



# Remote Ischemic Preconditioning to Prevent Acute Kidney Injury After Cardiac Surgery: A Meta-Analysis of Randomized Controlled Trials

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Liu Z, Zhao Y, Lei M, Zhao G, Li D, Sun R and Liu X (2021) Remote Ischemic Preconditioning to Prevent Acute Kidney Injury After Cardiac Surgery: A Meta-Analysis of Randomized Controlled Trials. Front. Cardiovasc. Med. 8:601470. doi: 10.3389/fcvm.2021.601470 **Objective:** Randomized controlled trials (RCTs) evaluating the influence of remote ischemic preconditioning (RIPC) on acute kidney injury (AKI) after cardiac surgery showed inconsistent results. We performed a meta-analysis to evaluate the efficacy of RIPC on AKI after cardiac surgery.

# **Methods:** Relevant studies were obtained by search of PubMed, Embase, and Cochrane's Library databases. A random-effect model was used to pool the results. Meta-regression and subgroup analyses were used to determine the source of heterogeneity.

**Results:** Twenty-two RCTs with 5,389 patients who received cardiac surgery -2,702 patients in the RIPC group and 2,687 patients in the control group—were included. Moderate heterogeneity was detected (*p* for Cochrane's *Q* test = 0.03,  $l^2 = 40\%$ ). Pooled results showed that RIPC significantly reduced the incidence of AKI compared with control [odds ratio (OR): 0.76, 95% confidence intervals (CI): 0.61–0.94, *p* = 0.01]. Results limited to on-pump surgery (OR: 0.78, 95% CI: 0.64–0.95, *p* = 0.01) or studies with acute RIPC (OR: 0.78, 95% CI: 0.63–0.97, *p* = 0.03) showed consistent results. Meta-regression and subgroup analyses indicated that study characteristics, including study design, country, age, gender, diabetic status, surgery type, use of propofol or volatile anesthetics, cross-clamp time, RIPC protocol, definition of AKI, and sample size did not significantly affect the outcome of AKI. Results of stratified analysis showed that RIPC significantly reduced the risk of mild-to-moderate AKI that did not require renal replacement therapy (RRT, OR: 0.76, 95% CI: 0.60–0.96, *p* = 0.02) but did not significantly reduce the risk of severe AKI that required RRT in patients after cardiac surgery (OR: 0.73, 95% CI: 0.50–1.07, *p* = 0.11).

**Conclusions:** Current evidence supports RIPC as an effective strategy to prevent AKI after cardiac surgery, which seems to be mainly driven by the reduced mild-to-moderate AKI events that did not require RRT. Efforts are needed to determine the influences of patient characteristics, procedure, perioperative drugs, and RIPC protocol on the outcome.

Keywords: remote ischemic preconditioning, acute kidney injury, cardiac surgery, off-pump, meta-analysis

# INTRODUCTION

Acute kidney injury (AKI) is common for patients after cardiac surgery, particularly for patients who receive cardiac surgery with complex procedures, such as on-pump surgery, concomitant coronary artery bypass graft (CABG) with valvular surgery, longer aortic cross-clamp time during cardiopulmonary bypass, and surgeries of open total aortic arch replacement (1-3). Previous studies showed that patients with postoperative AKI, even of mild degree, had worse clinical outcome after cardiac surgery (4). Currently, multiple criteria have been applied to define AKI after cardiac surgery, while consensus regarding the optimal definition of AKI remains lacking (5). Among which, the Acute Kidney Injury Network (AKIN) and the Risk, Injury, Failure, Loss, End Stage Kidney Disease (RIFLE) have been the mostly used criteria to define AKI, while the Kidney Disease: Improving Global Outcomes (KDIGO) criteria for AKI staging has been shown to demonstrate greater sensitivity to detect AKI and to predict associated in-hospital mortality, than do the RIFLE or AKIN criteria (5). Since treatment options for postoperative AKI remain limited, identification of strategies to prevent the incidence of AKI after cardiac surgery are clinically important (6). Remote ischemic preconditioning (RIPC), which is known as a strategy that protects the target organ by inducing brief episodes of ischemia and reperfusion in distant tissue, has been suggested to be effective for preventing AKI (7-9). However, randomized clinical trials (RCTs) evaluating the influence of RIPC on AKI after cardiac surgery showed inconsistent results (10-31). Since most of these studies were of limited sample size, the individual studies may have been statistically insufficient to detect a significant influence of RIPC on postoperative AKI. Furthermore, several previous meta-analyses that pooled studies published through 2016 showed that RIPC was not effective at preventing AKI in patients after on-pump cardiac surgery (32-34). Although, these results were also inconsistent (35). Many RCTs have been published since 2016 (22, 24-31), warranting an updated meta-analysis. In addition, the potential impacts of study and patient features on the efficacy of RIPC on postoperative AKI in these patients were rarely analyzed. Therefore, we performed an updated meta-analysis with comprehensive subgroup analyses to investigate the role of RIPC to prevent AKI after cardiac surgery.

# METHODS

The Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) statement (36) and the Cochrane Handbook guidelines (37) were followed during the designing and implementation of the study.

#### Search Strategy

PubMed, Embase, and the Cochrane Library (Cochrane Center Register of Controlled Trials) databases were searched for relevant studies with a combined strategy of: [1] "ischemic preconditioning" OR "remote ischemic preconditioning" OR "RIPC"; [2] "cardiac surgery" OR "coronary artery bypass" OR "surgical coronary revascularization" OR "valve surgery" OR "valve replacement"; and [3] "random" OR "randomized" OR "randomized" OR "randomly." Only clinical studies were considered. The references of related reviews and original articles were also searched as a complementation. The latest database search was conducted on 5th May 2020.

# **Study Selection**

Inclusion criteria were as follows: [1] peer-reviewed articles in English or Chinese; [2] designed as parallel-group RCTs; [3] included adult patients scheduled for open heart surgery who were randomly allocated to a RIPC treatment group or a control group; and [4] reported the incidence of AKI in the perioperative periods. Reviews, studies including children or neonates, preclinical studies, observational studies, and repeated reports were excluded.

# **Data Extraction and Quality Assessment**

Study search, data extraction, and quality evaluation were achieved by two independent authors. If disagreement occurred, it was resolved by consensus between the two authors. We extracted data regarding study information (first author, publication year, and study country), study design (blind or open-label), patient information (number of participants, mean age, gender, diabetic status, and patients with severe cardiac systolic dysfunction), surgery type, perioperative anesthetics (use of propofol or volatile anesthetics, and cross-clamp time), RIPC protocol, and definition of AKI. Quality evaluation was achieved using the Cochrane's Risk of Bias Tool (37) according to the following aspects: [1] random sequence generation; [2] allocation concealment; [3] blinding of participants and personnel; [4] blinding of outcome assessors; [5] incomplete outcome data; [6] selective outcome reporting; and [7] other potential bias.

# **Statistical Analysis**

Incidence of AKI in each arm was evaluated via odds ratio (OR) and its 95% confidence intervals (CIs). We used the Cochrane's Q test to detect the heterogeneity, and significant heterogeneity was suggested if p < 0.10 (38). The  $I^2$  statistic was also calculated, and an  $I^2 > 50\%$  reflected significant heterogeneity. Pooled analyses were calculated using a randomeffect model because this method incorporates the influence of potential heterogeneity and retrieves a more generalized result (37). Stratified analyses comparing the results in complex and simple cardiac surgeries were also performed. We defined complex surgeries as double-valve or triple-valve surgery, mitral valve surgery, coronary artery bypass graft (CABG) plus valve(s), open total aortic arch replacement, or any "redo" operation. Accordingly, CABG or single-valve surgeries were defined as simple surgeries. Predefined meta-regression analyses, sensitivity analyses, and subgroup analyses were performed to explore the potential influences of study characteristics on the outcome. These characteristics included study design, country, age, gender, diabetic status, surgery type, use of propofol, or volatile anesthetics, cross-clamp time, RIPC protocol, definition of AKI, and sample size of the RCT (39). For continuous variables, medians were used for cut-off. To evaluate the influence of RIPC on the AKI events with different severity, we defined patients with mild-to-moderate AKI as those that did not require renal replacement therapy (RRT) and those with severe AKI as patients that required RRT. Stratified analysis was performed accordingly. Publication bias was evaluated by visual inspection of funnel plots, and the Egger's regression asymmetry test (40). p < 0.05 were considered statistically significant. The RevMan (Version 5.1; Cochrane, Oxford, UK) and Stata software (Version 12.0; Stata, College Station, TX) were applied for statistical analyses.

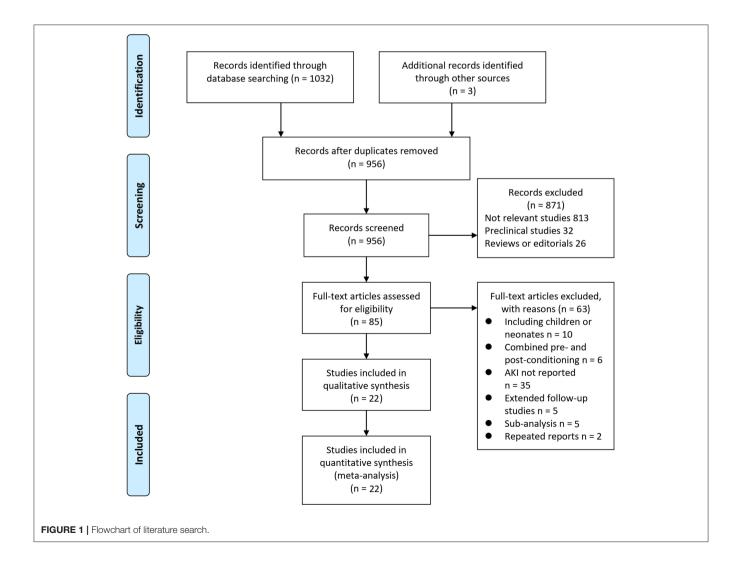
# RESULTS

# **Search Results**

In summary, 1,035 articles were obtained through the database search. After exclusion of duplicate studies, 956 articles were screened. Among them, 871 articles were subsequently excluded based on titles and abstracts primarily because these studies were irrelevant. Among the 85 potentially relevant articles, 63 were further excluded via full-text review based on reasons listed in **Figure 1**. Finally, 22 RCTs (10–31) were included.

