

Supplementary Online Content

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This supplementary material has been provided by the authors to give readers additional information about their work.

eMethods 1. Modified version of the Newcastle-Ottawa Scale for Assessing the Quality of Nonrandomized Studies in Meta-Analyses, used to assess potential sources of bias.

Potential sources of bias were assessed using a modified version of the Newcastle-Ottawa Scale for Assessing the Quality of Nonrandomized Studies in Meta-Analyses. The studies are scored on:

Selection

1. **MDD Definition:** Is the case definition adequate?
 - A) Cases were defined as MDD according DSM or ICD criteria according to a validated assessment tool or by an experienced clinician
 - B) Cases were defined as MDD according DSM or ICD criteria but the method for assessing MDD status was not stated.
 - C) Cases were described as ‘clinically depressed’ but no further description was given.
2. **MDD Generality:** Was a General sample of cases tested?
 - A) A General sample of MDD was tested.
 - B) Recruitment of MDD cases was restricted to a specific sub-sample.
3. **HC Selection:** Selection of Controls
 - A) Controls were selected from the same population as cases
 - B) Controls were not selected from the same population as cases
 - C) No description
4. **HC Definition:** Definition of Controls
 - A) HC were defined clearly defined as having no current or past psychopathology
 - B) Controls were not clearly defined as having no current or past psychopathology.

Comparability (Comparability of cases and controls on the basis of the design or analysis)

1. **Does the study control for Age:** Yes/No/Unclear
2. **Does the study control for Gender:** Yes/No/Unclear
3. **Does the study control for IQ:** Yes/No/Unclear
4. **Does the study control for Personal or Household Income/Occupation:** Yes/No/Unclear

eTable 1. Quality ratings for ‘Selection’ component of the above rating scale.

Author, Year	Selection			
	MDD Definition	MDD Generality	HC selection	HC definition
Baek, 2017 ¹	A	B (suicide attempters)	B	A
Charpentier, 2017 ²	A	A	A	B
Chung, 2017 ³	A	A	C	B
Clarke, 2011 ⁴	A	B (elderly, suicide attempters)	B	A
Dombrovski, 2012 ⁵	A	B (elderly, suicide attempters)	C	A
Murphy, 2001 ⁶	A	A	B	B
Subramaniapillai, 2019 ⁷	A	A	C	A
Treadway, 2012 ⁸	A	A	B	A
Yang, 2014 ⁹	A	A	B	B
Admon, 2017 ¹⁰	A	A	A	A
Arrondo, 2015 ¹¹	A	A	B	B
Carl, 2016 ¹²	A	A	B	B
Chase, 2010 ¹³	A	A	B	A
DelDonno, 2015 ¹⁴	A	A	A	A
DelDonno, 2019a ¹⁵	A	A	A	A
DelDonno, 2019b ¹⁶	A	A	A	A
Pizzagalli, 2009 ¹⁷	A	A	B	A
Sankar, 2019 ¹⁸	A	B (women only)	A	A
Smoski, 2011 ¹⁹	A	A	B	B
Takamura, 2017 ²⁰	A	A	B	A
Xie, 2014 ²¹	A	A	B	A
Clery-Melin, 2011 ²²	A	A	B	A
Aylward, 2019 ²³	A	A	A	A
Henriques, 2000 ²⁴	A	A	A	A
Lawlor, 2019 ²⁵	A	A	B	B
Liu, 2011 ²⁶	A	A	B	B
Pizzagalli, 2008 ²⁷	A	A	B	A
Vrieze, 2013 ²⁸	A	A	B	A
Alexopoulos, 2015 ²⁹	A	B (elderly)	A	A
Cella, 2010 ³⁰	A	A	B	A
Deisenhammer, 2018 ³¹	A	B (suicide attempters)	B	B
Dezfouli, 2019 ³²	A	A	B	A
Gradin, 2011 ³³	A	A	A	B
Hall, 2014 ³⁴	A	A	B	A
Hegedus, 2018 ³⁵	A	B (suicide attempters)	B	A
Herzallah, 2013 ³⁶	A	B (melancholic)	B	A
Jollant, 2005 ³⁷	A	B (suicide attempters)	C	A
Jollant, 2016 ³⁸	A	A	C	A
Kumar, 2018 ³⁹	A	A	A	A
Liu, 2017 ⁴⁰	A	A	B	A
McGovern, 2014 ⁴¹	A	B (elderly)	A	A
Moutoussis, 2018 ⁴²	A	A	B	B
Must, 2006 ⁴³	A	A	B	A
Nord, 2018 ⁴⁴	A	A	B	A
Rothkirch, 2017 ⁴⁵	A	A	B	A
Saperia, 2019 ⁴⁶	A	A	C	A
Thoma, 2015 ⁴⁷	A	A	B	A
Walsh, 2018 ⁴⁸	A	A	A	A

