnosis as the longer the disease, the higher the certainty that PD was correctly diagnosed. Clinical data reported included MDS-UPDRS (46 of 62 studies), MoCa (26 of 62 studies), medication (25 of 62 studies), and MMSE (24 of 62 studies). In addition, only 27 studies explained in detail the exclusion and

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inclusion criteria they used to enroll study participants and properly reported pathologies other than PD.

Biological specimens from which EVs were isolated included plasma (35 of 62 studies), serum (15 of 62 studies), urine (8 of 62 studies), CSF (6 of 62 studies), and saliva (2 of 62 studies); therefore, plasma was the most commonly used biological specimen. The EV isolation methods were EV isolation kits (25 of 62 studies; ExoQuick: 13, ExoEasy: 5, MAC-SPlex Exosome Assays: 2, Total Exosome Isolation Kit: 2. XYCO EV Enrichment Kit: 1. PureExo: 1. and miRCURY Exosome Isolation Serum/Plasma Kit: 1); ultracentrifugation (24 of 62 studies); immunoprecipitation (IP) (8 of 62 studies, L1 cell adhesion molecule [L1CAM]: 6, and total EV: 2); size-exclusion chromatography (SEC) (2 of 62 studies); fluorescence-activated cell sorting (FACS) (1 of 62 studies); and one study performing enzyme-linked immunosorbent assay (ELISA) directly without isolation. In nine studies, a secondary processing method was performed to select a specific subgroup of EVs with IP or FACS. In particular, L1CAM-IP followed ExoQuick isolation (5 of 62 studies), myelin-oligodendrocyte glycoprotein IP followed ExoQuick (1 of 62 studies), and L1CAM-IP followed ultracentrifugation (1 of 62 studies). Collectively, there is no indication for a preferred isolation method, although the majority of studies after 2019 use isolation kits. Interestingly, L1CAM-IP was first performed in 2014 by Shi et al and seemed to gain popularity after 2019.81

The majority of studies (42 of 62) used at least two different methods to characterize the isolated vesicles. The most commonly used method is immunoblotting analysis (44 of 62 studies) employing positive and negative exosome markers. Next is microscopy (37 of 62 studies), specifically electron microscopy, transmission electron microscopy, or atomic force microscopy, to visualize the vesicle shape and size, and then nanoparticle tracking analysis (33 of 62 studies), which measures the size and number of isolated vesicles. Individual studies employed dynamic light scattering, exosome antibody array, augmented colorimetric nanoplasmonic method, acetylcholinesterase activity assay, and flow cytometry. Conclusively, immunoblotting and a second method-either microscopy or a particle analysis method-seem to be the gold standards for EV characterization. As EV size varies and overlaps with the size of other types of vesicles, size and shape estimation are insufficient, and EV markers are required to confirm proper isolation. It is worth highlighting that some of the studies that analyzed RNA biomarkers did not perform any vesicle characterization but proceeded directly to RNA extraction.

The different methods of analysis used to measure the biomarkers were immunoblotting (13 of 62 studies), ELISA (13 of 62 studies), real-time quantitative polymerase chain reaction (RT qPCR) (9 of 62 studies), next-generation sequencing (NGS) (8 of 62 studies), mass spectrometry (6 of 62 studies), electrochemiluminescence (ECL) (6 of 62 studies), flow cytometry (4 of 62 studies), immunomagnetic reduction assay (3 of 62 studies), Meso Scale Discovery (MSD) (2 of 62 studies), single-molecule array (2 of 62 studies), Luminex (2 of 62 studies), and Thioflavin-T (ThT) based assay (2 of 62 studies).

The biomarkers analyzed from exosomal extraction were mainly proteins. A total of 24 studies reported data on α -synuclein, including total aSyn (24 of 24 studies), oligomeric aSyn (6 of 24 studies), and phosphorylated Ser129 aSyn (5 of 24 studies); 5 studies reported data on leucine-rich repeat kinase 2 (LRRK2), including phosphorylated Ser1292 LRRK2 (3 of 5 studies); two studies reported data on tau-protein; two studies reported data on DJ-1; and two studies reported data on clusterin. Other proteins related with neurode-generative disorders, synaptic proteins, surface markers, and inflammatory proteins were also reported and are presented in Table S1. Only biomarkers reported in more than three studies were included in the meta-analysis.

RNA biomarkers from EVs were reported in 15 studies, including miRNAs (12 of 15 studies), long non-coding RNAs (2 of 15 studies), and mRNAs (2 of 15 studies). Four miRNAs were identified in at least three studies: miR-22, miR-24, miR-331, and miR-151. All other RNAs are presented in Table S1. Due to the incomplete vesicle characterization in some of the studies reporting miRNA biomarkers, they were not included in the meta-analysis.

Risk of Bias and Applicability Assessment

Fifty studies were classified as "low risk of bias," 12 studies were classified as "moderate risk of bias," and 2 studies were classified as "high risk of bias" and were not included in the meta-analysis. Data produced by the QUADAS tool on bias and applicability risk are presented in Table S2.

Meta-Analysis

We meta-analyzed data for aSyn, including total aSyn and oligomeric aSyn (Fig. 2), and LRRK2, including total LRRK2 and phosphorylated Ser1292 LRRK2 (Fig. 4).

Meta-analysis of total aSyn in all studies showed a nonsignificant increase in PD (Std.MD [standardized mean difference] = 1.02, 95% confidence interval [CI] = 0.23-1.82, P = 0.01) (Fig. 2A). However, performing the meta-analysis of total aSyn in blood EVs based on their source showed different results: total EVs isolated from blood samples did not show any change between PD patients and HCs (healthy controls,