#### **Study Characteristics**

Table 1 shows the characteristics of the included studies. Overall, 22 RCTs (10-31) with 5,389 patients who received cardiac surgery -2,702 patients in the RIPC group and 2,687 patients in the control group-were included. These studies were published between 2010 and 2020. Patients with normal kidney function at baseline were included except one study that included patients with chronic kidney disease (CKD) who were scheduled for cardiac surgery (17). With regard to the surgery procedures, 20 studies included patients who received on-pump surgeries (10-28, 30), while the remaining two studies included patients who received off-pump surgeries (29, 31). Proportions of diabetic patients varied among the included studies, and patients with severe cardiac systolic function (left ventricular ejection fraction <30%) were rarely included. The use of propofol and volatile anesthetics varied among the patients of the included studies. RIPC was performed after anesthesia induction and before CPB in most of the included studies (acute RIPC), except in one study in which RIPC was performed 24-48 h before the surgery (chronic RIPC) (25). The protocol of RIPC included 3-4 cycles of upper or lower limb ischemia (5-10 min blood pressure cuff



#### TABLE 1 | Characteristics of the included studies.

Study	Country	Design	Surgical procedure	No. of patients	Mean age (years)	Male (%)	DM (%)	Baseline eGFR (ml/min)	LVEF <30% (%)	Propofol used	Volatile anesthetics	Cross- clamp time (min)	Protocols of RIPC	Control	Definition of AKI
Venugopal et al. (10)	UK	R, SB	On-pump CABG with or without AVR	78	65.1	82.1	0	NR	2.6	Partial	61.5	52	UL, 200 mmHg, 5 min × 3, after anesthesia induction and before CPB	Uninflated cu	IIF AKIN
Choi et al. (2011)	South Korea	R, DB	On-pump complex valvular heart surgery	76	68.5	39.5	6.6	78.5	0	None	100	103	LL, 250 mmHg, 10 min $\times$ 3, after anesthesia induction and before CPB	Uninflated cu	Iff AKIN
Zimmerman et al. (2011)	USA	R, SB	On-pump CABG with or without AVR	118	63.5	68.5	38.1	NR	10.2	NR	100	71	LL, 200 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated cu	Iff AKIN
Young et al. (13)	New Zealand	R, DB	On-pump complex heart surgery	96	65.0	62.5	5.2	NR	4.2	All	100	111	UL, 200 mmHg, 5 min $\times$ 3, after an esthesia induction and before CPB	Uninflated cu	iff RIFLE
Meybohm et al. (14)	Germany	R, DB	On-pump heart surgery	180	69.0	81.2	21.1	NR	0	All	0	80	UL, 200 mmHg, 5 min $\times$ 4, after anesthesia induction and before CPB	Uninflated cu	IIF AKIN
Wang et al. (15)	China	R, DB	On-pump valvular heart surgery	31	49.4	33.1	0	NR	NR	NR	NR	54	LL, 600 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated cu	Iff AKIN
Candilio et al. (16)	UK	R, DB	On-pump CABG with or without valvular surgery	178	65.5	78.0	29.5	NR	4.5	All	100	63	UL and LL, 200 mmHg, $5 \text{ min} \times 2$ , after anesthesia induction and before CPB	Uninflated cu	Iff AKIN
Zarbock et al. (20)	Germany	R, DB	On-pump heart surgery	240	70.4	62.9	37.5	56.6	15	None	100	82	UL, 200 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated cu	iff KDIGO
Gallagher et al. (17)	UK	R, SB	CKD patients that received on-pump CABG with or without AVR	86	70.8	80.2	64.0	51	10.4	NR	87.2	62	UL, SBP + 50 mmHg, 5 mir × 3, after anesthesia induction and before CPB	า Uninflated cu	Iff AKIN
Meybohm et al. (19)	Germany	R, DB	On-pump heart surgery	1,385	65.9	74.2	24.9	NR	0	All	0	77	UL, 200 mmHg, 5 min $\times$ 4, after an esthesia induction and before CPB	Uninflated cu	iff RIFLE
Hausenloy et al. (18)	UK	R, DB	On-pump CABG with or without valvular surgery	1,612	76.2	71.6	25.9	NR	NR	All	42.1	70	UL, 200 mmHg, $5 \text{ min} \times 4$ , after anesthesia induction and before CPB	Uninflated cu	iff KDIGO
Walsh et al. (24)	Canada, USA, India, and China	R, DB	On-pump heart surgery	258	72.2	58.5	30.6	NR	NR	Partial	84.4	99	LL, 300 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated cu	Iff AKIN

RIPC for AKI After Cardiac Surgery

Study	Country	Design	Surgical procedure	No. of patients	Mean age (years)	Male (%)	DM (%)	Baseline eGFR (ml/min)	LVEF <30% (%)	Propofol used	Volatile anesthetics	Cross- clamp time (min)	Protocols of RIPC	Control	Definition: of AKI
Pinaud et al. (23)	France	R, SB	On pump AVR with or without CABG	99	73.4	51.5	14.1	92.4	0	All	100	57	UL, 200 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated c	uff AKIN
Hu et al. (21)	China	R, DB	On-pump valvular heart surgery	201	47.1	37.8	0	NR	NR	All	100	59	LL, 600 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated c	uff AKIN
Nouraei et al. (22)	Iran	R, DB	On-pump CABG without valvular surgery	99	60.3	70.7	46.5	NR	0	Partial	38.5	41	LL, SBP $+$ 20 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated c	uff AKIN
Song et al. (26)	South Korea	R, DB	On-pump AVR due to aortic stenosis	72	66.5	50.0	0	NR	0	None	100	59	UL, 300 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated c	uff AKIN
Kim et al. (25)	South Korea	R, DB	On-pump heart surgery	160	62.3	53.1	0	NR	0	All	0	147	UL, 200 mmHg, 5 min $\times$ 4, 24 $\sim$ 48 h before surgery	Uninflated c	uff AKIN
Bagheri et al. (27)	Iran	R, DB	On-pump CABG without valvular surgery	180	63.6	57.6	35.0	NR	8	Partial	8.5	29	UL, 200 mmHg, 5 min $\times$ 3, after anesthesia induction and before CPB	Uninflated c	uff AKIN
Gasparovic et al. (28)	USA	R, DB	On-pump CABG without valvular surgery	66	62.0	82.0	36.0	NR	0	None	100	56	UL, 200 mmHg, 5 min $\times$ 3, after an esthesia induction and before CPB	Uninflated c	uff RIFLE
Wang et al. (29)	China	R, DB	Off-pump CABG without valvular surgery	65	60.5	73.5	NR	NR	0	None	100	NA	UL, SBP + 40 mmHg, 5 mir $\times$ 4, after anesthesia induction and before CPB	n Uninflated c	uff AKIN
Zhou et al. (35)	China	R, DB	Open TAAR under CPB with or without CABG	130	46.6	54	6.2	NR	NR	Partial	32.3	96	UL, SBP + 50 mmHg, 5 mir $\times$ 4, after anesthesia induction and before CPB	n Uninflated c	uff KDIGO
Stokfisz et al. (31)	Poland	R, DB	Off-pump CABG without valvular surgery	28	66.0	65.5	46.5	NR	NR	All	0	NA	UL, 200 mmHg, 5 min $\times$ 3, after an esthesia induction and before CPB	Uninflated c	uff KDIGO

DM, diabetes mellitus; eGFR, estimated glomerular filtrating rate; LVEF, left ventricular ejection fraction; RIPC, remote ischemic preconditioning; AKI, acute kidney injury; UK, United Kingdom; USA, United States of America; R, randomized; DB, double-blinded; SB, single-blinded; AVR, aortic valvular replacement; TAAR, total aortic arch replacement; CKD, chronic kidney disease; CABG, coronary artery bypass graft; NR, not reported; NA, not applicable; UL, upper limb; LL, lower limb; SBP, systolic blood pressure; CPB, cardiopulmonary bypass; AKIN, Acute Kidney Injury Network; RIFLE, Risk, Injury, Failure, Loss, End-Stage Kidney Disease; KDIGO, Kidney Disease: Improving Global Outcomes.

inflation to a pressure of 200 mmHg or at least a pressure that was 20 mmHg higher than the systolic arterial pressure), followed by 5–10 min reperfusion (with the cuff deflated). Uninflated cuffs were used on patients in the control group. Most of the studies used the AKIN definition of AKI, while the RIFLE (13, 19, 28) and the KDIGO (18, 20, 30, 31) definitions were used in three and four studies, respectively.

#### **Data Quality**

**Table 2** shows the details of study quality evaluation. Most of the included RCTs were double blinded except for four studies, which were single blinded (10, 12, 17, 23). Methods of random sequence generation were reported in 19 studies, and information of allocation concealment was reported in 14 studies. The overall quality score varied between 4 and 7, which suggested a generally good study quality.