MDD=Depressed Group; HC=Healthy Control Group.

Please refer to the above rating tool for the meanings of ‘MDD definition’, ‘MDD Generality’, ‘HC selection’ and ‘HC definition’ as well as ‘A’, ‘B’ and ‘C’ in each column. In question 2 (MDD Generality), where recruitment was restricted only to a specific subsample of MDD patients, this subsample is specified in parentheses.

eTable 2. Quality ratings for ‘Comparability’ component above.

Author, Year	Comparability			
	Age	Gender	IQ	Inc/Occ
Baek, 2017 ¹	Y	Y	N	N
Charpentier, 2017 ²	Y	Y	Y	N
Chung, 2017 ³	Y	N	Y	N
Clarke, 2011 ⁴	Y	Y	N	N
Dombrovski, 2012 ⁵	Y	Y	Y	N
Murphy, 2001 ⁶	Y	Y	Y	N
Subramaniapillai, 2019 ⁷	Y	Y	N	N
Treadway, 2012 ⁸	Y	Y	N	N
Yang, 2014 ⁹	Y	Y	N	N
Admon, 2017 ¹⁰	Y	Y	N	N
Arrondo, 2015 ¹¹	Y	Y	Y	N
Carl, 2016 ¹²	Y	Y	Y	N
Chase, 2010 ¹³	Y	Y	Y	N
DelDonno, 2015 ¹⁴	Y	Y	Y	N
DelDonno, 2019a ¹⁵	N	Y	Y	N
DelDonno, 2019b ¹⁶	Y	Y	Y	N
Pizzagalli, 2009 ¹⁷	Y	Y	N	N
Sankar, 2019 ¹⁸	Y	Y	N	N
Smoski, 2011 ¹⁹	Y	Y	Y	N
Takamura, 2017 ²⁰	Y	Y	Y	N
Xie, 2014 ²¹	Y	Y	N	N
Clery-Melin, 2011 ²²	Y	Y	N	N
Aylward, 2019 ²³	Y	Y	Y	N
Henriques, 2000 ²⁴	Y	Y	N	N
Lawlor, 2019 ²⁵	Y	Y	N	N
Liu, 2011 ²⁶	Y	Y	Y	N
Pizzagali, 2008 ²⁷	Y	Y	N	N
Vrieze, 2013 ²⁸	Y	Y	N	N
Alexopoulos, 2015 ²⁹	Y	N	N	N
Cella, 2010 ³⁰	Y	Y	N	N
Deisenhammer, 2018 ³¹	Y	Y	N	N
Dezfouli, 2019 ³²	Y	Y	Y	N
Gradin, 2011 ³³	Y	Y	Y	N
Hall, 2014 ³⁴	Y	Y	N	N
Hegedus, 2018 ³⁵	Y	Y	N	N
Herzallah, 2013 ³⁶	Y	Y	N	N
Jollant, 2005 ³⁷	Y	N	Y	N
Jollant, 2016 ³⁸	Y	Y	Y	N
Kumar, 2018 ³⁹	Y	Y	N	N
Liu, 2017 ⁴⁰	Y	Y	Y	N
McGovern, 2014 ⁴¹	Y	U	N	N
Moutoussis, 2018 ⁴²	Y	Y	N	N
Must, 2006 ⁴³	Y	Y	N	N
Nord, 2018 ⁴⁴	Y	Y	Y	N
Rothkirch, 2017 ⁴⁵	Y	Y	N	N
Saperia, 2019 ⁴⁶	Y	Y	N	N
Thoma, 2015 ⁴⁷	Y	Y	Y	N
Walsh, 2018 ⁴⁸	Y	Y	Y	N

IQ=Intelligence Quotient; Inc/Occ= Personal or Household Income/Occupation; Y=case and control groups were controlled for on the relevant measure; N= case and control groups were not controlled for on the relevant measure; U=unclear whether case and control groups were controlled for on the relevant measure.

eMethods 2. List of additional moderators used to assess the potential sources of bias.