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Study or Subaroup		PD			нс			Std. Mean Difference	Std. Mean Difference
ottaaj ot oangi oap	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI	IV, Random, 95% Cl
Agliardi 2021	3.5	3.96	32	6.564	6.143	40	4.8%	-0.57 [-1.05, -0.10]	-
Cao 2019	11.39	16.2	74	12.23	39.52	60	4.8%	-0.03 [-0.37, 0.31]	t t
Cerri 2018	35.3	185.66	39	15.82	76.91	33	4.8%	0.13 [-0.33, 0.60]	t
Chung 2021	56	3.7	116	74.5	4.3	66	4.7%	-4.69 [-5.26, -4.12]	
Dutta 2021	107	124	51	96	110	50	4.8%	0.09 [-0.30, 0.48]	t
Han 2019	1.48	0.59	20	1.38	0.92	20	4.7%	0.13 [-0.49, 0.75]	t
Hong 2021	0.38	1.39	170	0.43	1.22	131	4.8%	-0.04 [-0.27, 0.19]	1
Jiang 2020	29.32	20.5	230	12.72	6.1	144	4.8%	1.00 [0.78, 1.22]	
Jiang 2021	22.5	4.8	290	10.8	4.8	144	4.8%	2.43 [2.18, 2.69]	
Kluge 2022	1.51	0.97	30	1	0.24	50	4.8%	0.81 [0.34, 1.28]	-
Niu 2020	2.39	2.15	53	0.64	0.58	21	4.8%	0.93 [0.41, 1.46]	-
Shi 2014	25.07	1.8	264	11.8	0.78	215	4.7%	9.23 [8.62, 9.85]	
Shim 2021	234	183.2	34	219.7	135.3	29	4.8%	0.09 [-0.41, 0.58]	ł
Si 2019	0.2	0.13	49	0.27	0.16	18	4.7%	-0.50 [-1.05, 0.05]	-
Stuendl 2016	13.04	5.31	76	16.01	7.4	58	4.8%	-0.47 [-0.82, -0.12]	-
Stuendl 2021	12.66	0.83	96	8.08	0.54	42	4.6%	6.04 [5.23, 6.85]	
Xia 2019	205.48	64 17	20	151 78	76.37	15	4 7%	0 75 [0 06 1 45]	-
Van 2022	7 556 3	5 551 3	20	3 636 8	3 100 6	10	4.170	0.88 [0.45, 1.30]	-
7hoo 2010	7,550.5	0,001.0	20	5,050.0	1 17	40	4.0%	0.50 [0.43, 1.50]	Ļ
Zhao 2019	1.15	2.14	39	0.0	1.17	40	4.070	0.09 [0.14, 1.04]	_
Znao 2020	139.0	21.74	30	07.09	11.2	30	4.0%	4.10 [3.19, 5.02]	
Zou 2020	7.84	2.46	93	6.07	1.35	85	4.8%	0.88 [0.57, 1.19]	l'
			4050			42.40	400.00/	4 00 50 00 4 000	
l otal (95% CI)			1850			1340	100.0%	1.02 [0.23, 1.82]	
Heterogeneity: Tau ² = 3	3.38; Chi ² =	= 1657.9	6, df = 2	20 (P < 0.	00001); l ²	= 99%			-10 -5 0 5 10
Test for overall effect: 2	Z = 2.52 (P	9 = 0.01)							Higher in HC Higher in PD
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B)				O	igome	eric a	Syn		
_)		PD			нс		S	td. Mean Difference	Std. Mean Difference
Study or Subgroup	Mean	SD	Total	Mean	SD T	otal V	Veight	IV, Random, 95% CI	IV, Random, 95% CI
Agliardi 2021	0.572	0.154	20	0.200	0 1 4 0	40	24 40/	1 21 [0 70 1 72]	
	0.073	0.104	32	0.369	0.140	40	21.170	1.21 [0.70, 1.72]	
Cao 2017	7.03	6.31	74	0.92	0.61	60	21.2%	1.29 [0.91, 1.66]	
	uu	0.7	30	1	0.74	50	17.5%	12.15 [10.17, 14.13]	
Kluge 2022	0.0				75 0 4	5	10 60/	0 70 10 60 9 001	
Xiuge 2022 Xia 2019	206.205	64.7	5	151.78	/ 5.84	5	19.5%	0.70 [-0.60, 2.00]	_
Kiuge 2022 Xia 2019 Zheng 2021	206.205 1.38	64.7 0.13	5 36	151.78 1	0.01	36	20.6%	4.08 [3.25, 4.90]	· · · ·
Kluge 2022 Xia 2019 Zheng 2021	206.205 1.38	64.7 0.13	5 36	151.78	0.01	36	20.6%	4.08 [3.25, 4.90]	
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI)	206.205 1.38	64.7 0.13	5 36 177	151.78	0.01	36 191 1	20.6%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58]	•
Xiuge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² =	206.205 1.38 4.61; Chi ²	64.7 0.13 = 148.83	5 36 177 3, df = 4	151.78 1 4 (P < 0.0	0.01 00001); l ²	36 191 1 = 97%	20.6% 00.0%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58]	•
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Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Aqliardi 2021	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5	64.7 0.13 = 148.83 P = 0.000 PD <u>SD</u> 3.96	5 36 177 3, df = 4 03) <u>Total</u> 32	4 (P < 0.0 aSyn <u>Mean</u> 6,564 (0.01 00001); I ² in tota IC <u>SD To</u> 5.143	36 191 1 = 97% al exe tal W 40 1	20.6% 20.6% 00.0% DSOME Sto eight 4.4%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 	-20 -10 0 10 : Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1	64.7 0.13 = 148.83 P = 0.000 PD <u>SD</u> 3.96 185.66	5 36 177 3, df = 4 03) <u>Total</u> 32 39	4 (P < 0.0 aSyn 6.564 (15.82	0.01 00001); I ² in tota IC <u>SD To</u> 5.143 76.91	36 191 1 = 97% al exe tal W 40 1 33 1	20.6% 20.6% 00.0% DSOME Sto eight 4.4% 4.4%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% Cl -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60]	-20 -10 0 10 : Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 3.5.3 1 56	64.7 0.13 = 148.83 P = 0.000 PD <u>SD</u> 3.96 185.66 3.7	5 36 177 3, df = - 03) <u>Total</u> 32 39 116	151.78 1 4 (P < 0.0 aSyn 6.564 15.82 74.5	75.84 0.01 00001); I ² in tota HC <u>SD To</u> 5.143 6.91 4.3	36 191 1 = 97% al exe tal W 40 1 33 1 66 1	20.6% 20.0% 00.0% DSOME Sta 4.4% 4.4% 4.3%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-52, 6, -4.12]	-20 -10 0 10 2 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Hen 2010	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1 56	64.7 0.13 = 148.83 P = 0.000 <u>SD -</u> 3.96 185.66 3.7 0.59	5 36 177 3, df = 4 03) Total 32 39 116 20	aSyn 6.564 (15.22 74.5 138	75.64 0.01 00001); I ² in tota HC <u>SD To</u> 5.143 76.91 4.3 0.92	36 191 1 = 97% 1 exc $1 exc1 exc$	20.6% 20.0% 00.0% DSOME street eight 4.4% 4.3% 4.3%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 [3.25, 4.90] -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75]	-20 -10 0 10 2 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1 56 1.48 234	64.7 0.13 = 148.83 P = 0.000 <u>PD</u> <u>3.96</u> 185.66 3.7 0.59	5 36 177 3, df = 4 03) Total 32 39 116 20 34	aSyn 6.564 (15.82 7 74.5 230 7	75.84 0.01 00001); I ² in tota HC <u>SD To</u> 5.143 76.91 4.3 0.92 4.3	36 191 1 = 97% al exe tal W 40 1 33 1 66 1 20 1	20.6% 20.0% 00.0% DSOME Street 4.4% 4.4% 4.3% 4.3% 4.4%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference 11, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.04 1, 0.75]	-20 -10 0 10 1 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1 56 1.48 234 234	64.7 0.13 = 148.83 P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.59	5 36 177 3, df = 4 03) Total 32 39 116 20 34 20	aSyn 4 (P < 0.0 aSyn <u>Hean</u> 6.564 (1 15.82 7 1.38 219.7 200	75.64 0.01 00001); I ² in tota IC <u>SD To</u> 5.143 76.91 4.3 0.92 135.3	36 191 1 = 97% al exa tal W 40 1 33 1 66 1 20 1 29 1	20.6% 20.0% 00.0% DSOME Eight 4.4% 4.3% 4.3% 4.3%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [-2.2, 6.51]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Yie 2020	4.61; Chi ² Z = 3.66 (i <u>Mean</u> 3.5 35.3 1 56 1.48 234 12.66 205 10	64.7 0.13 = 148.83 P = 0.000 PD <u>SD</u> <u>3.96</u> 185.66 3.7 0.59 183.2 0.83 2.083 2.083 2.085 2.0	5 36 177 3, df = 4 03) Total 32 39 116 20 34 96 20	aSyn 4 (P < 0.0 aSyn <u>Hean</u> 6.564 (15.82) 74.5 1.38 219.7 8.08	75.64 0.01 00001); I ² in tota IC SD To 3.143 0.92 135.3 0.54 0.52	36 191 1 = 97% al exe tal W 40 1 33 1 66 1 20 1 29 1 42 1 42 1	20.6% 20.0% 00.0% CSCOME 20.6%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.0, -17]	-20 -10 0 10 1 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48	64.7 0.13 = 148.83 P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17	5 36 177 3, df = - 03) Total 32 39 116 20 34 96 20	aSyn 4 (P < 0.0 aSyn 6.564 (15.82 74.5 1.38 219.7 8.08 15.78	75.84 0.01 00001); I ² in tota IC SD To 3.143 76.91 4.3 0.92 135.3 0.54 76.37	36 191 1 = 97% al exe tal W 40 1 33 1 66 1 20 1 29 1 42 1 15 1	20.6% 20.0% 00.0% CSOME stre eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.1% 4.2%		-20 -10 0 10 1 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stiuendl 2021 Xia 2019	206.205 1.38 4.61; Chi ² Z = 3.66 (I Mean 3.5 35.3 1.48 234 12.66 205.48	64.7 0.13 = 148.83 P = 0.000 3.96 185.66 3.7 0.59 183.2 0.83 64.17	5 36 177 3, df = - 03) Total 32 39 116 20 34 96 20	aSyn 6.564 (1 1.5.82 1 1.5.82 1 74.5 1.38 219.7 8.08 151.78 1	75.84 0.01 00001); I ² in tota HC <u>SD To</u> 5.143 76.91 4.3 0.92 135.3 0.54 76.37	36 191 1 = 97% 1 exc $tal w40 133 166 120 129 142 115 1$	20.6% 20.0% 00.0% DSOME Sta eight 4.4% 4.3% 4.3% 4.3% 4.4% 4.1% 4.2%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference <u>IV, Random, 95% C1</u> -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.25 [4.65, 2.15]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI)	206.205 1.38 4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1 56 1.48 234 12.66 205.48	64.7 0.13 = 148.8: P = 0.000 <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17	5 36 177 3, df = - 03) Total 32 39 116 20 34 96 20 357	aSyn 4 (P < 0.0 aSyn 6.564 (1) 15.82 (1) 74.5 1.38 219.7 8.08 151.78 (1) 151.78	75.84 0.01 00001); I ² in tota 4C <u>SD To</u> 5.143 76.91 4.3 0.92 135.3 0.54 76.37 2	$\begin{array}{c} 36 \\ 191 & 1 \\ = 97\% \\ \hline al exa \\ \hline bala \\ 40 & 1 \\ 33 & 1 \\ 33 & 1 \\ 20 & 1 \\ 20 & 1 \\ 29 & 1 \\ 42 & 1 \\ 15 & 1 \\ 45 & 10 \\ \hline \end{array}$	20.6% 20.0% 00.0% CSCOME eight 4.4% 4.4% 4.3% 4.3% 4.3% 4.1% 4.2% 0.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1	206.205 1.38 4.61; Chi ² = 3.66 (I <u>Mean</u> 3.5 3.5 3.5 1.48 234 12.66 205.48 6.54; Chi ² =	64.7 0.13 = 148.8: P = 0.000 PD SD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25	5 36 177 3, df = - 03) 32 39 116 20 34 96 20 357 , df = 6	A (P < 0.0 A (P <	75.84 0.01 00001); I ² in tota 4C <u>SD To</u> 3.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² =	36 191 1 = 97% tal W 40 1 33 1 66 1 20 1 29 1 42 1 15 1 45 10 99%	20.6% 20.0% 00.0% CSOME Sta eight 4.4% 4.4% 4.3% 4.3% 4.4% 4.3% 4.4% 4.3% 4.2% 0.0%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16]	-20 -10 0 10 2 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Gerri 2018 Chung 2021 Stuendi 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = i Test for overall effect: <i>X</i>	206.205 1.38 206.205 2 = 3.66 (1 <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79)	5 36 177 3, df = 4 33 39 116 20 34 20 34 20 357 , df = 6	A (P < 0.0 A (P <	75.84 0.01 00001); I ² in tota ic to	36 36 191 1 = 97% 10 20 11 20 120 1 20 1 20 1 20 1 20 1 20 1 15 1 15 1 15 10 99%	20.6% 20.0% 00.0% 2550 2550 2550 2550 2550 2550 2550 25	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4.08 an Difference 1.057 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16]	-20 -10 0 10 2 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² =	206.205 1.38 2.4.61; Chi ² Z = 3.66 (l <u>Mean</u> 3.5 3.5.3 1 56 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.8; P = 0.000 SD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79)	5 36 177 33, df = - 03) 32 39 116 20 34 96 20 357 , df = 6	A (P < 0.0 A (P <	75.84 0.01 00001); I ² in tota IC SD To 3.143 0.92 135.3 0.54 76.37 2 0001); I ² =	36 36 191 1 = 97% 10 20 11 20 1 5 1 5 1 5 1 5 1 5 1 5 1 5 1 5	20.6% 20.0% 00.0% DSOME eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.1% 4.2% 00.0%	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% C1 -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16]	-20 -10 0 10 : Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Corri 2018 Chung 2021 Han 2019 Shim 2021 Stuendi 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = (Test for overall effect: 2	206.205 1.38 2.4.61; Chi ² = 2 = 3.66 (I Mean 3.5 1 56 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.8: P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 = 0.79) aSyn i	5 36 177 3, df = - 03) 116 20 34 96 20 357 , df = 6	aSyn 4 (P < 0.0 aSyn 6.564 (15.82) 74.5 1.38 219.7 8.08 15.78 (P < 0.0 (P < 0.0)	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 3.143 0.92 135.3 0.54 76.91 4.3 0.54 76.37 2 0001); I ² =	36 36 191 1 = 97% 1 exe tal w 40 1 33 1 66 1 20 1 29 1 42 1 15 1 45 10 99% ►	20.6% 20.6% 00.0% DSOME Sta eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOME DSOME State	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] e exosomes	-20 -10 0 10 : Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 C)	206.205 1.38 2.4.61; Chi ² Z = 3.66 (I Mean 3.5 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.83 P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i	5 36 177 3, df = - 33 39 116 20 34 96 20 357 , df = 6	4 (P < 0.0 aSyn Mean 6.564 (15.82) 74.5 1.38 219.7 8.08 151.78 (P < 0.00 eurona	75.84 0.01 00001); I ² in tota 4C <u>SD To</u> 5.