#### **Meta-Analysis Results**

Moderate heterogeneity was detected (*p* for Cochrane's *Q* test = 0.03,  $I^2 = 40\%$ ) among the included RCTs. Pooled results with a random-effect model showed that RIPC significantly reduced the incidence of AKI after cardiac surgery compared with the control (OR: 0.76, 95% CI: 0.61–0.94, p = 0.01; **Figure 2**). Stratified analyses indicated that the effect of RIPC on postoperative AKI was not significantly different in studies with simple, complex, or mixed procedures (*p* for subgroup difference = 0.88, **Figure 3**).

Results of the univariate meta-regression analyses suggested that study sample size, age, gender, diabetic status, volatile anesthetic use, and cross-clamp time in CPB did not significantly affect the outcome (p all >0.05; Table 3). Moreover, sensitivity analyses confirmed a potential preventative efficacy of RIPC on postoperative AKI in patients with normal renal function at baseline (21 studies, OR: 0.75, 95% CI: 0.60–0.93, p = 0.01), in studies with on-pump surgery only (20 studies, OR: 0.78, 95% CI: 0.64–0.95, p = 0.01), and in studies with acute RIPC (21) studies, OR: 0.78, 95% CI: 0.63–0.97, *p* = 0.03; **Table 4**). Further subgroup analyses indicated that study characteristics, including study design, country, age, gender, diabetic status, surgery type, use of propofol or volatile anesthetics, cross-clamp time, RIPC protocol, definition of AKI, or sample size of the RCT did not significantly affect AKI outcome (p for subgroup difference all >0.10; Table 4). Stratified analysis according to the severity of AKI showed that RIPC significantly reduced the risk of mildto-moderate AKI that did not require RRT (OR: 0.76, 95% CI: 0.60–0.96, p = 0.02) but did not significantly reduce the risk of severe AKI that required RRT in patients after cardiac surgery (OR: 0.73, 95% CI: 0.50–1.07, *p* = 0.11; **Figure 4**).

#### **Publication Bias**

The funnel plots were symmetrical, suggesting low-risk of publication bias (**Figure 5**). Egger's regression tests showed similar results (p = 0.28).

#### **TABLE 2** | Details of study quality evaluation using the Cochrane's risk-of-bias tool.

Study	Random sequence generation	Allocation concealment	Blinding of participants	Blinding of outcome assessment	Incomplete outcome data addressed	Selective reporting	Other sources of bias	Total
/enugopal et sl. (10)	Low	Unclear	Low	High	Low	Low	High	4
Choi et al. (11)	Low	Unclear	Low	Low	Low	Low	Low	6
Zimmerman et al. (12)	Low	Low	Low	High	Low	Low	Unclear	5
Young et al. (13)	Low	Low	Low	Low	Low	Low	Low	7
Meybohm et al. (14)	Unclear	Low	Low	Low	Low	Low	Unclear	5
Nang et al. (15)	Low	Unclear	Low	Low	Low	Low	High	5
Candilio et al. (16)	Low	Low	Low	Low	Low	Low	Unclear	6
Zarbock et al. (20)	Low	Low	Low	Low	Low	Low	Low	7
Gallagher et al. (17)	Unclear	Unclear	Low	High	Low	Low	Low	4
Vleybohm et al. (19)	Low	Low	Low	Low	Low	Low	Low	7
Hausenloy et al. (18)	Low	Low	Low	Low	Low	Low	Low	7
Walsh et al. (24)	Low	Low	Low	Low	Low	Low	Unclear	6
Pinaud et al. (23)	Low	Unclear	Low	High	Low	Low	Unclear	4
Hu et al. (7)	Unclear	Unclear	Low	Low	Low	Low	Unclear	4
Nouraei et al. (22)	Low	Low	Low	Low	Low	Low	Unclear	6
Song et al. (26)	Low	Low	Low	Low	Low	Low	Unclear	6
Kim et al. (25)	Low	Low	Low	Low	Low	Low	Unclear	6
Bagheri et al. (27)	Low	Low	Low	Low	Low	Low	Unclear	6
Gasparovic et al. (28)	Low	Low	Low	Low	Low	Low	Unclear	6
Vang et al. (29)	Low	Unclear	Low	Low	Low	Low	Unclear	5
Zhou et al. (30)	Low	Low	Low	Low	Low	Low	Low	7
Stokfisz et al. (31)	Low	Unclear	Low	Low	Low	Low	Unclear	5

Stokfi Zimm Venuç	v or Subgroup sz 2020	Events 4	Total	Events	Total	Waight	N/ D 0.50/ 01	IV Developed AF0/ AL
Zimm Venug		4			10101	weight	IV, Random, 95% C	IV, Random, 95% CI
Venug			14	13	14	0.8%	0.03 [0.00, 0.32]	
	erman 2011	12	59	28	59	4.6%	0.28 [0.13, 0.64]	
	gopal 2010	4	38	10	40	2.4%	0.35 [0.10, 1.24]	
Candi	lio 2015	9	89	19	89	4.3%	0.41 [0.18, 0.98]	
Kim 2	017	24	80	38	80	6.1%	0.47 [0.25, 0.91]	
Zhou	2019	36	65	47	65	5.3%	0.48 [0.23, 0.99]	
Noura	ei 2016	7	50	12	49	3.3%	0.50 [0.18, 1.41]	+
Zarbo	ck 2015	45	120	63	120	7.8%	0.54 [0.32, 0.91]	-
Wang	2014	7	15	9	16	1.9%	0.68 [0.17, 2.80]	
Hu 20	16	69	101	71	100	6.7%	0.88 [0.48, 1.61]	-
Hause	enloy 2015	287	794	293	772	12.7%	0.93 [0.75, 1.14]	+
Baghe	eri 2018	38	87	41	90	6.8%	0.93 [0.51, 1.68]	+
Gallag	gher 2015	12	43	12	43	3.8%	1.00 [0.39, 2.57]	_ <del></del>
Wang	2019	4	32	4	33	1.8%	1.04 [0.24, 4.55]	
Pinau	d 2016	13	50	12	49	4.0%	1.08 [0.44, 2.68]	_ <del></del>
Walsh	2016	27	128	25	130	6.6%	1.12 [0.61, 2.06]	+
Meybo	ohm 2013	9	90	8	90	3.4%	1.14 [0.42, 3.10]	_ <del></del> _
Young	2012	7	48	6	48	2.7%	1.20 [0.37, 3.86]	_ <del></del> _
Meybo	ohm 2015	42	692	35	693	8.6%	1.21 [0.77, 1.93]	+
Choi 2	2011	14	38	12	38	3.7%	1.26 [0.49, 3.27]	
Gaspa	arovic 2018	4	33	3	33	1.6%	1.38 [0.28, 6.71]	<del></del> _
Song	2017	3	36	2	36	1.2%	1.55 [0.24, 9.85]	<del></del>
Total	(95% CI)		2702		2687	100.0%	0.76 [0.61, 0.94]	•
Total	events	677		763				
	ogeneity: Tau² = or overall effect:				(P = 0	.03); l² = 4	.0%	0.005 0.1 1 10 200 Favours RIPC Favours control

# DISCUSSION

In this updated meta-analysis, we pooled the results of 22 RCTs that involved 5,389 patients who received cardiac surgery and found that RIPC significantly reduced the incidence of postoperative AKI in these patients. Sensitivity analyses showed consistent results in studies with on-pump surgery only, in studies of patients with normal renal function at baseline, and in studies evaluating the acute efficacy of RIPC. In addition, results of meta-regression and subgroup analyses indicated that characteristics, including study design, country, age, gender, diabetic status, surgery type, use of propofol or volatile anesthetics, cross-clamp time, RIPC protocol, definition of AKI, and sample size of the RCT did not significantly impact the efficacy of RIPC on postoperative AKI. Moreover, stratified analysis showed that RIPC significantly reduced the risk of mild-to-moderate AKI that did not require RRT but did not significantly reduce the risk of severe AKI that required RRT in patients after cardiac surgery. To sum up, these results suggest that RIPC is effective for preventing AKI after cardiac surgery, which seems to be mainly driven by the reduced mild-tomoderate AKI events that did not require RRT. Further efforts are needed to determine the influences of patient characteristics, procedure, perioperative drugs, and RIPC protocol on AKI outcome.

Some meta-analyses have been previously published evaluating the use of RIPC for preventing AKI after cardiac surgery (32-35). Compared with these studies, our updated meta-analysis has the following strengths. First, we included the most recent RCTs with more than 5,000 patients, which is a much larger sample population than was included in the previous meta-analyses. The relatively large number of available RCTs and sample size of this meta-analysis may overcome the potential statistical inadequacy of previously published individual RCTs and meta-analyses. Secondly, some RCTs (41, 42) evaluating the efficacy of combined remote ischemic pre- and post-conditioning, rather than RIPC alone, on AKI were included in previous meta-analyses (34, 35), which may have confounded the results. In our meta-analysis, only studies that compared RIPC and control on AKI after cardiac surgery were included, which therefore reflects the potential benefits of RIPC only. Finally, the relative larger number of available RCTs enables us to perform comprehensive meta-regression and subgroup analyses of the influences of study characteristics on AKI outcome. This is important because these factors were rarely considered in previous meta-analyses, probably due to the limited RCTs available. In this study, we synthesized the current evidence from RCTs and showed that RIPC is effective for preventing AKI after cardiac surgery. Besides the potential preventative efficacy on AKI, previous studies have shown that RIPC is also effective to attenuate perioperative myocardial