In addition to those listed in eTables 1 and 2, other potential sources of bias were assessed using the following moderators:

- **Reward Processing Category:** Which of the four reward processing categories the studies belonged to.
- **Medication Status of MDD Group:** Whether the entire study sample was unmedicated.
- **Proportion of Females:** The proportion of female participants in the total study sample.
- **Mean Age:** The mean age of participants in the total study sample.
- **Risk-based OV task:** Within the Option Valuation category, whether the reward processing task was risk-based.
- **Difference in Anhedonia Score:** The difference in anhedonia questionnaire score between cases and controls (using standardized mean differences).
- **Difference in Cold Cognitive Task Performance:** The difference in performance on a cold cognitive task between cases and controls (using standardized mean differences).

eTable 3. Additional moderators used to assess the potential sources of bias.

Author, Year	Reward Processing Category	Medication Status of MDD Group	Proportion of Females	Mean Age	Risk-based OV task	Difference in Anhedonia Score		Difference in Cold Cognitive Task Performance	
						d	var _d	d	var _d
Baek, 2017 ¹	OV	M	0.449	24.2	Y	NA	NA	NA	NA
Charpentier, 2017 ²	OV	U	0.790	25.47	Y	NA	NA	NA	NA
Chung, 2017 ³	OV	M	0.684	35.46	Y	2.046	0.067	NA	NA
Clarke, 2011 ⁴	OV	M	0.495	68.5	Y	NA	NA	NA	NA
Dombrovski, 2012 ⁵	OV	M	0.543	68.675	Y	NA	NA	NA	NA
Murphy, 2001 ⁶	OV	M	0.563	37.9	Y	NA	NA	NA	NA
Subramaniapillai, 2019 ⁷	OV	M	0.537	39.985	N	1.177	0.115	0.470	0.100
Treadway, 2012 ⁸	OV	M	0.657	41.05	N	1.936	0.170	NA	NA
Yang, 2014 ⁹	OV	M	0.448	26.63	N	1.351	0.051	NA	NA
Admon, 2017 ¹⁰	RRV	U	0.723	25.8	NA	1.573	0.128	NA	NA
Arrondo, 2015 ¹¹	RRV	M	0.244	33.705	NA	1.432	0.112	NA	NA
Carl, 2016 ¹²	RRV	U	0.679	32.15	NA	NA	NA	NA	NA
Chase, 2010 ¹³	RRV	M	0.452	46.31	NA	NA	NA	NA	NA
DelDonno, 2015 ¹⁴	RRV	U	0.760	29.015	NA	1.351	0.120	0.677	0.104
DelDonno, 2019a ¹⁵	RRV	U	0.775	27.12	NA	1.051	0.201	NA	NA
DelDonno, 2019b ¹⁶	RRV	M	0.645	22.92	NA	NA	NA	NA	NA
Pizzagalli, 2009 ¹⁷	RRV	U	0.459	40.985	NA	NA	NA	NA	NA
Sankar, 2019 ¹⁸	RRV	M	1.000	30.125	NA	NA	NA	NA	NA
Smoski, 2011 ¹⁹	RRV	M	NA	30.3	NA	NA	NA	NA	NA
Takamura, 2017 ²⁰	RRV	M	0.500	41.15	NA	NA	NA	NA	NA
Xie, 2014 ²¹	RRV	M	0.667	31.915	NA	NA	NA	NA	NA
Clery-Melin, 2011 ²²	(no category)	M	0.708	44.05	NA	NA	NA	NA	NA
Aylward, 2019 ²³	RB	U	0.627	28.88	NA	NA	NA	0.283	0.060
Henriques, 2000 ²⁴	RB	U	0.606	33.055	NA	NA	NA	0.831	0.133
Lawlor, 2019 ²⁵	RB	U	0.656	37.03	NA	3.337	0.051	0.333	0.032
Liu, 2011 ²⁶	RB	U	0.540	27.8	NA	0.889	0.051	NA	NA
Pizzagali, 2008 ²⁷	RB	U	0.438	41.005	NA	3.756	0.230	NA	NA
Vrieze, 2013 ²⁸	RB	M	0.606	44.75	NA	2.506	0.051	NA	NA
Alexopoulos, 2015 ²⁹	RL	U	NA	72.505	NA	NA	NA	0.115	0.052
Cella, 2010 ³⁰	RL	M	0.462	35.45	NA	NA	NA	NA	NA
Deisenhammer, 2018 ³¹	RL	M	0.670	39.975	NA	NA	NA	NA	NA
Dezfouli, 2019 ³²	RL	M	0.559	22.6	NA	NA	NA	NA	NA
Gradin, 2011 ³³	RL	M	0.594	42.955	NA	NA	NA	NA	NA