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² il LLCA	36 191 1 191 1 191 1 191 1 197% tal W 40 1 33 1 66 1 20 1 29 1 42 1 15 1 45 10 99%	20.6% 20.6% 00.0% DSOME State eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOME DSOME State Sta	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] e exosomes Std. Mean Difference	-20 -10 0 10 2 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendi 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D)	208.205 208.205 1.38 24.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.8; P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i PD	5 36 177 3, df = - 03) 116 20 34 96 20 357 , df = 6 7 7 7 7 7 7 7 7 7 7 7 7 7	$\frac{1}{4} (P < 0.0)$ $\frac{aSyn}{6.564}$ $\frac{1}{6.564}$ $\frac{1}{5.82}$ $\frac{74.5}{1.38}$ $\frac{219.7}{8.08}$ $\frac{1}{151.78}$ $\frac{1}{51.78}$ $\frac{1}{51.78}$ $\frac{1}{51.78}$ $\frac{1}{51.78}$ $\frac{1}{51.78}$ $\frac{1}{51.78}$	75.84 0.01 00001); I ² in tota 4C <u>SD To</u> 3.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² = hI L1CA HC	36 191 1 191 1 191 1 191 1 191 1 191 1 197% 101 20 1 20 1 20	20.6% 20.6% 00.0% Stor eight 4.4% 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% 0.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] CE EXOSOMES Std. Mean Difference IV, Random 95% CI	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendi 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D)	206.205 206.205 1.38 2.4.61; Chi ² Z = 3.66 (l <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i PD <u>SD</u> 2 = 0.79)	5 36 177 3, df = - 32 39 116 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 - - - - - - - - - - - - -	4 (P < 0.0 aSyn Mean 6.564 (15.82) 74.5 1.38 219.7 8.08 151.78 (P < 0.0 (P < 0.0) curona al Mea	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 5.143 76.91 4.3 0.92 135.3 0.94 76.071 12 12 12 12 12 12 12 12 12 1	36 191 1 = 97% 1 exe tal W 40 1 33 1 66 1 20 1 20 1 20 1 20 1 20 1 45 10 · 99% • 99%	20.6% 20.6% 00.0% DSOME eight 4.4% 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOITIV 1 Weight	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% C1 -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% C1	-20 -10 0 10 Higher in HC Higher in PC Std. Mean Difference IV, Random, 95% CI
Kuge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Corri 2018 Chung 2021 Han 2019 Shim 2021 Stuend! 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = { Test for overall effect: 2 D) Duta 2021 Lines 2020	206.205 1.38 2.4.61; Chi ² Z = 3.66 (I Mean 3.5 1 56 1.48 234 12.66 205.48 6.54; Chi ² : Z = 0.26 (P 2 Mean 107 2022	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> <u>3.96</u> 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i PD <u>SD</u> <u>12</u> 2 2 2 2 2 2 2 2 2 2 2 2 2	5 36 177 3, df = . 03) Total 32 39 116 20 34 96 20 357 , df = 6 0 Total 5 20 30 32 39 116 20 34 96 20 35 117 5 20 30 116 20 34 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 36 116 20 35 116 20 35 116 20 35 20 35 116 20 35 116 20 35 116 20 35 116 20 35 116 20 35 106 20 35 106 20 20 35 106 20 35 106 20 35 106 20 35 106 20 20 35 106 20 20 106 20 106 20 106 20 106 20 106 20 106 106 20 106 106 20 106 106 106 106 106 106 106 10	151.78 1 4 (P < 0.0	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 3.143 0.92 135.3 0.54 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² = HC <u>N</u> HC <u>N</u> SI 16 111 17 17 17 17 17 17 17 17 1	36 36 191 1 = 97% 1 kl exc kl exc <	20.6% 20.6% 00.0% DSOME Sta eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOIL 0.0% 0.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C27 [-1.05, 0.0, 0.45] C27 [-1.05, 0.0, 0.45] C28 [-1.05, 0.0, 0.45] C29 [-0.30, 0.48] C20 [-0.50, 0.40] C20 [-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020	206.205 1.38 2.4.61; Chi ² = 2 = 3.66 (l Mean 3.5 1.48 234 12.66 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P 4 Mean 107 29.32	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD b b b c c c c c c c c	5 36 177 3, df = . 03) Total 32 39 116 20 34 96 20 357 , df = 6 0 Total 4 5 5 23 Dota 5 7 7 7 7 7 7 7 7 7 7 7 7 7	$\begin{array}{c} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 $	75.84 0.01 00001); I ² in tota 4C <u>SD</u> To 5.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² = hC n SI 6 111 2 6 111 2 6	36 191 1 = 97% al exe tal w 40 1 33 1 20 1 20 1 20 1 20 1 20 1 20 1 45 10 99% NM-p 0 5 1 1 44 0 5 0 5 0 5 0 5 0 5 0 5 0 5 0 5	20.6% 20.6% 00.0% DSOME Ste eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOME Ste eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.2% 0.0% DSOME Ste 2.% 1.% 1.% 1.% 1.% 1.% 1.% 1.% 1	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% CI 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 0.09 [-0.30, 0.48] 6 0.09 [-0.30, 0.48] 7 0.00 [-0.78, 1.22] 7 0.00 [-0.78, 1.22]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Stuendl 2021 Xia 2019 Total (95% Cl) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2021	206.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (l <u>Mean</u> 3.5 3.5.3 1.48 2.34 1.2.66 205.48 6.54; Chi ² = 2.6. (P <u>Mean</u> 1.2.66 2.2.5.48 6.54; Chi ² = 2.2.6 (P	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD <u>SD</u> <u>183.2</u> 0.83 64.17 a Syn i PD <u>183.2</u> 0.83 64.17 b 2 0.79) b 2 0.79) c b 2 0.79) c c c c c c c c	5 36 177 3, df = - 32 39 116 20 34 96 20 357 , df = 6 0 0 0 155 20 34 96 20 34 96 20 34 96 20 357 0 157 16 - 20 34 20 357 16 - 16 - 16 - 17 17 16 - 16 - 16 - 16 - 16 - 17 - 16 - 16 - 16 - 16 - 17 - 10 - - 10 - - - - - - - - - - - - -	151.78 1 4 (P < 0.0	75.84 0.01 00001); ² in tota 1C SD To 3.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0.92 135.3 0.54 76.37 2 0001); ² = ILLCA HC N SI 6 111 2 6 8 4.3 0.54 111 12 12 12 13 13 14 12 12 12 12 12 12 12 12 12 12	36 36 1191 1 = 97% 1 bl exc tal W 40 1 20 1 20 1 20 1 20 1 99% 0 NM-p 0 5 0 5 0 1 8 14	20.6% 20.6% 00.0% 0.	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C EXOSOMES Std. Mean Difference t IV, Random, 95% CI 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 2.43 [2.18, 2.69]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuend! 2021 Xia 2019 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: 2 D) Study or Subgroup Duta 2021 Jiang 2020 Jiang 2021 Kluge 2022	206.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (l <u>Mean</u> 3.5 3.5.3 1 5.6 1.48 234 12.66 205.48 6.54; Chi ² = 0.26 (P <u>Mean</u> 107 29.32 22.5; 1.51	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD <u>SD</u> 7 12 2 20.5 4.1 0.9	5 36 177 3, df = - 03) Total 32 39 116 20 34 96 20 34 96 20 357 , df = 6 0 Total 20 34 20 35 20 35 20 35 20 37 20 37 20 39 20 30 20 357 57 57 57 57 57 57 57 57 57	al Mean (P < 0.0 aSyn Mean 6.564 (15.82 7 74.5 1.38 219.7 8.08 151.78 7 (P < 0.00 COMPANIES (P < 0.00 (P <	75.84 0.01 00001); I ² in tota HC SD To 3.143 0.92 135.3 0.54 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² = HC n SI 6 111 2 6. 14 2 0. 14 2 16 2 16 16 11 12 12 12 12 12 12 12 12 12	36 36 191 1 = 97% 1 bl exc tal W tal exc tal tal tal exc tal tal tal exc	20.6% 20.6% 00.0% DSOME Sta eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOIL 0.0% 0	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% Cl -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C27 [-0.30, 0.48] 6 0.09 [-0.30, 0.48] 6 0.68 [0.34, 1.28] 6 0.81 [0.34, 1.28] 6 0.54 [0.34, 1.28] 7 0.54 [0.54, 1.55] 7 0.55 [0.55,	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = (Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2020 Niu 2020	206.205 1.38 206.205 2 = 3.66 (I Mean 3.5 1.48 234 12.66 205.48 6.54; Chi ² = Z = 0.26 (P Mean 107 29.32 22.5 1.55 2.35	64.7 0.13 = 148.83 P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i PD PD 122 2 0.5 4.1 0.59 2 0.79) 2 0.5 4.1 0.5 2 0.2 5 4.1 0.5 2 0.5 4.1 0.5 5 4.1 0.5 5 2 0.5 5 4.1 0.5 5 2 0.5 5 4.1 0.5 5 2 0.5 5 4.1 0.5 5 2 0.5 5 0	5 36 177 3, df = . 32 39 116 20 34 96 20 357 df = 6 0 57 df = 6 0 20 357 df = 6 0 34 34 96 20 357 df = 6 0 20 357 df = 6 0 7 35 55	151.78 1 4 (P < 0.0	75.84 0.01 00001); I ² in tota HC SD To 5.143 76.91 0.92 135.3 0.54 76.37 2 0001); I ² HC n SI HC n 102 4.4 0.54	36 191 191 191 191 11 11 11 20 11 20 11 120 115 115 105 105 105 105 105 105 105 11 14 14 45 105 11 14 45 11 12 13 14 45 58 21	20.6% 20.6% 00.0% DSOME State eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 0.0% DSOIL 0.0% DSOIL 0.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% CI 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 0.43 [10.34, 1.28] 6 0.93 [0.41, 1.46]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = i Total (95% CI) Heterogeneity: Tau ² = z Total (95% CI)	208.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = 0.26 (P <u>Mean</u> 1.51 2.33 2.2.5 1.51 2.33 2.5.07	64.7 0.13 = 148.8; P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD a Syn i 7 12 2 20. 5 4. 1 0.9 9 2.11 7 1.	5 36 177 3, df = - 32 39 116 20 34 96 20 357 , df = 6 0 7 34 96 20 357 , df = - 20 34 96 20 357 34 96 20 357 34 55 5 5 8 20 35 8 20 34 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 8 20 35 7 116 1 1 1 1 1 1 1 1 1 1 1 1 1	151.78 1 4 (P < 0.0	75.84 0.01 00001); I ² in tota in	36 36 191 1 = 97% 1 exe tal w 40 1 33 1 20 1 45 10 99% ► ► ► ► ► ► ► ► ► ► ► ► ►	20.6% 20.6% 00.0% Store eight 4.4% 4.4% 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.2% 0.0% 0	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C EXOSOMES Std. Mean Difference t IV, Random, 95% CI 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 2.43 [2.18, 2.69] 6 0.81 [0.34, 1.28] 6 0.93 [0.41, 1.46] 6 9.23 [8.62, 9.85]	-20 -10 0 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2021 Kluge 2022 Niu 2020 Shi 2014 Si 2019	206.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 3.5.3 1 5.6 1.48 234 235 25.07 2.38 25.07 0.2	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD <u>PD</u> <u>1 S</u> 7 12 2 200 5 4.1 1 0.9 9 2.11 7 1.2 2 0.01	5 36 177 3, df = - 03) Total 32 39 116 20 34 96 20 357 , df = 6 0 Total 20 34 96 20 35 20 20 35 20 35 20 20 20 35 20 35 20 35 20 35 20 20 20 20 20 20 20 20 20 20	al Mean (P < 0.0 aSyn F Mean 6.564 (15.82 7 74.5 1.38 219.7 8.08 151.78 7 (P < 0.00 (P < 0.00 al Mea 1 9 0 12.7 0 10. 0 10. 9 0.2	75.84 0.01 00001); I ² in tota HC SD To 3.143 0.92 135.3 0.54 76.91 4.3 0.52 76.91 4.3 0.54 76.37 2 0001); I ² = HC n SI 6 111 2 6. 14 2 4 0.54 7 7 0.01 12 12 12 12 12 12 12 12 12 1	36 191 191 1 97% 1 2 1 1 4 5 2 1 1 4 5 2 1 1 4 5 2 1 1 4 5 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	20.6% 20.6% 00.0% DSOME Sta eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.2% 0.0% DSOITIV NOSIT	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% Cl -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C EXOSOMES Std. Mean Difference IV, Random, 95% Cl 6 0.09 [-0.30, 0.48] 6 0.09 [-0.30, 0.48] 6 0.93 [0.41, 1.42] 6 0.93 [0.41, 1.42] 6 0.93 [0.41, 1.48] 6 0.93 [0.41, 1.48] 6 0.93 [8.62, 9.85] 6 -0.50 [-1.05, 0.05]	-20 -10 0 10 Higher in HC Higher in PE Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2021 Kluge 2022 Niu 2020 Shi 2014 Si 2019 Yan 2022	206.205 = 205.	