	RIPC		Contr			Odds Ratio	Odds Ratio
Study or Subgroup	Events	Total	Events	Total	Weight	IV, Random, 95% Cl	IV, Random, 95% Cl
1.4.1 Simple							
Stokfisz 2020	4	14	13	14	0.8%	0.03 [0.00, 0.32]	
Venugopal 2010	3	35	7	32	1.8%	0.33 [0.08, 1.43]	
Nouraei 2016	7	50	12	49	3.2%	0.50 [0.18, 1.41]	+
Bagheri 2018	38	87	41	90	6.8%	0.93 [0.51, 1.68]	-+
Wang 2019	4	32	4	33	1.7%	1.04 [0.24, 4.55]	
Gasparovic 2018	4	33	3	33	1.6%	1.38 [0.28, 6.71]	_ <del>_</del>
Song 2017	3	36	2	36	1.2%	1.55 [0.24, 9.85]	
Subtotal (95% CI)		287		287	17.0%	0.64 [0.33, 1.25]	$\bullet$
Total events	63		82				
Heterogeneity: Tau <sup>2</sup> =		= 10.73	3. df = 6 (	P = 0.1	0); $ ^2 = 44$	%	
Test for overall effect:					-,,		
1.4.2 Complex							
Zhou 2019	36	65	47	65	5.3%	0.48 [0.23, 0.99]	
Venugopal 2010	1	3		8	0.5%	0.83 [0.05, 13.63]	
Young 2012	7	48	6	48	2.6%	1.20 [0.37, 3.86]	_ <del></del>
Choi 2011	14	38	12	38	3.6%	1.26 [0.49, 3.27]	_ <b>_</b>
Subtotal (95% CI)	14	154	12	159	12.1%	0.78 [0.45, 1.35]	
Total events	58	101	68	100	12.170	0110 [0110, 1100]	
Heterogeneity: Tau <sup>2</sup> =		= 3.26		P = 0 3F	5) $l^2 = 8\%$		
Test for overall effect:	Z = 0.89 (F	P = 0.3	7)				
1.4.3 Mixed							
Zimmerman 2011	12	59	28	59	4.6%	0 00 10 10 0 0 11	
	12	00	20	00	4.070	0.28 [0.13, 0.64]	-
Candilio 2015	9	89	20 19	89	4.3%	0.28 [0.13, 0.64] 0.41 [0.18, 0.98]	
Candilio 2015	9	89	19	89	4.3%	0.41 [0.18, 0.98]	
Candilio 2015 Kim 2017	9 24	89 80	19 38	89 80	4.3% 6.1%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91]	
Candilio 2015 Kim 2017 Zarbock 2015	9 24 45	89 80 120	19 38 63	89 80 120	4.3% 6.1% 7.9%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014	9 24 45 7	89 80 120 15	19 38 63 9	89 80 120 16	4.3% 6.1% 7.9% 1.9%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016	9 24 45 7 69	89 80 120 15 101	19 38 63 9 71	89 80 120 16 100	4.3% 6.1% 7.9% 1.9% 6.7%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015	9 24 45 7 69 287	89 80 120 15 101 794	19 38 63 9 71 293	89 80 120 16 100 772	4.3% 6.1% 7.9% 1.9% 6.7% 13.2%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015	9 24 45 7 69 287 12	89 80 120 15 101 794 43	19 38 63 9 71 293 12	89 80 120 16 100 772 43	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016	9 24 45 7 69 287 12 13	89 80 120 15 101 794 43 50	19 38 63 9 71 293 12 12	89 80 120 16 100 772 43 49	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013	9 24 45 7 69 287 12 13 27	89 80 120 15 101 794 43 50 128	19 38 63 9 71 293 12 12 25	89 80 120 16 100 772 43 49 130	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2015	9 24 45 7 69 287 12 13 27 9	89 80 120 15 101 794 43 50 128 90	19 38 63 9 71 293 12 12 12 25 8	89 80 120 16 100 772 43 49 130 90	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2015 <b>Subtotal (95% CI)</b>	9 24 45 7 69 287 12 13 27 9	89 80 120 15 101 794 43 50 128 90 692	19 38 63 9 71 293 12 12 12 25 8	89 80 120 16 100 772 43 49 130 90 693	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4%	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2015 <b>Subtotal (95% CI)</b> Total events	9 24 45 7 69 287 12 13 27 9 42 556	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b>	19 38 63 9 71 293 12 12 25 8 35 613	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b>	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b>	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2015 <b>Subtotal (95% CI)</b>	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup>	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33	19 38 63 9 71 293 12 25 8 35 613 3, df = 11	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b>	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b>	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2015 <b>Subtotal (95% CI)</b> Total events Heterogeneity: Tau <sup>2</sup> =	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup>	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33	19 38 63 9 71 293 12 25 8 35 613 3, df = 11	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b> (P = 0	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b>	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99]	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2013 <b>Subtotal (95% CI)</b> Total events Heterogeneity: Tau <sup>2</sup> = Test for overall effect:	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup>	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33 P = 0.04	19 38 63 9 71 293 12 25 8 35 613 3, df = 11	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b> (P = 0	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b> .04); I <sup>2</sup> = 4	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99] 6%	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2013 <b>Subtotal (95% CI)</b> Total events Heterogeneity: Tau <sup>2</sup> = Test for overall effect: <b>Total (95% CI)</b> Total events	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup> Z = 2.05 (F	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33 P = 0.04 <b>2702</b>	19 38 63 9 71 293 12 25 8 35 613 3, df = 11 4) 763	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b> (P = 0 2687	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b> .04); I <sup>2</sup> = 4	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99] 6%	
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2013 Subtotal (95% CI) Total events Heterogeneity: Tau <sup>2</sup> = Test for overall effect: Total (95% CI) Total events Heterogeneity: Tau <sup>2</sup> =	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup> Z = 2.05 (F 677 0.07; Chi <sup>2</sup>	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33 P = 0.04 <b>2702</b> = 34.75	19 38 63 9 71 293 12 25 8 35 613 3, df = 11 4) 763 8, df = 22	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b> (P = 0 2687	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b> .04); I <sup>2</sup> = 4	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99] 6%	• • • • • • • • • • • • • • • • • •
Candilio 2015 Kim 2017 Zarbock 2015 Wang 2014 Hu 2016 Hausenloy 2015 Gallagher 2015 Pinaud 2016 Walsh 2016 Meybohm 2013 Meybohm 2013 <b>Subtotal (95% CI)</b> Total events Heterogeneity: Tau <sup>2</sup> = Test for overall effect: <b>Total (95% CI)</b> Total events	9 24 45 7 69 287 12 13 27 9 42 556 0.07; Chi <sup>2</sup> Z = 2.05 (F 677 0.07; Chi <sup>2</sup> Z = 2.57 (F	89 80 120 15 101 794 43 50 128 90 692 <b>2261</b> = 20.33 P = 0.0 <b>2702</b> = 34.73 P = 0.0	19 38 63 9 71 293 12 25 8 35 613 3, df = 11 4) 763 8, df = 22 1)	89 80 120 16 100 772 43 49 130 90 693 <b>2241</b> (P = 0 <b>2687</b> (P = 0	4.3% 6.1% 7.9% 1.9% 6.7% 13.2% 3.7% 3.9% 6.6% 3.4% 8.7% <b>70.9%</b> .04); l <sup>2</sup> = 4 <b>100.0%</b>	0.41 [0.18, 0.98] 0.47 [0.25, 0.91] 0.54 [0.32, 0.91] 0.68 [0.17, 2.80] 0.88 [0.48, 1.61] 0.93 [0.75, 1.14] 1.00 [0.39, 2.57] 1.08 [0.44, 2.68] 1.12 [0.61, 2.06] 1.14 [0.42, 3.10] 1.21 [0.77, 1.93] 0.77 [0.60, 0.99] 6% 0.76 [0.62, 0.94]	0.005 0.1 1 10 200 Favours RIPC Favours control

injury (43) and reduce the risk of new onset atrial fibrillation (44) in patients undergoing cardiac surgery and may reduce mortality in patients receiving volatile inhalational agent anesthesia (34). Taken together, these findings support the benefits of RIPC use for patients undergoing cardiac surgeries.

Moderate heterogeneity was detected among the included studies in our meta-analysis. Although we aimed to evaluate the potential contribution of study characteristics on AKI outcome, results of meta-regression and subgroup analyses did not reveal any significant relationship between these factors and the efficacy of RIPC on postoperative AKI. It has been suggested by previous meta-analyses that patient age, complexity of the surgical procedure, and use of propofol may significantly modify the efficacy of RIPC on postoperative AKI (32, 35). However, by including more eligible RCTs, we found that these factors played no significant role on the outcome of AKI following RIPC. In fact, some recently published experimental studies suggest that the potential interaction between these factors, such as propofol, with organ protective efficacy of RIPC may be more complicated than expected. A recent study in a rat

#### TABLE 3 | Results of univariate meta-regression analysis.

Study characteristics	OR for the incidence of AKI after cardiac surgery								
	Coefficient	95% CI	p						
Number of subjects	0.13	-0.07 to 0.33	0.19						
Mean age (years)	0.09	-0.07 to 0.25	0.27						
Male (%)	0.02	-0.10 to 0.14	0.72						
DM (%)	-0.02	-1.12 to 1.08	0.97						
Volatile anesthetic use (%)	-0.12	-0.53 to 0.20	0.55						
Cross-clamp time (min)	-0.17	-0.82 to 0.48	0.59						

model found that the timing of propofol administration could affect RIPC-induced cardioprotection (45), and that different doses of propofol used in the studies may have also affected the results. Future studies are needed to determine the influences of patient characteristics, procedure, perioperative drugs, and RIPC protocol on the efficacy of RIPC on AKI prevention.