Please refer to the list above for the description of the moderators.

OV=Option Valuation; RV=Response Vigour; RB=Reward Bias; RL=Reinforcement Learning; MDD=Depressed Group; U=Entirely Unmedicated Depressed Group; M=At Least Partially Medicated Depressed Group; Y=the Reward Processing OV Task was risk-based; N=the Reward Processing OV task was not risk-based; d=Cohen's d; Var_d=Variance on d.

eTable 3. Additional moderators used to assess the potential sources of bias (continued).

Author, Year	Reward Processing Category	Medication Status of MDD Group	Proportion of Females	Mean Age	Risk-based OV task	Difference in Anhedonia Score		Difference in Cold Cognitive Task Performance	
						d	var _d	d	var _d
Hall, 2014 ³⁴	RL	M	0.700	47	NA	NA	NA	NA	NA
Hegedus, 2018 ³⁵	RL	U	0.638	34.985	NA	NA	NA	NA	NA
Herzallah, 2013 ³⁶	RL	U	NA	27.865	NA	NA	NA	NA	NA
Jollant, 2005 ³⁷	RL	M	0.405	40.14	NA	NA	NA	NA	NA
Jollant, 2016 ³⁸	RL	U	0.593	35.5	NA	NA	NA	-0.301	0.076
Kumar, 2018 ³⁹	RL	U	0.745	25.78	NA	3.395	0.191	NA	NA
Liu, 2017 ⁴⁰	RL	U	0.540	29.5	NA	1.009	0.120	NA	NA
McGovern, 2014 ⁴¹	RL	U	NA	71.625	NA	NA	NA	NA	NA
Moutoussis, 2018 ⁴²	RL	M	0.541	33.66	NA	NA	NA	NA	NA
Must, 2006 ⁴³	RL	M	0.580	43.15	NA	NA	NA	NA	NA
Nord, 2018 ⁴⁴	RL	U	0.426	27.375	NA	1.771	0.103	NA	NA
Rothkirch, 2017 ⁴⁵	RL	U	0.638	36.225	NA	1.997	0.105	NA	NA
Saperia, 2019 ⁴⁶	RL	M	0.521	32	NA	1.330	0.052	-0.317	0.043
Thoma, 2015 ⁴⁷	RL	M	0.529	41.555	NA	NA	NA	NA	NA
Walsh, 2018 ⁴⁸	RL	U	0.705	29.865	NA	2.874	0.092	NA	NA

Please refer to the list above for the description of the moderators.

OV=Option Valuation; RV=Response Vigour; RB=Reward Bias; RL=Reinforcement Learning; MDD=Depressed Group; U=Entirely Unmedicated Depressed Group; M=At Least Partially Medicated Depressed Group; Y=the Reward Processing OV Task was risk-based; N=the Reward Processing OV task was not risk-based; d=Cohen's d; Var_d=Variance on d.

eTable 4. Tasks and Measures Included in the Meta-Analysis.