64.7 0.13 = 148.83 P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD T 12 2 20.5 5 4.1 1 0.9 9 2.1 7 1.2 2 0.1 3 5,551.	5 36 177 3, df = . 03) Total 32 39 116 20 34 96 20 357 , df = 6 0 Total 4 5 5 23 39 10 32 39 7 32 39 7 32 39 7 32 39 7 30 32 39 7 32 39 7 30 32 39 34 4 50 35 7 36 36 37 37 37 37 37 37 37 37 37 37	151.78 1 4 (P < 0.0	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 5.143 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 HC n SI 6 111 2 6.8 4.3 0.54 HC n SI 6 111 2 6.8 4.3 0.77 1 0.2 1 0	36 191 191 191 191 191 191 191 191 11 11 120 115 115 115 115 115 115 115 115 115 115 115 11 445 10 5 11 14 5 11 14 15 11 14 15 10 5 14 5 14 5 14 15	20.6% 20.6% 00.0% DSOME Store eight 4.4% 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.2% 0.0% DSOIT 1.9%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] C26 [-1.65, 2.16] C26 [-1.65, 2.16] C24 [2.18, 2.69] 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 0.93 [0.41, 1.28] 6 0.50 [-1.05, 0.05] 6 0.88 [0.45, 1.30]	-20 -10 0 10 Higher in HC Higher in PE Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2020 Jiang 2020 Shi 2014 Si 2019 Yan 2022 Zhao 2019	208.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (I <u>Mean</u> 3.5 35.3 1.48 2.34 12.66 205.48 6.54; Chi ² = 2.64 2.25 2.25 1.51 2.33 2.25.07 0.2 7.556.3 7.75	64.7 0.13 = 148.8; P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) aSyn i PD aSyn i PD aSyn i 2 20.5 4.1 7 12 2 20.5 4.1 7 12 2 20.5 5 4.1 7 12 2 20.1 3 5,551.1 5 2.7	5 36 177 3, df = . 32 39 116 20 34 96 20 34 96 20 357 , df = 6 D Tott 4 5 5 238 20 357 5 5 238 20 34 34 34 34 34 357 34 34 357 5 5 5 338 238 34 34 34 34 357 5 5 338 34 34 34 34 357 34 34 357 357 358 34 357 3586 3586 3586 3586 3586 3586 3586 3586 35	$\begin{array}{c} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 $	75.84 0.01 00001); I² in tota IC SD To 5.143 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I² = ht L1CA n 5 1.1 0.2 4.3 0.92 135.3 0.54 76.37 2 0.61 1.1 0.2 4.3 3.3(100,0) 5 1.1*	36 191 1 = 97% 1 exe tal w 40 1 33 1 20 1 20 1 20 1 20 1 20 1 20 1 20 1 20 1 42 1 15 1 45 10 99% NM-p 5 1 1 445 10 0 5 5 1 1 445 20 8 21 8 21 8 21 6 4 4 7 4 4 5 7 4 7 4 7 4 7 4 7 4 7 4 7 4 7 4	20.6% 20.6% 00.0% DSOME Ste eight 4.4% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.3% 4.2% 0.0% DSOIL 10 10 10 10 10 10 10 10 10 10	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% CI 6 0.09 [-0.30, 0.48] 6 1.00 [0.78, 1.22] 6 2.43 [2.18, 2.69] 6 0.81 [0.34, 1.28] 6 0.93 [0.41, 1.46] 6 9.23 [8.62, 9.85] 6 0.08 [0.45, 1.30] 6 0.88 [0.45, 1.30] 6 0.59 [0.14, 1.04]	-20 -10 0 10 Higher in HC Higher in PE Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% Cl) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuendl 2021 Xia 2019 Total (95% Cl) Heterogeneity: Tau ² = 1 Total (95% Cl) Heterogeneity: Tau ² = 2 Total (95% Cl) Heterogeneity: Tau ² = 1 Total (95% Cl) Heterogeneity: Tau ² = 1 Total (95% Cl) Heterogeneity: Tau ² = 1 Total (95% Cl) Heterogeneity: Tau ² = 1 Xia 2019 Dutta 2021 Jiang 2020 Jiang 2020 Shi 2014 Si 2019 Yan 2022 Zhao 2019 Zhao 2020	206.205 208.205 1.38 2.4.61; Chi ² = 3.66 (l <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = 2.65 2.026 (P <u>Mean</u> 107 2.9.33 2.2.5 1.51 2.35 2.50 0.2 7.556.3 7.75 139.6 (l	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 2 = 0.79) a Syn i PD <u>5 2 4.</u> 1 0.9 9 2.11 7 1.2 2 200. 5 4. 1 0.9 9 2.17 7 1.2 2 20. 5 4. 1 0.9 9 2.17 7 1.2 2 20. 5 4. 1 0.9 9 2.17 7 1.2 2 2.0. 5 4. 1 0.9 9 2.17 7 1.2 2 2.0. 5 4. 1 7 1.2 2 2.0. 5 5 1. 5 5 2.7 5 2.7	5 36 177 3, df = - 32 39 116 20 34 96 20 357 , df = 6 0 0 0 0 0 16 - 0 34 96 20 357 7 8 20 34 96 20 34 96 20 357 7 38 20 34 34 34 34 34 34 34 34 34 34	al Mean (P < 0.0 aSyn Mean 6.564 (1) 15.82 (2) 74.5 1.38 219.7 8.08 (P < 0.0 (P < 0.0 (P < 0.0 0 0 10.0 0 10.0 0 10.0 0 10.0 0 10.0 0 10.	75.84 0.01 00001); I ² in tota ic SD To SD To SD To S.143 0.92 135.3 0.54 76.91 2.54 76.37 2 0001); I ² = HC N SI 6 111 2.6.8 4.1 0.54 7 1.1 8 0.54 7 1.1 8 0.51 1.1 9 1.1 1.1 1.1 1.1 1.1 1.1	36 36 191 1 = 97%	20.6% 20.6% 20.0% 20	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% Cl -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% Cl 6 0.09 [-0.30, 0.48] 6 0.09 [-0.30, 0.48] 6 0.03 [0.34, 1.22] 6 0.81 [0.34, 1.28] 6 0.93 [0.41, 1.46] 9 .923 [8.62, 9.85] 6 0.59 [0.14, 1.04] 6 0.59 [0.14, 1.04] 6 4.10 [3.19, 5.02]	-20 -10 0 10 Higher in HC Higher in PE Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Corri 2018 Chung 2021 Han 2019 Shim 2021 Stuend! 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = (Test for overall effect: 2 D) Study or Subgroup Dutta 2021 Jiang 2020 Jiang 2020 Jiang 2021 Kluge 2022 Niu 2020 Shi 2014 Si 2019 Yan 2022 Zhao 2019 Zhao 2020	206.205 208.205 1.38 4.61; Chi ² Z = 3.66 (l Mean 3.5 1.48 234 12.66 205.48 6.54; Chi ² = 0.26 (P Mean 107 2.9.32 2.25.6 1.51 2.35 2.25.77 1.59.6 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.25.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.35 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.57.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.77.77 1.51 2.55 2.57.77 1.51 2.55 2.57.77 1.51 2.55 2.57.77 1.51 2.55 2.57.77 1.51 2.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.55 2.57.77 1.57.75	64.7 0.13 = 148.8; P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD 5 4.1 7 12 2 20.5 5 4.1 0.9 3 5,551.5 5 2.7 6 21.7 4 2.4	5 36 177 3, df = . 32 39 116 20 34 96 20 34 96 20 357 df = 6 D Total 4 5 5 233 34 4 5 5 233 34 34 34 34 34 34 357 357 35 35 35 35 35 35 35 35 35 35 35 35 35 35 35 35 35 34 35 35 34 35 35 34 35 35 34 35 35 34 35 35 34 35 35 34 35 34 35 34 35 34 35 34 35 34 35 34 34 34 35 34 34 34 34 35 34 35 35 55 55 253 34 34 34 34 34 34 35 34 34 35 34 34 35 34 34 35 35 34 34 35 35 34 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 34 35 35 34 35 35 34 35 35 34 35 35 35 34 35	aSyn 4 (P < 0.0	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 3.143 76.91 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 4.3 0.92 135.3 0.54 HC n SI 6 111 2 6. 1 0.2 4 4 0.57 1 0.2 4 1 0.2 4 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 1 0.2 1 0.5 1 1 0.5 1 0 0 0 0 0 1 0 0 1 0 0 0 0 1 1 0 0 0 1 1 0 0 0 1 1 0 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 1 1 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 1 0 1 1 0 1 1 1 0 1 1 1 0 1 1 1 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1	36 36 191 = 97% 1 1 1 1 20 1 20 1 20 1 45 10 5 1 44 5 8 21 1 44 5	20.6% 20.6% 20.6% 00.0% Store eight 4.4% 4.4% 4.3% 4.4% 4.3% 4.4% 4.3% 4.4% 4.3% 4.4% 4.3% 4.4% 4.3% 4.2% 0.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 an Difference IV, Random, 95% CI -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16]	-20 -10 0 10 Higher in HC Higher in PE Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
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Kuge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Stuendl 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = 1 Total (95% CI) Heterogeneity: Tau ² = 2 Total (95% CI) Dutta 2021 Jiang 2020 Jiang 2021 Kluge 2022 Niu 2020 Shi 2014 Si 2019 Yan 2022 Zhao 2019 Zhao 2020 Zou 2020 Total (95% CI)	208.205 208.205 1.38 2.4.61; Chi ² Z = 3.66 (l <u>Mean</u> 3.5 35.3 1.48 234 12.66 205.48 6.54; Chi ² = 2.0.26 (P <u>Mean</u> 1.51 2.33 2.2.6 1.51 2.33 2.5.07 0.2 7.556.3 7.7.78 139.6 7.84	64.7 0.13 = 148.8; P = 0.000 PD 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD a Syn i PD a Syn i 7 12 2 20. 5 4. 1 0.9 9 2.11 7 1. 2 4. 1 3 5,551. 5 2.7 4 2.4	5 36 177 3, df = - 32 39 116 20 34 96 20 357 , df = 6 0 7 34 96 20 357 , df = - 6 20 34 96 20 357 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 96 20 34 34 96 20 34 34 96 20 357 34 45 55 55 55 55 55 55 55 8 20 34 45 55 55 55 55 55 8 20 34 4 33 4 5 55 55 55 55 55 55 55 55	$\begin{array}{c} 151.78 \\ 1 \\ 1 \\ 4 (P < 0.0 \\ \hline aSyn \\ \hline Mean \\ \hline 6.564 \\ (15.82 \\ 74.5 \\ 1.38 \\ 219.7 \\ 8.08 \\ 151.78 \\ 1.38 \\ 219.7 \\ 6.08 \\ 1.38 \\ 219.7 \\ 74.5 \\ 1.38 \\ 219.7 \\ 74.5 \\ 1.38 \\ 219.7 \\ 1.38 \\ 200 \\ 1.38 \\ 1.38 \\ 200 \\ 1.38 \\ $	75.84 0.01 0.0001); I ² in tota in tota in in tota in<	36 36 36 191 11 = 97% 1 1 Image: state of the sta	19.5% 20.6% 00.0%	4.08 [3.25, 4.90] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4. Mean Difference IV, Random, 95% Cl -0.57 [-1.05, -0.10] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% Cl 6.0.09 [-0.30, 0.48] 6.0.09 [-0.30, 0.48] 6.0.09 [-0.30, 0.48] 6.0.30 [-4.5, 2.16] E EXOSOMES Std. Mean Difference IV, Random, 95% Cl 6.0.09 [-0.30, 0.48] 6.0.93 [0.41, 1.42] 6.0.88 [0.45, 1.30] 6.0.88 [0.45, 1.30] 6.0.88 [0.57, 1.19] 6.1.84 [0.76, 2.93]	-20 -10 0 10 Higher in HC Higher in PC Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI
Kluge 2022 Xia 2019 Zheng 2021 Total (95% CI) Heterogeneity: Tau ² = Test for overall effect: C) Study or Subgroup Agliardi 2021 Cerri 2018 Chung 2021 Han 2019 Shim 2021 Stuend! 2021 Xia 2019 Total (95% CI) Heterogeneity: Tau ² = C) Study or Subgroup Duta 2021 Jiang 2020 Jiang 2021 Kluge 2022 Niu 2020 Shi 2014 Si 2019 Yan 2022 Zhao 2019 Zhao 2020 Total (95% CI) Heterogeneity: Tau ²	206.205 206.205 1.38 2.4.61; Chi ² = 3.66 (l <u>Mean</u> 3.5 3.5.3 1 56 1.48 234 12.66 205.48 6.54; Chi ² = 0.26 (P <u>Mean</u> 107 29.32 22.5 1.57 2.38 25.07 0.2 7.556.3 7.75 139.6 (Z 139.6 7.84 = 3.31; Ch	64.7 0.13 = 148.8; P = 0.000 PD <u>SD</u> 3.96 185.66 3.7 0.59 183.2 0.83 64.17 = 478.25 P = 0.79) a Syn i PD <u>7 12</u> 2 20. 5 4. 1 0.9 9 2.11 7 1.2 2 20. 5 4. 1 0.9 9 2.11 7 1.2 2 2.0. 5 4. 1 0.9 9 2.11 7 1.2 2 .1. 5 .551. 5 2.1. 7 4 2.4 8 2.5 8 2.7 8 2.7 8 2.7 8 2.7 8 2.7 8 2.7 8 2.7 8 2.7 8 2.5 8 2.7 8 2.7 8 2.7 8 2.7 8 2.7 8 2.5 8 2.7 8 2.7	5 36 177 3, df = - 32 39 116 20 34 96 20 357 , df = 6 0 16 20 34 96 20 357 , df = 6 0 177 3 4 4 5 5 5 2 3 4 3 4 3 4 3 6 9 116 20 3 4 3 4 3 5 5 5 5 5 5 3 4 3 4 3 5 5 5 5 5 5 5 5 5 5 5 5 5	al Mean (P < 0.0 aSyn Mean 6.564 (15.82 7 74.5 1.38 219.7 8.08 151.78 7 (P < 0.0 (P < 0.0 al Mea 1 9 0 12.7 0 10. 9 0.2 4 3,636. 9 6.0 3 6.0 3 10 (P < 0 10 (P < 0.0 10 (P < 0.0	75.84 0.01 00001); I ² in tota HC <u>SD</u> To 3.143 0.92 135.3 0.54 76.91 4.3 0.92 135.3 0.54 76.37 2 0001); I ² = HC <u>n SI</u> 6 111 2 6. 8 4.1 1 0.22 4 0.55 8 4.1 1 0.22 4 0.55 1.11 8 1.12 0.11 1.13 0.00001); I 1.13 0.00001); I 0.14 1.15 1.1	36 36 191 1 = 97% 1 all exact w all exact 1 block 1	20.6% 20.6% 20.6% 00.0% State St	4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 3.63 [1.69, 5.58] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.08 [3.25, 4.90] 3.63 [1.69, 5.58] 4.09 [5.26, 4.12] 0.13 [-0.33, 0.60] -4.69 [-5.26, -4.12] 0.13 [-0.49, 0.75] 0.09 [-0.41, 0.58] 6.04 [5.23, 6.85] 0.75 [0.06, 1.45] 0.26 [-1.65, 2.16] Cl C EXOSOMES Std. Mean Difference IV, Random, 95% Cl C e EXOSOMES Std. Mean Difference IV, Random, 95% Cl C e EXOSOMES Std. Mean Difference IV, Random, 95% Cl C e EXOSOMES Std. Mean Difference IV, Random, 95% Cl C e EXOSOMES Std. Mean Difference IV, Random, 95% Cl C e EXOSOMES Std. Mean Difference IV, Random, 95% Cl 6 0.03 [0.41, 1.46] C e Cost [-1.05, 0.05] 6 0.88 [0.45, 1.30] 6 0.59 [0.14, 1.0	-20 -10 0 10 : Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI -10 -5 0 5 10 Higher in HC Higher in PD Std. Mean Difference IV, Random, 95% CI