The potential mechanisms underlying the preventative role of RIPC on AKI after cardiac surgery are likely multifactorial (8, 9). Generally, RIPC is considered to mediate organ protection via regulating both the sympathetic and parasympathetic components of the autonomic nervous system (46–48). Ischemia

#### TABLE 4 | Results of sensitivity and subgroup analyses.

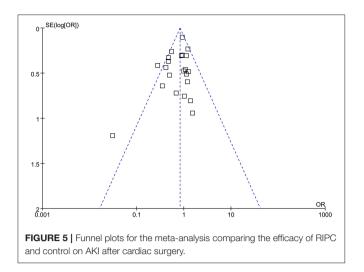
Study characteristics	Datasets number	OR (95% CI)	l <sup>2</sup>	p for subgroup effect	<i>p</i> for subgroup difference
Baseline renal function					
Only patients without CKD	21	0.75 [0.60, 0.93]	43%	0.01	-
Surgery characteristics					
Only studies of on-pump surgery	20	0.78 [0.64, 0.95]	31%	0.01	-
Timing of RIPC					
Only studies with acute RIPC	21	0.78 [0.63, 0.97]	38%	0.03	-
Study origin					
Asian	9	0.73 [0.55, 0.96]	0%	0.02	
Non-Asian	12	0.72 [0.51, 1.00]	59%	0.05	0.94
Study design					
DB	18	0.80 [0.64, 0.99]	35%	0.04	
SB	4	0.58 [0.28, 1.19]	55%	0.14	0.41
Surgery type					
CABG $\pm$ valvular surgery	10	0.66 [0.46, 0.96]	49%	0.03	
Valvular surgery $\pm$ CABG	5	0.99 [0.66, 1.50]	0%	0.96	
Any cardiac surgery	7	0.78 [0.55, 1.12]	50%	0.18	0.36
Propofol use					
All patients	10	0.72 [0.50, 1.02]	63%	0.06	
Partial patients	5	0.71 [0.47, 1.08]	29%	0.11	
None patients	5	0.75 [0.50, 1.12]	0%	0.16	0.99
Volatile anesthetics use					
All patients	10	0.72 [0.51, 1.02]	28%	0.06	
Partial patients	6	0.85 [0.68, 1.06]	12%	0.15	
None patients	4	0.61 [0.25, 1.49]	78%	0.28	0.62
Cross-clamp time (min)					
≤70	10	0.79 [0.59, 1.05]	0%	0.10	
>70	10	0.77 [0.58, 1.03]	56%	0.07	0.92
RIPC protocol					
Upper limb	16	0.76 [0.59, 0.98]	41%	0.03	
Lower limb	6	0.73 [0.46, 1.16]	46%	0.19	0.87
Definition of AKI					
AKIN	15	0.74 [0.58, 0.96]	19%	0.02	
RIFLE	3	1.22 [0.81, 1.85]	0%	0.34	
KDIGO	4	0.53 [0.28, 1.01]	78%	0.05	0.15
Sample size of RCT					
≤100	11	0.81 [0.53, 1.24]	22%	0.34	
>100	11	0.73 [0.57, 0.95]	55%	0.22	0.68

CKD, chronic kidney disease; RIPC, remote ischemic preconditioning; AKI, acute kidney injury; DB, double-blinded; SB, single-blinded; CABG, coronary artery bypass graft; UL, upper limb; LL, lower limb; AKIN, Acute Kidney Injury Network; RIFLE, Acute Dialysis Quality Initiative; Kidney Disease Improving Global Outcomes.