Author, Year	Task	Measure
Baek, 2017 ¹	Gambling Task	Risk Aversion in Gain Condition
Charpentier, 2017 ²	Gambling Task	Risk Aversion
Chung, 2017 ³	Risky Decision Making task	Risk Preference
Clarke, 2011 ⁴	Cambridge Gambling Task	Risk Adjustment
Dombrovski, 2012 ⁵	Cambridge Gambling Task	Risk Adjustment
Murphy, 2001 ⁶	Cambridge Gambling Task	Risk Adjustment
Subramaniapillai, 2019 ⁷	Effort Expenditure for Reward Task	Mean Proportion trials in which High Cost/High Reward chosen
Treadway, 2012 ⁸	Effort Expenditure for Reward Task	Mean Proportion trials in which High Cost/High Reward chosen
Yang, 2014 ⁹	Effort Expenditure for Reward Task	Mean Proportion trials in which High Cost/High Reward chosen
Admon, 2017 ¹⁰	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Arrondo, 2015 ¹¹	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Carl, 2016 ¹²	Monetary Incentive Delay	Group x Trial interaction; Reaction time
Chase, 2010 ¹³	Choice Reinforcement Reaction Time Task	Winnings
DelDonno, 2015 ¹⁴	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
DelDonno, 2019a ¹⁵	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
DelDonno, 2019b ¹⁶	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Pizzagalli, 2009 ¹⁷	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Sankar, 2019 ¹⁸	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Smoski, 2011 ¹⁹	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Takamura, 2017 ²⁰	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Xie, 2014 ²¹	Monetary Incentive Delay	Neutral Reaction Time minus Reward Reaction Time
Clery-Melin, 2011 ²²	Grip Force Task	Force Production
Aylward, 2019 ²³	Tone Identification Task	Response Bias
Henriques, 2000 ²⁴	Signal Detection Task (words)	Response Bias, Neutral vs Reward Condition
Lawlor, 2019 ²⁵	Probabilistic Reward Task	Response Bias, 2nd block
Liu, 2011 ²⁶	Probabilistic Reward Task	Response Bias, block 3, no-stress condition
Pizzagali, 2008 ²⁷	Probabilistic Reward Task	Response Bias, block 3
Vrieze, 2013 ²⁸	Probabilistic Reward Task	Response Bias, block 3
Alexopoulos, 2015 ²⁹	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks
Cella, 2010 ³⁰	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks, phase 1, last block
Deisenhammer, 2018 ³¹	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks
Dezfouli, 2019 ³²	Instrumental Learning Task	Proportion, Higher Reward Probability Option Chosen
Gradin, 2011 ³³	Instrumental Learning Task	Number of Scores

eTable 4. Tasks and Measures Included in the Meta-Analysis (continued).

Author, Year	Task	Measure
Hall, 2014 ³⁴	Contingency Reversal Reward Paradigm	Number of Successful Switches
Hegedus, 2018 ³⁵	Iowa Gambling Task	ABCD deck, block 5 score (Advantageous Decks minus Disadvantageous Decks)
Herzallah, 2013 ³⁶	Instrumental Learning Task	% of Correct Responses in Block 4 for Reward Trials
Jollant, 2005 ³⁷	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks, last block
Jollant, 2016 ³⁸	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks in second half of task (last 50 trials)
Kumar, 2018 ³⁹	probabilistic instrumental learning task	% Choice Accuracy
Liu, 2017 ⁴⁰	Reinforcement Learning task	% High-Probability Stimuli Chosen (reward trials)
McGovern, 2014 ⁴¹	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks, last block
Moutoussis, 2018 ⁴²	Go/No-Go task	Appetitive Motivational Exchange Rate for Rewards
Must, 2006 ⁴³	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks, first phase, last block
Nord, 2018 ⁴⁴	Go/No-Go task	Probability of Making a 'Go' response
Rothkirch, 2017 ⁴⁵	Reinforcement Learning Task	Optimal Choice Rate, 5th block, Reward Trials
Saperia, 2019 ⁴⁶	Iowa Gambling Task	Advantageous Decks minus Disadvantageous Decks, last block.
Thoma, 2015 ⁴⁷	Probabilistic Selection Task (active)	% Correct Responses, block 4
Walsh, 2018 ⁴⁸	Probabilistic Instrumental Learning Task	Money Won in Win Trials at Baseline Session

eTable 5. Reward Processing Categories into Which the Studies were Categorised, Medication Status of the Depressed Group and Sample Sizes.