FIG. 2. Forest plots of aSyn (α -synuclein) and oligomeric aSyn in EVs (extracellular vesicles) and subgroup forest plots of aSyn in total and neuronalderived EVs. Forest plots showing the standardized mean difference of (**A**) aSyn and (**B**) oligomeric aSyn and subgroup forest plots showing the standardized mean differences of aSyn in (**C**) total and (**D**) L1CAM (L1 cell adhesion molecule)–positive neuronal-derived EVs. Standardized mean difference and their corresponding 95% confidence intervals are indicated by green squares. Standardized differences and 95% confidence intervals for each plot are represented by diamonds. [Color figure can be viewed at wileyonlinelibrary.com]

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aSyn in neuronal exosomes from serum

PD				HC			Std. Mean Difference	Std. Mean Difference		
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI	IV, Random, 95% CI	
Jiang 2020	29.32	20.5	230	12.72	6.1	144	25.6%	1.00 [0.78, 1.22]	-	
Jiang 2021	22.5	4.8	290	10.8	4.8	144	25.5%	2.43 [2.18, 2.69]		
Si 2019	0.2	0.13	49	0.27	0.16	18	24.2%	-0.50 [-1.05, 0.05]	•	
Yan 2022	7,556.3	5,551.3	44	3,636.8	3,100.6	49	24.8%	0.88 [0.45, 1.30]	•	
Total (95% CI)			613	-		355	100.0%	0.97 [-0.08, 2.02]	•	
Heterogeneity: Tau ² = Test for overall effect	= 1.11; Chi² : Z = 1.82 (l	= 126.05 P = 0.07)	, df = 3	(P < 0.00	001); I² =	98%			-10 -5 0 5 10 Higher in HC Higher in PD	