Study or Subgroup         Events         Total         Weight         IV. Random. 95% CI         IV. Random. 95% CI           11.21 Mild to moderate AKI without RRT         Stokfisz 2020         3         14         12         14         0.9%         0.05 [0.01, 0.33]           Zimmerman 2011         12         59         28         59         4.0%         0.28 [0.13, 0.64]           Venugopal 2010         4         38         0         40         2.1%         0.35 [0.10, 1.24]           Candilio 2015         8         89         16         89         3.5%         0.44 [0.22, 0.87]           Candilio 2016         7         50         12         49         2.9%         0.50 [0.18, 1.41]           Thouraei 2016         7         50         12         49         2.9%         0.50 [0.14, 1.41]           Zhou 2019         33         65         41         65         4.9%         0.60 [0.30, 1.22]           Zarbock 2015         38         120         44         120         6.7%         0.80 [0.47, 1.37]           Wang 2014         7         15         8         16         1.7%         0.38 [0.27, 1.59]           Baybeni 2018         33         33         1.2%         1.00 [0.		RIPC		Contr			Odds Ratio	Odds Ratio
Stokfisz 2020       3       14       12       14       0.9%       0.05 [0.01, 0.33]         Zimmerman 2011       12       59       28       59       4.0%       0.28 [0.13, 0.64]         Venugopal 2010       4       38       10       40       2.1%       0.25 [0.10, 1.24]         Kim 2017       18       80       32       80       5.0%       0.44 [0.22, 0.87]         Candilio 2015       8       89       16       89       3.5%       0.45 [0.18, 1.41]         Nouraei 2016       7       50       12       49       2.9%       0.50 [0.18, 1.41]         Zhou 2019       3       65       41       65       4.9%       0.60 [0.30, 1.22]         Gallagher 2015       10       43       12       43       0.5%       0.88 [0.21, 3.59]         Bagheri 2018       38       87       41       90       6.0%       0.93 [0.51, 1.68]         Hausenloy 2015       268       794       270       772       1.5%       0.95 [0.77, 1.17]         Gasparovic 2018       3       33       3       1.2%       1.26 [0.49, 3.27]       1.50         Pinaud 2016       4       5       48       2.1%       1.23 [0.35, 4.33]<					Total	Weight	IV, Random, 95% C	IV, Random, 95% Cl
Zimmerman 2011 12 59 28 59 4.0% 0.28 [0.13, 0.64] Venugopal 2010 4 38 10 40 2.1% 0.35 [0.10, 1.24] Kim 2017 18 80 32 80 5.0% 0.44 [0.22, 0.87] Candillo 2015 8 89 16 89 3.5% 0.45 [0.18, 1.11] Nouraei 2016 7 50 12 49 2.9% 0.50 [0.18, 1.41] Zarbock 2015 10 43 12 43 3.1% 0.78 [0.30, 2.07] Zarbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Walsh 2016 19 128 23 130 5.3% 0.98 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.98 [0.77, 1.17] Gasparovic 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.95 [0.77, 1.17] Gasparovic 2018 3 33 3 33 1.2% 1.00 [0.19, 5.36] Wang 2014 7 15 8 16 1.7% 0.88 [0.24, 3.50] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 7 120 19 120 3.5% 0.33 [0.01, 0.83] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); I <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candillo 2015 1 89 3 89 0.7% 0.33 [0.03, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.11, 0.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 214 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfiz 2020 1 144 14 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 68 02 3.3% 1.00 [0.3, 3.42] Young 2012 1 48 148 0.5% 1.00 [0.06, 14.64] Gasparovic 2018 1 33 0 33 0.4% 3.29 [0.12, 78.70] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 0.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112.45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 1.2 (P = 0.41); I <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)								
Venugopal 2010 4 38 10 40 2.1% 0.35 [0.10, 1.24] Kim 2017 18 80 32 80 5.0% 0.44 [0.22, 0.87] Candillo 2015 8 89 16 89 3.5% 0.45 [0.18, 1.11] Nouraei 2016 7 50 12 49 2.9% 0.50 [0.18, 1.41] Zhou 2019 33 65 41 65 4.9% 0.60 [0.30, 1.22] Gallagher 2015 10 43 12 43 3.1% 0.78 [0.30, 2.07] Zarbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Wang 2014 7 15 8 16 1.7% 0.88 [0.21, 3.59] Bagheri 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.85 [0.77, 1.77] Gasparovic 2018 3 33 3 33 1.2% 1.00 [0.19, 5.36] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 ( $P = 0.02$ ); $P = 42%$ Test for overall effect: Z = 2.29 ( $P = 0.02$ ) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candillo 2015 1 89 3 89 0.7% 0.33 [0.01, 3, 0.8] Zhou 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Heusenloy 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Heusenloy 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Heusenloy 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112, 45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12 ( $P = 0.41$ ); $P = 3\%$ Test for overall effect: Z = 1.62 ( $P = 0.11$ )								
Kim 2017       18       80       32       80       5.0%       0.44       [0.22, 0.87]         Candilio 2015       8       89       16       89       3.5%       0.44       [0.22, 0.87]         Nouraei 2016       7       50       12       49       2.9%       0.50       [0.18, 1.41]         Zhou 2019       33       65       41       65       4.9%       0.60       [0.30, 1.22]         Gallagher 2015       10       43       12       43       3.1%       0.76       [0.30, 2.07]         Zarbock 2015       38       87       41       90       6.0%       0.80       [0.47, 1.37]         Wang 2014       7       15       8       16       1.7%       0.88       [0.77, 1.17]         Gasparovic 2018       3       33       33       1.2%       1.00       [0.19, 5.36]         Wang 2019       4       32       4       33       1.6%       1.44       [0.42, 3.10]         Young 2012       6       48       5.48       2.1%       1.23       [0.35, 4.33]         Choi 2011       14       38       123       1.55       [0.89, 2.72]         Subtotal (95% Cl)       2.565	Zimmerman 2011		59	28	59	4.0%	0.28 [0.13, 0.64]	
Candillo 2015 8 89 16 89 3.5% 0.45 [0.18, 1.11] Nouraei 2016 7 50 12 49 2.9% 0.50 [0.18, 1.41] Candillo 2019 33 65 41 65 4.9% 0.60 [0.30, 1.22] Gallagher 2015 10 43 12 43 3.1% 0.78 [0.30, 2.07] Carbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Wang 2014 7 15 8 16 1.7% 0.88 [0.21, 3.59] Bagheri 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.95 [0.77, 1.77] Gasparovic 2018 3 33 3 33 1.2% 1.00 [0.19, 5.36] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2019 4 32 4 33 1.6% 1.46 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candillo 2015 1 89 3 89 0.7% 0.33 [0.01, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Young 2012 1 148 148 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 448 148 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 448 148 0.5% 1.00 [0.06, 17.75] Walsh 2016 8 128 2 130 1.4% 4.27 [0.88, 20.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112, 45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Venugopal 2010	4	38		40	2.1%	0.35 [0.10, 1.24]	
Nouraei 2016 7 50 12 49 2.9% 0.50 [0.18, 1.41] Zhou 2019 33 65 41 65 4.9% 0.60 [0.30, 1.22] Zarbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Wang 2014 7 15 8 16 1.7% 0.88 [0.21, 3.59] Bagheri 2018 38 67 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.95 [0.77, 1.17] Gasparovic 2018 3 33 3 33 1.2% Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Heterogeneity: Tau <sup>2</sup> = 0.10; Ch <sup>2</sup> = 3.288, df = 19 (P = 0.02); P = 42% Test for overall effect: Z = 2.29 (P = 0.02) <b>1.12.2 Severe AKI with RRT</b> Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.03, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 3.19] Zarbock 2015 7 120 19 2.0 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 3.19] Zarbock 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 16, 46] Gasparovic 2018 1 33 0 0.3 0.31 0.4% 3.09 [0.12, 78.70] Walsh 2016 8 122 2 1.30 1.4% 4.27 [0.88, 0.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112.45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Ch <sup>2</sup> = 12.42, df = 12 (P = 0.41); P = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Kim 2017	18	80	32	80	5.0%	0.44 [0.22, 0.87]	
Zhou 2019 33 65 41 65 4.9% 0.60 [0.30, 1.22] Gallagher 2015 10 43 12 43 3.1% 0.78 [0.30, 2.07] Zarbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Wang 2014 7 15 8 16 1.7% 0.88 [0.21, 3.59] Bagheri 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.95 [0.77, 1.17] Gasparovic 2018 3 33 3 33 1.2% 1.00 [0.19, 5.36] Wang 2019 4 32 4 33 1.6% 1.04 [0.42, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.36, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.03, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.82] Wang 2014 0 15 1 18 9 3 .0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.18] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.18] Min 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.26, 17.87] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 20.50] Gallagher 2015 2 4 3 0 43 0.4% 5.24 [0.24, 112.45] Multicl (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Candilio 2015	8	89	16	89	3.5%	0.45 [0.18, 1.11]	
Gallagher 2015       10       43       12       43       3.1%       0.78       0.30, 2.07         Zarbock 2015       38       120       44       120       6.7%       0.80       [0.47, 1.37]         Walsh 2016       19       128       23       130       5.3%       0.81       [0.42, 1.57]         Wang 2014       7       15       8       16       1.7%       0.88       [0.21, 3.59]         Bagheri 2018       38       87       41       90       6.0%       0.93       [0.57, 1.17]         Gasparovic 2018       3       33       3       3.3       1.2%       1.00       [0.12, 4.65]         Wang 2019       4       32       4       33       1.6%       1.04       [0.24, 4.55]         Meybohm 2013       9       9       8       90       3.0%       1.14       [0.42, 3.63]         Choi 2011       14       38       12       38       3.2%       1.26       [0.49, 3.27]         Pinaud 2016       13       50       10       49       3.3%       1.37       [0.54, 3.50]         Meybohm 2015       12       6.92       21       6.93       6.4%       0.19       [0.01, 4.02]<	Nouraei 2016	7	50	12	49	2.9%	0.50 [0.18, 1.41]	
Zarbock 2015 38 120 44 120 6.7% 0.80 [0.47, 1.37] Walsh 2016 19 128 23 130 5.3% 0.81 [0.42, 1.57] Wang 2014 7 15 8 16 1.7% 0.86 [0.21, 3.59] Bagheri 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.95 [0.77, 1.17] Gasparovic 2018 3 33 3 3 1.2% 1.00 [0.19, 5.36] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.40] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.60] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); I <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) <b>1.12.2 Severe AKI with RRT</b> Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.03, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 19 794 23 772 5.7% 0.88 [0.43, 1.48] Stokfisz 2020 1 1 44 1 14 0.5% 1.00 [0.66, [1.7.5] Kim 2017 6 80 6 80 2.3% 1.00 [0.36, 13, 2.4] Young 2012 1 48 1 48 0.5% 1.00 [0.66, [1.7.5] Kim 2017 6 80 6 80 2.3% 1.00 [0.26, [1.7.87] Kim 2017 6 80 6 80 2.3% 1.00 [0.26, [1.7.87] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 20.50] Galagher 2015 2 4 3 0 43 0.4% 5.24 [0.24, 112.45] Subtotal (95% CI) 2171 2.152 2.1.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Zhou 2019	33	65	41	65	4.9%	0.60 [0.30, 1.22]	
Walsh 2016       19       128       23       130 $5.3\%$ 0.81       [0.42, 1.57]         Wang 2014       7       15       8       16       1.7%       0.88       [0.21, 3.59]         Bagheri 2018       38       87       41       90       6.0%       0.93       [0.51, 1.68]         Hausenloy 2015       268       794       270       772       11.5%       0.95       [0.77, 1.17]         Gasparovic 2018       3       33       3       3       1.01       [0.24, 3.56]         Wang 2019       4       32       4       33       1.6%       1.04       [0.24, 4.55]         Meybohm 2013       9       90       8       90       3.0%       1.41       [0.42, 3.27]         Young 2012       6       48       5       48       2.1%       1.23       [0.35, 4.33]         Choi 2011       14       38       1.26       [0.49, 3.27]       [0.43, 3.50]         Pinaud 2016       526       2551       78.3%       0.76       [0.60, 0.96]       [0.14, 4.02]         Candilio 2015       1       89       3       89       0.7%       0.33       [0.01, 4.02]       [0.43, 1.48]       [0.44]	Gallagher 2015	10	43	12	43	3.1%	0.78 [0.30, 2.07]	
Wang 2014       7       15       8       16       1.7%       0.88       [0.21, 3.59]         Bagheri 2018       38       87       41       90       6.0%       0.93       [0.51, 1.68]         Hausenloy 2015       268       794       270       772       11.5%       0.95       [0.77, 1.17]         Gasparovic 2018       3       33       3       33       1.2%       1.00       [0.19, 5.36]         Wang 2019       4       32       4       33       1.6%       1.04       [0.24, 4.55]         Meybohm 2013       9       90       8       90       3.0%       1.37       [0.54, 3.50]         Young 2012       6       48       5       48       2.1%       1.23       [0.35, 4.33]         Choi 2011       14       38       3.2%       1.26       [0.49, 3.27]       Pinaud 2016       1.35       [0.49, 3.27]         Vindagenetic       2565       2551       78.3%       0.76       [0.60, 0.96]       Interval         Total events       546       612       Pinaud 2016       0       50       2.35       0.33       [0.01, 4.02]       Pinaud 2016       50       93       89       0.7%       0.33	Zarbock 2015	38	120	44	120	6.7%	0.80 [0.47, 1.37]	
Bagheri 2018 38 87 41 90 6.0% 0.93 [0.51, 1.68] Hausenloy 2015 268 794 270 772 11.5% 0.55 [0.77, 1.17] Gasparovic 2018 3 33 3 3 1.2% 1.00 [0.19, 5.36] Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% CI) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); I <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2015 19 794 23 772 5.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.48 [0.43, 1.48] Stokfisz 2020 1 144 1 44 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.43, 1.48] Stokfisz 2020 1 144 1 44 0.5% 1.00 [0.06, 16, 7.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 16, 7.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 16, 7.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 16, 7.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 16, 7.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.06, 16, 7.75] Kim 2017 1 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 1.242, df = 12 (P = 0.41); I <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Walsh 2016	19	128	23	130	5.3%	0.81 [0.42, 1.57]	