Author, Year	Reward Processing Category	Sample Size (HC)	Sample Size (MDD)
Baek, 2017 ¹	OV	75	92
Charpentier, 2017 ²	OV	23	13
Chung, 2017 ³	OV	28	47
Clarke, 2011 ⁴	OV	22	73
Dombrovski, 2012 ⁵	OV	15	31
Murphy, 2001 ⁶	OV	26	22
Subramaniapillai, 2019 ⁷	OV	20	21
Treadway, 2012 ⁸	OV	15	20
Yang, 2014 ⁹	OV	50	46
Admon, 2017 ¹⁰	RRV	19	22
Arrondo, 2015 ¹¹	RRV	21	24
Carl, 2016 ¹²	RRV	20	33
Chase, 2010 ¹³	RRV	21	21
DelDonno, 2015 ¹⁴	RRV	22	19
DelDonno, 2019a ¹⁵	RRV	15	9
DelDonno, 2019b ¹⁶	RRV	35	11
Pizzagalli, 2009 ¹⁷	RRV	31	30
Sankar, 2019 ¹⁸	RRV	20	20
Smoski, 2011 ¹⁹	RRV	13	9
Takamura, 2017 ²⁰	RRV	12	12
Xie, 2014 ²¹	RRV	20	40
Aylward, 2019 ²³	RB	47	26
Henriques, 2000 ²⁴	RB	15	18
Lawlor, 2019 ²⁵	RB	36	258
Liu, 2011 ²⁶	RB	44	43
Pizzagalli, 2008 ²⁷	RB	25	23
Vrieze, 2013 ²⁸	RB	63	79
Clery-Melin, 2011 ²²	(no category)	26	22
Alexopoulos, 2015 ²⁹	RL	30	53
Cella, 2010 ³⁰	RL	20	19
Deisenhammer, 2018 ³¹	RL	26	48
Dezfouli, 2019 ³²	RL	34	34
Gradin, 2011 ³³	RL	17	15
Hall, 2014 ³⁴	RL	15	15
Hegedus, 2018 ³⁵	RL	46	59
Herzallah, 2013 ³⁶	RL	22	13
Jollant, 2005 ³⁷	RL	82	66
Jollant, 2016 ³⁸	RL	30	24
Kumar, 2018 ³⁹	RL	26	25
Liu, 2017 ⁴⁰	RL	17	21
McGovern, 2014 ⁴¹	RL	36	60
Moutoussis, 2018 ⁴²	RL	22	39
Must, 2006 ⁴³	RL	20	30
Nord, 2018 ⁴⁴	RL	28	26
Rothkirch, 2017 ⁴⁵	RL	28	29
Saperia, 2019 ⁴⁶	RL	51	43
Thoma, 2015 ⁴⁷	RL	16	18
Walsh, 2018 ⁴⁸	RL	42	46

OV=Option Valuation; RRV=Reward Response Vigour; RB=Reward Bias; RL=Reinforcement Learning; MDD=Depressed Group; HC=Healthy Control Group; U=Entirely Unmedicated Depressed Group; M=At Least Partially Medicated Depressed Group.

eMethods 3. Formulae used to convert study measures into Cohen's ds and associated variances⁴⁹

The following formulae were used to convert study measures into Cohen's ds and associated variances:

$$d = \frac{M_1 - M_2}{SD_{pooled}}$$

Equation 1. Cohen's d from Means and Standard Deviations of 2 samples.

d is Cohen's d, M_1 is the mean of one sample M_2 is the mean of the other sample, SD_{pooled} is the pooled standard deviation of the two samples (please see below.)

$$SD_{pooled} = \sqrt{\frac{(N_1 - 1)SD_1^2 + (N_2 - 1)SD_2^2}{N_1 + N_2 - 2}}$$

Equation 2. Pooled standard deviation of 2 samples.