1	R)

aSyn in neuronal exosomes from plasma

(0)		PD			нс			Std. Mean Difference	Std. Mean I	Difference	
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% Cl	IV, Randor	m, 95% Cl	
Kluge 2022	1.51	0.97	30	1	0.24	50	14.3%	0.81 [0.34, 1.28]	·	•	
Niu 2020	2.39	2.15	53	0.64	0.58	21	14.3%	0.93 [0.41, 1.46]		•	
Shi 2014	25.07	1.8	264	11.8	0.78	215	14.2%	9.23 [8.62, 9.85]		+	
Yan 2022	7,556.3	5,551.3	44	3,636.8	3,100.6	49	14.4%	0.88 [0.45, 1.30]		•	
Zhao 2019	7.75	2.74	39	6.5	1.17	40	14.3%	0.59 [0.14, 1.04]		•	
Zhao 2020	139.6	21.74	30	67.69	11.2	30	14.0%	4.10 [3.19, 5.02]			
Zou 2020	7.84	2.46	93	6.07	1.35	85	14.4%	0.88 [0.57, 1.19]		•	
Total (95% CI)			553			490	100.0%	2.48 [0.55, 4.41]		•	
Heterogeneity: Tau ² =	6.70; Chi ²	= 690.54	. df = 6	(P < 0.00	001); l ² =	99%					_