Hausenloy 2015 268 794 270 772 11.5% 0.95 $[0.77, 1.17]$ Gasparovic 2018 3 33 3 3 1.2% 1.00 $[0.19, 5.36]$ Wang 2019 4 32 4 33 1.6% 1.04 $[0.24, 4.55]$ Waybohm 2013 9 90 8 90 3.0% 1.14 $[0.42, 3.10]$ Young 2012 6 48 5 48 2.1% 1.23 $[0.35, 4.33]$ Choi 2011 1 14 38 12 38 3.2% 1.26 $[0.49, 3.27]$ Pinaud 2016 13 50 10 49 3.3% 1.37 $[0.54, 3.50]$ Meybohm 2015 32 692 21 693 6.4% 1.55 $[0.89, 2.72]$ Subtotal (95% Cl) 2565 2551 78.3% 0.76 $[0.60, 0.96]$ Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); I <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 $[0.01, 4.02]$ Candilio 2015 1 89 3 89 0.7% 0.33 $[0.03, 3.19]$ Zarbock 2015 7 120 19 120 3.5% 0.33 $[0.13, 0.82]$ Wang 2014 0 15 1 16 0.4% 0.33 $[0.01, 4.03]$ Candilio 2015 19 794 23 772 5.7% 0.80 $[0.43, 1.48]$ Zhou 2019 3 65 6 65 1.7% 0.48 $[0.11, 1.99]$ Meybohm 2015 10 692 14 693 4.0% 0.71 $[0.31, 1.61]$ Hausenloy 2015 19 794 23 772 5.7% 0.80 $[0.43, 1.48]$ Stokfiz 2020 1 1 14 1 14 0.5% 1.00 $[0.06, 17.75]$ Kim 2017 6 80 6 80 2.3% 1.00 $[0.31, 3.24]$ Young 2012 1 48 1 48 0.5% 1.00 $[0.36, 13.24]$ Young 2012 1 48 1 48 0.5% 1.00 $[0.36, 13.24]$ Young 2012 1 48 1 48 0.4% 5.24 $[0.24, 112.45]$ Subtotal (95% Cl) 2171 2152 21.7% 0.73 $[0.50, 1.07]$ Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Wang 2014	7	15	8	16	1.7%	0.88 [0.21, 3.59]	
Gasparovic 2018       3       33       33       1.2%       1.00 [0.19, 5.36]         Wang 2019       4       32       4       33       1.6%       1.04 [0.24, 4.55]         Meybohm 2013       9       90       8       90       3.0%       1.14 [0.42, 4.55]         Meybohm 2012       6       4.8       5       4.8       2.1%       1.23 [0.35, 4.33]         Choi 2011       14       38       1.2       38       3.2%       1.26 [0.49, 3.27]         Pinaud 2016       13       50       10       49       3.3%       1.37 [0.54, 3.50]         Meybohm 2015       32       692       21       693       6.4%       1.55 [0.89, 2.72]         Subtotal (95% CI)       2565       2551       78.3%       0.76 [0.60, 0.96]         Total events       546       612         Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42%         Test for overall effect: Z = 2.29 (P = 0.02)         1.12.2 Severe AKI with RRT         Pinaud 2016       0       50       2       49       0.4%       0.19 [0.01, 4.02]       1.33 [0.01, 8.83]       1.48       1.48       0.5%       1.00 [0.6, 17.75]       Kim 2015       10       692       14       <	Bagheri 2018	38	87	41	90	6.0%	0.93 [0.51, 1.68]	+
Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% Cl) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42% Test for overall effect: $Z = 2.29$ (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfisz 2020 1 14 1 14 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.34, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.04, 1.75] Subtotal (95% Cl) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: $Z = 1.62$ (P = 0.11)	Hausenloy 2015	268	794	270	772	11.5%	0.95 [0.77, 1.17]	+
Wang 2019 4 32 4 33 1.6% 1.04 [0.24, 4.55] Meybohm 2013 9 90 8 90 3.0% 1.14 [0.42, 3.10] Young 2012 6 48 5 48 2.1% 1.23 [0.35, 4.33] Choi 2011 14 38 12 38 3.2% 1.26 [0.49, 3.27] Pinaud 2016 13 50 10 49 3.3% 1.37 [0.54, 3.50] Meybohm 2015 32 692 21 693 6.4% 1.55 [0.89, 2.72] Subtotal (95% Cl) 2565 2551 78.3% 0.76 [0.60, 0.96] Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42% Test for overall effect: $Z = 2.29$ (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfisz 2020 1 14 1 14 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.34, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.04, 1.75] Subtotal (95% Cl) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: $Z = 1.62$ (P = 0.11)	•							
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Meybohm 2015 32 692 21 693 6.4% 1.55 $[0.89, 2.72]$ Subtotal (95% Cl) 2565 2551 78.3% 0.76 $[0.60, 0.96]$ Total events 546 612 Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42% Test for overall effect: Z = 2.29 (P = 0.02) 1.12.2 Severe AKI with RRT Pinaud 2016 0 50 2 49 0.4% 0.19 $[0.01, 4.02]$ Candilio 2015 1 89 3 89 0.7% 0.33 $[0.13, 0.82]$ Wang 2014 0 15 1 16 0.4% 0.33 $[0.11, 1.99]$ Meybohm 2015 10 692 14 693 4.0% 0.71 $[0.31, 1.61]$ Hausenloy 2015 19 794 23 772 5.7% 0.80 $[0.43, 1.48]$ Stokfisz 2020 1 144 1 14 0.5% 1.00 $[0.06, 16.46]$ Gasparovic 2018 1 33 0 33 0.4% 3.09 $[0.12, 78.70]$ Walsh 2016 8 128 2 130 1.4% 4.27 $[0.89, 20.50]$ Gallagher 2015 2 43 0 43 0.4% 5.24 $[0.24, 112.45]$ Subtotal (95% Cl) 2171 2152 21.7% 0.73 $[0.50, 1.07]$ Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)		13						- <del>-</del> -
Subtotal (95% Cl)2565255178.3%0.76 [0.60, 0.96]Total events546612Heterogeneity: Tau <sup>2</sup> = 0.10; Chi <sup>2</sup> = 32.88, df = 19 (P = 0.02); l <sup>2</sup> = 42%Test for overall effect: Z = 2.29 (P = 0.02) <b>1.12.2 Severe AKI with RRT</b> Pinaud 20160502490.4%0.19 [0.01, 4.02]Candilio 20151893890.7%0.33 [0.03, 3.19]Zarbock 20157120191203.5%0.33 [0.13, 0.82]Wang 20140151160.4%0.33 [0.01, 8.83]Zhou 20193656651.7%0.48 [0.11, 1.99]Meybohm 201510692146934.0%0.71 [0.31, 1.61]Hausenloy 201519794237725.7%0.80 [0.43, 1.48]Stokfisz 20201141140.5%1.00 [0.06, 17.75]Kim 20176806802.3%1.00 [0.01, 3, 3.24]Young 20121481480.5%1.00 [0.06, 16.46]Gasparovic 20181330330.4%3.09 [0.12, 78.70]Walsh 2016812821301.4%4.27 [0.89, 20.50]Gallagher 20152430430.4%5.24 [0.24, 112.45]Subtotal (95% Cl)2171215221.7%0.73 [0.50, 1.07]Total events5978Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0		32		21	693			
Heterogeneity: $Tau^2 = 0.10$ ; $Chi^2 = 32.88$ , $df = 19$ (P = 0.02); $l^2 = 42\%$ Test for overall effect: Z = 2.29 (P = 0.02) <b>1.12.2 Severe AKI with RRT</b> Pinaud 2016 0 50 2 49 0.4% 0.19 [0.01, 4.02] Candilio 2015 1 89 3 89 0.7% 0.33 [0.03, 3.19] Zarbock 2015 7 120 19 120 3.5% 0.33 [0.13, 0.82] Wang 2014 0 15 1 16 0.4% 0.33 [0.01, 8.83] Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfisz 2020 1 14 1 14 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 16.46] Gasparovic 2018 1 33 0 33 0.4% 3.09 [0.12, 78.70] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 20.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112.45] <b>Subtotal</b> (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Subtotal (95% CI)		2565		2551	78.3%		•
Candilio 2015       1       89       3       89 $0.7\%$ $0.33$ [ $0.03$ , $3.19$ ]         Zarbock 2015       7       120       19       120 $3.5\%$ $0.33$ [ $0.13$ , $0.82$ ]         Wang 2014       0       15       1       16 $0.4\%$ $0.33$ [ $0.13$ , $0.82$ ]         Wang 2014       0       15       1       16 $0.4\%$ $0.33$ [ $0.01$ , $8.83$ ]         Zhou 2019       3       65       6       65 $1.7\%$ $0.48$ [ $0.11$ , $1.99$ ]         Meybohm 2015       10       692       14       693 $4.0\%$ $0.71$ [ $0.31$ , $1.61$ ]         Hausenloy 2015       19       794       23 $772$ $5.7\%$ $0.80$ [ $0.43$ , $1.48$ ]         Stokfisz 2020       1       14       1 $40.5\%$ $1.00$ [ $0.06$ , $17.75$ ]         Kim 2017       6       80       6 $80$ $2.3\%$ $1.00$ [ $0.06$ , $16.46$ ]         Gasparovic 2018       1       33 $0.33$ $0.4\%$ $3.09$ [ $0.12$ , $78.70$ ]         Walsh 2016       8       128 $2$ $130$ $1.4\%$ $4.27$ [ $0.89$ , $20.50$ ]         Gallagher 2015 $2$ $43$	1.12.2 Severe AKI wi	th RRT		,				
Zarbock 2015712019120 $3.5\%$ $0.33$ [0.13, 0.82]Wang 2014015116 $0.4\%$ $0.33$ [0.01, 8.83]Zhou 2019365665 $1.7\%$ $0.48$ [0.11, 1.99]Meybohm 20151069214693 $4.0\%$ $0.71$ [0.31, 1.61]Hausenloy 20151979423772 $5.7\%$ $0.80$ [0.43, 1.48]Stokfisz 2020114114 $0.5\%$ $1.00$ [0.06, 17.75]Kim 2017680680 $2.3\%$ $1.00$ [0.06, 16.46]Gasparovic 2018133033 $0.4\%$ $3.09$ [0.12, 78.70]Walsh 201681282130 $1.4\%$ $4.27$ [0.89, 20.50]Gallagher 20152430 $43$ $0.4\%$ $5.24$ [0.24, 112.45]Subtotal (95% CI)2171215221.7% $0.73$ [0.50, 1.07]Total events5978Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3%Test for overall effect: Z = 1.62 (P = 0.11)	Pinaud 2016	0	50	2	49	0.4%	0.19 [0.01, 4.02]	
Wang 2014015116 $0.4\%$ $0.33$ [0.01, 8.83]Zhou 2019365665 $1.7\%$ $0.48$ [0.11, 1.99]Meybohm 20151069214693 $4.0\%$ $0.71$ [0.31, 1.61]Hausenloy 20151979423772 $5.7\%$ $0.80$ [0.43, 1.48]Stokfisz 2020114114 $0.5\%$ $1.00$ [0.06, 17.75]Kim 2017680680 $2.3\%$ $1.00$ [0.06, 16.46]Gasparovic 2018133033 $0.4\%$ $3.09$ [0.12, 78.70]Walsh 201681282130 $1.4\%$ $4.27$ [0.89, 20.50]Gallagher 20152430 $43$ $0.4\%$ $5.24$ [0.24, 112.45]Subtotal (95% CI)2171215221.7% $0.73$ [0.50, 1.07]Total events5978Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3%Test for overall effect: Z = 1.62 (P = 0.11)	Candilio 2015	1	89	3	89	0.7%	0.33 [0.03, 3.19]	
Zhou 2019 3 65 6 65 1.7% 0.48 [0.11, 1.99] Meybohm 2015 10 692 14 693 4.0% 0.71 [0.31, 1.61] Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfisz 2020 1 14 1 14 0.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 16.46] Gasparovic 2018 1 33 0 33 0.4% 3.09 [0.12, 78.70] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 20.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112.45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: $Z = 1.62$ (P = 0.11)	Zarbock 2015	7	120	19	120	3.5%	0.33 [0.13, 0.82]	
Meybohm 2015       10       692       14       693       4.0% $0.71$ [0.31, 1.61]         Hausenloy 2015       19       794       23       772 $5.7\%$ $0.80$ [0.43, 1.48]         Stokfisz 2020       1       14       1       14 $0.5\%$ $1.00$ [0.06, 17.75]         Kim 2017       6       80       6       80 $2.3\%$ $1.00$ [0.06, 16.46]         Gasparovic 2018       1       33       0       33 $0.4\%$ $3.09$ [0.12, 78.70]         Walsh 2016       8       128       2       130 $1.4\%$ $4.27$ [0.89, 20.50]         Gallagher 2015       2       43       0 $43$ $0.4\%$ $5.24$ [0.24, 112.45]         Subtotal (95% CI)       2171       2152       21.7% $0.73$ [0.50, 1.07]         Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11) $I$ $I$ $I$ $I$	Wang 2014	0	15	1	16	0.4%	0.33 [0.01, 8.83]	
Hausenloy 2015 19 794 23 772 5.7% 0.80 [0.43, 1.48] Stokfisz 2020 1 14 1 40.5% 1.00 [0.06, 17.75] Kim 2017 6 80 6 80 2.3% 1.00 [0.31, 3.24] Young 2012 1 48 1 48 0.5% 1.00 [0.06, 16.46] Gasparovic 2018 1 33 0 33 0.4% 3.09 [0.12, 78.70] Walsh 2016 8 128 2 130 1.4% 4.27 [0.89, 20.50] Gallagher 2015 2 43 0 43 0.4% 5.24 [0.24, 112.45] Subtotal (95% CI) 2171 2152 21.7% 0.73 [0.50, 1.07] Total events 59 78 Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); I <sup>2</sup> = 3% Test for overall effect: $Z = 1.62$ (P = 0.11)	Zhou 2019	3	65	6	65	1.7%	0.48 [0.11, 1.99]	
Stokfisz 2020       1       14       1       14       0.5%       1.00 [0.06, 17.75]         Kim 2017       6       80       6       80       2.3%       1.00 [0.31, 3.24]         Young 2012       1       48       1       48       0.5%       1.00 [0.06, 16.46]         Gasparovic 2018       1       33       0       33       0.4%       3.09 [0.12, 78.70]         Walsh 2016       8       128       2       130       1.4%       4.27 [0.89, 20.50]         Gallagher 2015       2       43       0       43       0.4%       5.24 [0.24, 112.45]         Subtotal (95% CI)       2171       2152       21.7%       0.73 [0.50, 1.07]         Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)	Meybohm 2015	10	692	14	693	4.0%	0.71 [0.31, 1.61]	
Kim 2017       6       80       6       80       2.3%       1.00 [0.31, 3.24]         Young 2012       1       48       1       48       0.5%       1.00 [0.06, 16.46]         Gasparovic 2018       1       33       0       33       0.4%       3.09 [0.12, 78.70]         Walsh 2016       8       128       2       130       1.4%       4.27 [0.89, 20.50]         Gallagher 2015       2       43       0       43       0.4%       5.24 [0.24, 112.45]         Subtotal (95% CI)       2171       2152       21.7%       0.73 [0.50, 1.07]         Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)	Hausenloy 2015	19	794	23	772	5.7%	0.80 [0.43, 1.48]	
Young 2012       1       48       1       48       0.5%       1.00       [0.06, 16.46]         Gasparovic 2018       1       33       0       33       0.4%       3.09       [0.12, 78.70]         Walsh 2016       8       128       2       130       1.4%       4.27       [0.89, 20.50]         Gallagher 2015       2       43       0       43       0.4%       5.24       [0.24, 112.45]         Subtotal (95% CI)       2171       2152       21.7%       0.73       [0.50, 1.07]         Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)	Stokfisz 2020		14	1	14	0.5%	1.00 [0.06, 17.75]	
Gasparovic 2018       1       33       0       33       0.4%       3.09 $[0.12, 78.70]$ Walsh 2016       8       128       2       130       1.4%       4.27 $[0.89, 20.50]$ Gallagher 2015       2       43       0       43       0.4%       5.24 $[0.24, 112.45]$ Subtotal (95% CI)       2171       2152       21.7%       0.73 $[0.50, 1.07]$ Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)								
Walsh 2016       8       128       2       130       1.4%       4.27 [0.89, 20.50]         Gallagher 2015       2       43       0       43       0.4%       5.24 [0.24, 112.45]         Subtotal (95% CI)       2171       2152       21.7%       0.73 [0.50, 1.07]         Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)	-							
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Subtotal (95% CI)       2171       2152       21.7%       0.73 [0.50, 1.07]         Total events       59       78         Heterogeneity:       Tau² = 0.02; Chi² = 12.42, df = 12 (P = 0.41); I² = 3%         Test for overall effect:       Z = 1.62 (P = 0.11)	Walsh 2016							
Total events       59       78         Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3%         Test for overall effect: Z = 1.62 (P = 0.11)		2		0				
Heterogeneity: Tau <sup>2</sup> = 0.02; Chi <sup>2</sup> = 12.42, df = 12 (P = 0.41); l <sup>2</sup> = 3% Test for overall effect: Z = 1.62 (P = 0.11)	Subtotal (95% CI)		2171		2152	21.7%	0.73 [0.50, 1.07]	•
Test for overall effect: Z = 1.62 (P = 0.11)								
Total (95% CI) 4736 4703 100.0% 0.76 [0.63, 0.93]					: (P = 0	.41); l² = 3	%	
	Total (95% CI)		4736		4703	100.0%	0.76 [0.63, 0.93]	•
Total events 605 690	. ,	605		690				
Heterogeneity: Tau <sup>2</sup> = 0.08; Chi <sup>2</sup> = 45.82 df = 32 (P = 0.05); l <sup>2</sup> = 30%			= 45.82		(P = 0	.05); l² = 3	0%	
Test for overall effect: $Z = 2.71 (P = 0.007)$ 0.001 0.1 1 10	Test for overall effect:	Z = 2.71 (F	<b>&gt;</b> = 0.00	)7)		,.		0.001 0.1 1 10 1000 Favours RIPC Favours control