SD_{pooled} is the pooled standard deviation of the two samples, N_1 is the size of one sample N_2 is the size of the other sample, SD_1 is the standard deviation of one sample, SD_2 is the standard deviation of the other sample.

$$d = \frac{t}{\sqrt{\frac{1}{\frac{1}{N_1} + \frac{1}{N_2}}}}$$

Equation 3. Cohen's d from t-statistic.

d is Cohen's d, N_1 is the size of one sample N_2 is the size of the other sample, t is the t-statistic.

$$d = \sqrt{\frac{F}{\frac{1}{\frac{1}{N_1} + \frac{1}{N_2}}}}$$

Equation 4. Cohen's d from F-statistic.

d is Cohen's d, N_1 is the size of one sample N_2 is the size of the other sample, F is the F-statistic.

$$Var_d = \frac{N_1 + N_2}{N_1 \times N_2} + \frac{d^2}{2(N_1 + N_2)}$$

Equation 5. Variance on Cohen's d.

Var_d is the Variance on Cohen's d, N_1 is the size of one sample N_2 is the size of the other sample, d is Cohen's d.

eTable 6. The summary statistic(s) that were converted to a Cohen's d, Cohen's d and the Variance on 'd' for each included study.

Author, Year	t	F	M (HC)	SD (HC)	M (MDD)	SD (MDD)	d	Var _d
Baek, 2017 ¹			0.289	0.199	0.339	0.301	0.191	0.024
Charpentier, 2017 ²			0.311	0.583	0.565	0.458	0.469	0.123
Chung, 2017 ³			0.500	0.310	0.460	0.310	0.129	0.057
Clarke, 2011 ⁴			0.920	0.725	0.672	0.876	0.294	0.060
Dombrovski, 2012 ⁵			0.608	0.608	0.624	0.671	-0.024	0.099
Murphy, 2001 ⁶			1.569	0.692	1.194	0.945	0.459	0.086
Subramaniapillai, 2019 ⁷			0.570	0.321	0.380	0.224	0.690	0.103
Treadway, 2012 ⁸			0.542	0.170	0.428	0.098	0.856	0.127
Yang, 2014 ⁹			0.682	0.286	0.580	0.323	0.334	0.042
Admon, 2017 ¹⁰			34.950	37.142	36.990	29.680	-0.061	0.098
Arrondo, 2015 ¹¹			0.035	0.038	0.036	0.042	-0.025	0.089
Carl, 2016 ¹²		0.330					0.163	0.081
Chase, 2010 ¹³		5.637					-0.733	0.102
DelDonno, 2015 ¹⁴			-13.929	63.831	0.848	48.182	-0.259	0.099
DelDonno, 2019a ¹⁵			18.159	23.440	4.347	13.432	0.678	0.187
DelDonno, 2019b ¹⁶			5.359	15.226	5.489	14.828	-0.009	0.119
Pizzagalli, 2009 ¹⁷	2.090						0.535	0.068
Sankar, 2019 ¹⁸			7.310	21.650	4.270	26.460	0.126	0.100
Smoski, 2011 ¹⁹			23.880	23.170	28.310	28.040	-0.176	0.189
Takamura, 2017 ²⁰			23.585	19.710	6.043	18.493	0.918	0.184
Xie, 2014 ²¹			5.610	15.810	4.175	11.483	0.110	0.075
Clery-Melin, 2011 ²²			0.171	0.190	0.003	0.047	1.173	0.098
Aylward, 2019 ²³			0.533	0.166	0.411	0.149	0.757	0.064
Henriques, 2000 ²⁴	2.380						0.832	0.133
Lawlor, 2019 ²⁵			0.106	0.170	0.102	0.208	0.021	0.032
Liu, 2011 ²⁶			0.125	0.167	-0.068	0.128	1.298	0.056
Pizzagali, 2008 ²⁷			0.228	0.171	0.101	0.184	0.716	0.089
Vrieze, 2013 ²⁸	2.450						0.413	0.029
Alexopoulos, 2015 ²⁹			0.07*	0.96	-0.04*	1.03	0.109	0.052
Cella, 2010 ³⁰			4.510	8.810	0.230	8.587	0.492	0.106
Deisenhammer, 2018 ³¹			11.200	32.100	15.700	24.080	-0.166	0.059
Dezfouli, 2019 ³²			0.635	0.110	0.575	0.116	0.534	0.061
Gradin, 2011 ³³			50.590	1.800	48.500	1.190	1.352	0.154
Hall, 2014 ³⁴		3.87					0.718	0.142
Hegedus, 2018 ³⁵		4.171					0.402	0.039
Herzallah, 2013 ³⁶			80.040	16.280	57.230	28.300	1.064	0.139
Jollant, 2005 ³⁷			6.317	30.192	0.394	23.145	0.217	0.028
Jollant, 2016 ³⁸			15.200	19.500	7.300	26.700	0.344	0.076

t=t-statistic; F=F-statistic; M=mean, SD=standard deviation; MDD=Depressed Group; HC=Healthy Control Group; d=Cohen's d; Var_d=Variance on d; *Z-scores.