Test for overall effect: Z = 2.52 (P = 0.01)





aSyn in neuronal exosomes from plasma - IP

(0)		PD			HC			Std. Mean Difference	Std. Mean Difference		
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI	IV, Random, 95% CI		
Jiang 2020	29.32	20.5	230	12.72	6.1	144	16.9%	1.00 [0.78, 1.22]	•		
Jiang 2021	22.5	4.8	290	10.8	4.8	144	16.9%	2.43 [2.18, 2.69]			
Niu 2020	2.39	2.15	53	0.64	0.58	21	16.7%	0.93 [0.41, 1.46]	-		
Shi 2014	25.07	1.8	264	11.8	0.78	215	16.6%	9.23 [8.62, 9.85]			
Zhao 2020	139.6	21.74	30	67.69	11.2	30	16.2%	4.10 [3.19, 5.02]	-		
Zou 2020	7.84	2.46	93	6.07	1.35	85	16.8%	0.88 [0.57, 1.19]	•		
Total (95% CI)			960			639	100.0%	3.08 [1.33, 4.82]	•		
Heterogeneity: Tau ² =	4.67; Ch	ni² = 704	.90, df	= 5 (P ·	< 0.00	001); l ²	= 99%	-			
Test for overall effect:	Z = 3.46	(P = 0.	0005)						Higher in HC Higher in PD		

(D)

 (\mathbf{C})

aSyn in neuronal exosomes from plasma - combined IP

(-)	PD				HC			Std. Mean Difference	Std. Mean Difference
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI	IV, Random, 95% CI
Dutta 2021	107	124	51	96	110	50	21.1%	0.09 [-0.30, 0.48]	•
Kluge 2022	1.51	0.97	30	1	0.24	50	19.8%	0.81 [0.34, 1.28]	•
Si 2019	0.2	0.13	49	0.27	0.16	18	18.5%	-0.50 [-1.05, 0.05]	•
Yan 2022	7,556.3	5,551.3	44	3,636.8	3,100.6	49	20.5%	0.88 [0.45, 1.30]	-
Zhao 2019	7.75	2.74	39	6.5	1.17	40	20.1%	0.59 [0.14, 1.04]	-
Total (95% CI)			213			207	100.0%	0.39 [-0.08, 0.85]	
Heterogeneity: Tau ² =	0.23; Chi ²	= 21.17,	df = 4 (P = 0.000	3); l ² = 81	1%		-	
Test for overall effect:	Z = 1.63 (P = 0.10)							Higher in HC Higher in PD

FIG. 3. Forest plots of aSyn in EVs (extracellular vesicles) from different sources and isolation methods. Forest plots showing the standardized mean differences in aSyn isolated from (A) serum and (B) plasma and further classification of plasma EVs isolated using (C) IP (immunoprecipitation) only and (D) IP combined with another method. Standardized mean differences and their corresponding 95% confidence intervals are indicated by green squares. Standardized differences and 95% confidence intervals for each plot are represented by diamonds. [Color figure can be viewed at wileyonlinelibrary.com]

(Δ)												
(~)		PD			HC		:	Std. Mean Difference		Std. Mean Difference		
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI		IV, Random, 95% CI		
Fraser 2016a	0.418	0.45	16	0.343	0.3	16	19.9%	0.19 [-0.50, 0.89]				
Ho 2014	0.249	0.087	26	0.23	0.23	21	26.9%	0.11 [-0.46, 0.69]				
Wang 2022 17.1 73.5 66 36 36 67 53.3% -0.33 [-0.67, 0.02]												
Total (95% CI)			108			104	100.0%	-0.11 [-0.45, 0.24]		•		
Heterogeneity: Tau ² =	0.03; Ch	ni² = 2.7	3, df = 3	2 (P = 0	.25); l ^a	² = 27%			+		-	
Test for overall effect: $Z = 0.61$ (P = 0.54)-2-1012Higher in HCHigher in PD												
phospho Ser1292 LRRK2												

(В)		PD			HC			Std. Mean Difference	Std. Mean Difference		
	Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% CI	IV, Random, 95% CI		
	Fraser 2016a	0.221	0.213	16	0.0299	0.11	16	29.8%	1.10 [0.35, 1.85]			
	Fraser 2016b	0.154	0.57	79	0.073	0.657	79	35.3%	0.13 [-0.18, 0.44]	+		
	Wang 2017	0.39	0.04	65	0.43	0.05	65	34.9%	-0.88 [-1.24, -0.52]	-		
	Total (95% CI)			160			160	100.0%	0.07 [-0.88, 1.01]	•		
	Heterogeneity: Tau ² =											
	Test for overall effect:	Higher in HC Higher in PD										

FIG. 4. Forest plots of LRRK2 (leucine-rich repeat kinase 2) and phospho Ser1292 LRRK2. Forest plots showing the standardized mean differences in (A) LRRK2 and (B) phospho Ser1292 LRRK2. Standardized mean differences and their corresponding 95% confidence intervals are indicated by green squares. Standardized differences and 95% confidence intervals for each plot are represented by diamonds. [Color figure can be viewed at wileyonlinelibrary.com]

Std.MD = 0.26, 0.95% CI = -1.65-2.16, P = 0.79) (Fig. 2C), whereas neuronal L1CAM-positive EVs showed a significant increase in the PD group (Std. 95% CI = 0.76 - 2.93, P = 0.0009) MD = 1.84, (Fig. 2D). Interestingly, further filtering based on the biological specimen indicated significance only for the neuronal EVs isolated from plasma (Std.MD = 2.48, 95% CI = 0.57-4.41, P = 0.01) (Fig. 3B) and not from (Std.MD = 0.97,95% CI = -0.08 - 2.02, serum P = 0.07) (Fig. 3A). Surprisingly, further analysis of the EV isolation method indicated the observed effects are mostly attributed to studies that performed direct IP from the plasma sample (Std.MD = 3.08, 95% CI = 1.33-4.82, P = 0.00001) (Fig. 3C) rather than combining IP following another isolation method (Std. MD = 0.39, 95% CI = -0.08 - 0.85, P = 0.0003) (Fig. 3D). Collectively, aSyn quantified in neuronal EVs isolated from blood plasma with L1CAM-IP seems to be the best method for differentiating PD and HC.

Meta-analysis of oligomeric aSyn levels showed significantly higher levels in PD patients compared to HCs (Std.MD = 3.36, 95% CI = 1.69–5.58, P = 0.0003) (Fig. 2B). Due to the small number of studies measuring oligomeric aSyn and the combination of biological specimens and methods used, subgroup analysis was not performed.

Meta-analysis of LRRK2 indicated no change in total LRRK2 among PD patients and HCs in urinal EVs (Std.MD = -0.11, 95% CI = -0.45-0.24, P = 0.54) (Fig. 4A) and no change in phospho Ser1292 LRRK2 (Std.MD = 0.07, 95% CI = -0.88-1.01, P = 0.89)

(Fig. 4B). For this meta-analysis, PD patients and HCs with and without LRRK2 mutation were pooled as only two of three studies reported data on these groups separately. In these two studies, LRRK2 levels differed among mutant and nonmutant carriers in both groups with higher levels of LRRK2 and phospho Ser1292 LRRK2 detected in urinal EVs of the mutant carriers. More studies will be needed to assess the diagnostic value of EV LRRK2 and phospho Ser1292 LRRK2 levels for LRRK2 mutant carriers.

Discussion

The development of a biomarker that will enable the early and accurate diagnosis of PD and support novel therapeutic approaches is an urgent need. Here we present reports on the collection and extraction of EV biomarkers from CSF, plasma, serum, urine, and saliva and present the optimal biomarker differentiating PD from healthy individuals. Our meta-analysis indicated that total aSyn in neuronal L1CAM-positive EVs isolated from blood plasma using IP is a promising biomarker. In addition, interesting results were found for oligomeric aSyn isolated from EVs irrespective of their origin and isolation method. Finally, although some studies assess LRRK2 levels in urinal EVs, the meta-analysis failed to identify a significant change between PD and HC (Fig. 5).