has been associated with activation of sensory (afferent) Cfiber neurons, which trigger the vagus nerve of a distant organ, such as the kidney, consequently exerting protective effects (49). Experimental studies showed that RIPC was associated with the release of various humoral factors, which confers anti-inflammatory and anti-oxidative efficacies (50), two predominant pathogenic mechanisms in AKI. However, the exact mechanisms and key molecular pathways involved in the potential renoprotective efficacy of RIPC remain to be determined.

Our study has some limitations. First, the definitions of AKI varied across the included studies. Although the difference between subgroups was not significantly different, our subgroup analyses suggest that RIPC reduced AKI after cardiac surgery



in studies that applied the AKIN definition but not in those that used the RIFLE or KDIGO definition. Since the optimal definition of AKI has yet to be established (51), we could not exclude the possibility that difference in the definition of AKI may affect the outcome of the meta-analysis. Secondly, no access to individual patient data was obtained. Therefore, subgroup analyses were based on study-level data, and results of the subgroup analyses should be interpreted with caution. Future large-scale RCTs or meta-analysis based on individual patient data are needed to validate the results. Moreover, the sample sizes of the included RCTs varied significantly. The two RCTs with largest sample sizes comprised over half of the included patients of the meta-analysis, which may primarily

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contribute to the results of overall meta-analysis and therefore comprise the importance of the meta-analysis. Besides, variations in the details of protocols for RIPC may affect its influence on AKI. We have performed subgroup analysis to evaluate the potential difference of RIPC performed in upper and lower limbs. However, influences of other details of RIPC, such as number of cycles, pressure applied and durations, and more importantly, the combinations of these factors could not be evaluated in the current meta-analysis. Finally, we did not perform stratified analyses according to the degree of AKI since data regarding the severity of AKI were rare in the included studies. Future studies are warranted in this regard.

In conclusion, results of this updated meta-analysis suggest that RIPC is effective for preventing AKI after cardiac surgery, which seems to be mainly driven by the reduced mild-to-moderate AKI events that did not require RRT. More studies are warranted to determine the influence of patient characteristics, procedure, perioperative drugs, and RIPC protocol on AKI outcome.

#### DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

#### **AUTHOR CONTRIBUTIONS**

ZL and YZ contributed to the conception and design of the study. ZL, YZ, ML, and GZ performed the statistical analysis. ZL, YZ, DL, and RS wrote the first draft of the manuscript. XL wrote sections of the manuscript. All authors contributed to manuscript revision and read and approved the submitted version.

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**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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