eTable 6. The summary statistic(s) that were converted to a Cohen's d, Cohen's d and the Variance on 'd' for each included study (continued).

Author, Year	t	F	M (HC)	SD (HC)	M (MDD)	SD (MDD)	d	Var _d
Kumar, 2018 ³⁹			87.820	12.510	80.370	19.180	0.462	0.081
Liu, 2017 ⁴⁰			72.500	15.6	67.300	19.600	0.290	0.108
McGovern, 2014 ⁴¹			0.780	1.470	1.800	1.050	-0.834	0.048
Moutoussis, 2018 ⁴²	0.277						0.074	0.071
Must, 2006 ⁴³			6.400	4.510	-1.700	4.600	1.775	0.115
Nord, 2018 ⁴⁴			75.100	15.124	64.120	17.888	0.665	0.078
Rothkirch, 2017 ⁴⁵	1.010						-0.268	0.071
Saperia, 2019 ⁴⁶			4.400	10.800	1.800	12.900	0.220	0.043
Thoma, 2015 ⁴⁷			42.000	32.100	33.830	32.700	0.252	0.119
Walsh, 2018 ⁴⁸	0.690						0.147	0.046

t=t-statistic; F=F-statistic; M=mean, SD=standard deviation; MDD=Depressed Group; HC=Healthy Control Group; d=Cohen's d; Var_d=Variance on d.

eResults. Moderator Analyses.

Of the total variation in effect sizes, 68% was due to between-study differences. The medication status of the MDD sample (unmedicated *vs* at least partially medicated) explained 0% of the variance in the global effect size ($p=0.999$). Additional models revealed no effect of medication status in any sub-component category. Those studies that clearly matched the case and control groups for gender ($n=44$) yielded a larger effect to those that didn't ($p < 0.0001$). This was influenced by one study where it was unclear whether they matched for gender⁴¹ (SMD = -0.83 95% CI: -1.26 to -0.40). However, a significant difference remained between those studies that clearly balanced for gender (SMD = 0.389, 95% CI: 0.258 to 0.519, $n=44$) and those that didn't (SMD = 0.156, 95% CI: -0.289 to 0.599, $n=3$). Within the reward processing subcategories, gender-matching was only a moderator in the *Reinforcement Learning* subcategory ($p<0.0002$). As with the overall result, the one study in which it is unclear whether they balanced for gender⁴¹ yielded the lowest effect size (please see above) while those studies that did not match for gender ($n=2$) yielded a lower summary effect SMD = 0.167, 95% CI: -0.397 to 0.732) than those that did ($n=17$, SMD = 0.444, 95% CI: 0.226 to 0.662). However, the proportion of female participants in the total study population did not have any effect on the reward processing effect size ($p=0.840$).

Studies that tested an exclusively elderly sample ($n=4$), yielded a smaller ($p<0.001$) and non-significant effect (SMD=-0.127, 95%CI: -0.555-0.300) than those that did not ($n=44$, SMD=0.390, 95%CI: 0.254-0.525). Again, within the reward processing subcategories, this was only significant for *Reinforcement Learning* ($p < 0.0005$) where those that tested elderly samples ($n=2$), yielded a lower (and insignificant: SMD=-0.367, 95% CI: -0.990 to 0.256) effect compared to the non-elderly samples ($n=18$) (SMD = 0.432, 95% CI: 0.209 to 0.654). In those studies that reported group anhedonia scores ($n=17$) or the result of a cold-cognitive task ($n=8$) we did not find any significant effects of either anhedonia or cold cognition on reward processing impairment (either overall, or in any subcategory). All other moderation analyses were non-significant.

Supplementary References

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