To date, there is only one additional study presenting a similar meta-analysis,¹⁰⁵ but we used different





FIG. 5. Overview of the protein-level changes in the EVs (extracellular vesicles) isolated from different biological specimens in PD (Parkinson's disease) patients compared to HCs (healthy controls). Oligomeric aSyn (oligo-aSyn) is increased in total EVs isolated from blood and saliva. Total aSyn levels are found increased in EVs isolated from blood samples, with higher levels observed in neuronal EVs isolated from plasma using direct immunoprecipitation. [Color figure can be viewed at wileyonlinelibrary.com]

exclusion criteria. Most studies on biomarker discovery focus on aSyn, which clearly plays a major role in PD pathogenesis.¹⁰⁶ Studies on aSyn levels in CSF indicate that aSyn is significantly lower in PD patients compared to controls but presented moderate specificity.¹⁰⁷⁻¹¹⁰ Conversely, studies on aSyn levels in blood samples show high variability.¹¹¹⁻¹¹³ This can be attributed to the fact that aSyn in the blood can originate from the aSyn-rich red blood cells or any peripheral tissue.^{114,115} aSyn has also been assessed in saliva and urine, but the diagnostic value of these biological specimens for aSyn remains uncertain.¹¹⁶⁻¹¹⁸

The presence of aSyn in EVs was first reported by Lee and colleagues¹¹⁹ followed by the characterization of those vesicles as EVs and the importance of this finding in aSyn spread and pathology.^{120,121} Using specific neuronal markers like L1CAM, it became clear that a portion of aSyn can be released from the CNS via the BBB to the blood and other peripheral biofluids.⁸¹

Assessing those EVs is like opening a window to the PD brain and allows us to differentiate brain-derived aSyn from aSyn from other tissues that may end up in the blood. Indeed, neuronal EVs showed the best results in differentiating PD, whereas total EVs isolated from blood showed no difference between PD and HC. This

indicates that we should measure neuronal-derived aSyn in the blood as opposed to total aSyn that also originates from other peripheral tissues. This suggests that the increased levels of aSyn in blood EVs are from neuronal EVs. The levels of aSyn in total blood EVs are not as specific as the neuronal-only aSyn, highlighting the importance of measuring specific types of EVs and not the entire blood pool. Interestingly, this is not the case for oligomeric aSyn, which is significantly increased in PD in total EVs isolated from saliva and total as well as neuronal EVs isolated from plasma or serum. This indicates oligomeric aSyn is a more robust biomarker as it is the most toxic form and is also implicated in disease propagation.^{121,122}

Saliva

The majority of studies assess EV aSyn in plasma and serum, whereas only a few assessed urine, CSF, and saliva EVs. We could compare only plasma and serum as EV sources. More studies are needed to assess the diagnostic potential of EVs from other fluids. We show that plasma EVs have a higher and significant increase in aSyn in PD compared to HC versus a smaller nonsignificant change that was found in serum EVs. The main difference between plasma and serum is clot formation. Different coagulants are used to prevent clotting in the case of plasma. Studies comparing EV isolation from plasma and serum showed that although there is no significant change in the number of EVs isolated from these two samples, there are many differences in EV contents, and the optimal source should be assessed based on the research questions.¹²³⁻¹²⁶ Although it is not clear why the EVs derived from plasma and serum differ, plasma EVs seem to be richer in protein contents and present more neuronal surface markers.^{124,126}

In addition to the EV source, the isolation method can introduce further variability. Each technique has advantages and disadvantages that are relevant to the research question. Classic ultracentrifugation has been the first isolation method used for EVs and is being gradually replaced by EV isolation kits that are based on precipitation principles. Although ultracentrifugation seems to be the most inexpensive and easiest method with a very high yield, it also requires more careful handling and purification steps to ensure that EV preparation is free of other vesicles and impurities in the form of aggregated protein and nucleic acids.¹²⁷ Exosome isolation kits, on the contrary, are more expensive and easier to use due to standardized procedures but have the disadvantage of impurities either from the sample itself or from the reagents employed by the kit for precipitation and isolation.^{127,128} Precipitation of EVs in such kits is performed either chemically or by affinity IP with the use of certain antibodies; precipitation-based kits seem to work well for RNA-seq studies, but affinity IP should be preferred in protein studies because chemical precipitation is mostly accountable for impurities such as sodium dodecyl sulfate, high-salt concentrations, and binding reagents that can hinder protein quality.¹²⁸ Similar to antibody-based total EV IP isolation employed by different kits, IP with specific antibodies to isolate the EVs of interest is a method gaining ground. The major advantages of this method are EV purity and high specificity and efficiency, and the disadvantages are the high costs and the use of antibodies that need careful optimization and add high variability.^{128,129} Despite presented results with L1CAM-IP, it is worth highlighting that L1CAM is specific for not only neuronal EVs but also EVs of different origin,¹³⁰ underlining that the use of antibodies for the isolation of EVs with specific origin needs to be better characterized and standardized among laboratories. Fewer studies are using SEC or FACS. Overall, SEC appears to be an inexpensive and effective method but requires a large volume of samples and produces a large volume of samples with highly diluted EVs, and the EVs passing through the column are subjected to high friction that can lead to bursting or loss of surface proteins, limiting the possibility of certain applications. Another limitation for SEC is sample purity, as lipoproteins with similar size to that of EVs can be isolated in parallel. SEC coupled with other

approaches like ultracentrifugation, ultrafiltration, and chemical IP can minimize this issue and increase sample purity.¹³¹ FACS, on the contrary, can be a very specific and efficient method for EV isolation but is not yet standardized.¹³² One would think that combining two methods would give better results, but along with the advantages of each method, the disadvantages might also add up. Our meta-analysis indicated that for aSynlevel assessment, it is preferred to perform L1CAM-IP.

It is better to perform IP on a straight sample rather than perform IP after another isolation method. This could be due to impurities introduced in the sample that can interfere with the antibody binding or clustering of several EVs that decrease specificity.

Heterogeneity calculated using I^2 was very high in our meta-analysis, indicating high variability among studies. We tried to eliminate heterogeneity by performing subgroup analyses, but the EV source and isolation method do not seem to eliminate variability. To assess the source of heterogeneity, we performed a regression analysis using relevant parameters like sex, age, disease duration, and disease severity in the form of MDS-UPDRS score during sample collection. We did not find a correlation between these parameters and the mean difference of aSyn (data not shown). This indicates that all these variables, in addition to individual cohort recruitment and patient selection and interlab variabilities in the experimental procedures, can be the source of heterogeneity. Only a few studies reported on participant inclusion and exclusion criteria for participant recruitment. Some studies accepted certain pathologies for the non-PD group and chose to record and monitor them, whereas others excluded individuals with many complications like hypertension, cardiovascular and cerebrovascular diseases, diabetes, inflammatory diseases, autoimmune diseases, kidney disease, thyroid problems, blood deficiencies, and anemia. These additional variabilities between studies may increase heterogeneity. In terms of experimental procedures, variability can be introduced at every step, starting with the use of different coagulants for plasma preparation to the biomarker measurement method. We did not assess the biomarker analysis methods as the studies included in the meta-analysis quantify aSyn using antibody-based methods (ELISA, ECL, and MSD). Thus, the variability is expected to be high and analogous to the sensitivity and specificity of the different antibodies used in the different assays. To eliminate this, or at least point out a definite variability source, more studies are necessary. Moreover, such meta-analyses aid the establishment of standard procedures for biomarker measurement and help create reference standards that aid biomarker development.

Measuring aSyn levels has been a promising biomarker, but other methods of analysis might be proven more useful and require further studies to validate their potential. For example, the ratio of oligomeric aSyn levels to the total levels provides a good measure of disease while eliminating errors of systemic errors in analysis methods.¹⁰² In addition, measuring the amount of aSyn-positive EVs in blood showed good sensitivity and specificity as well as correlation with disease measures.^{69,72} Finally, the potential of the aSyn species packed in EVs to form aggregates and propagate disease was assessed in two studies with ThT-based measures in a simple setup⁹⁶ or a more systematic way by performing SAA.⁶⁸

Although several biomarkers are reported in the literature, there are not sufficient studies to perform a meta-analysis for most of them. In short, aSyn remains interesting for PD diagnosis and also in EV studies that showed a higher aSyn concentration in neuronal blood EVs isolated using IP from plasma. Very few studies performed exploratory analysis, only six studies assessed EV protein contents using mass spectrometry, and only eight studies assessed RNA contents using NGS. Identifying novel targets is important for diagnosis and mechanistic studies as the EV cargo reflects the content of the cell of origin and can shed light on overseen mechanisms.

Data Availability Statement

Data sharing not applicable-no new data generated

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Supporting Data

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