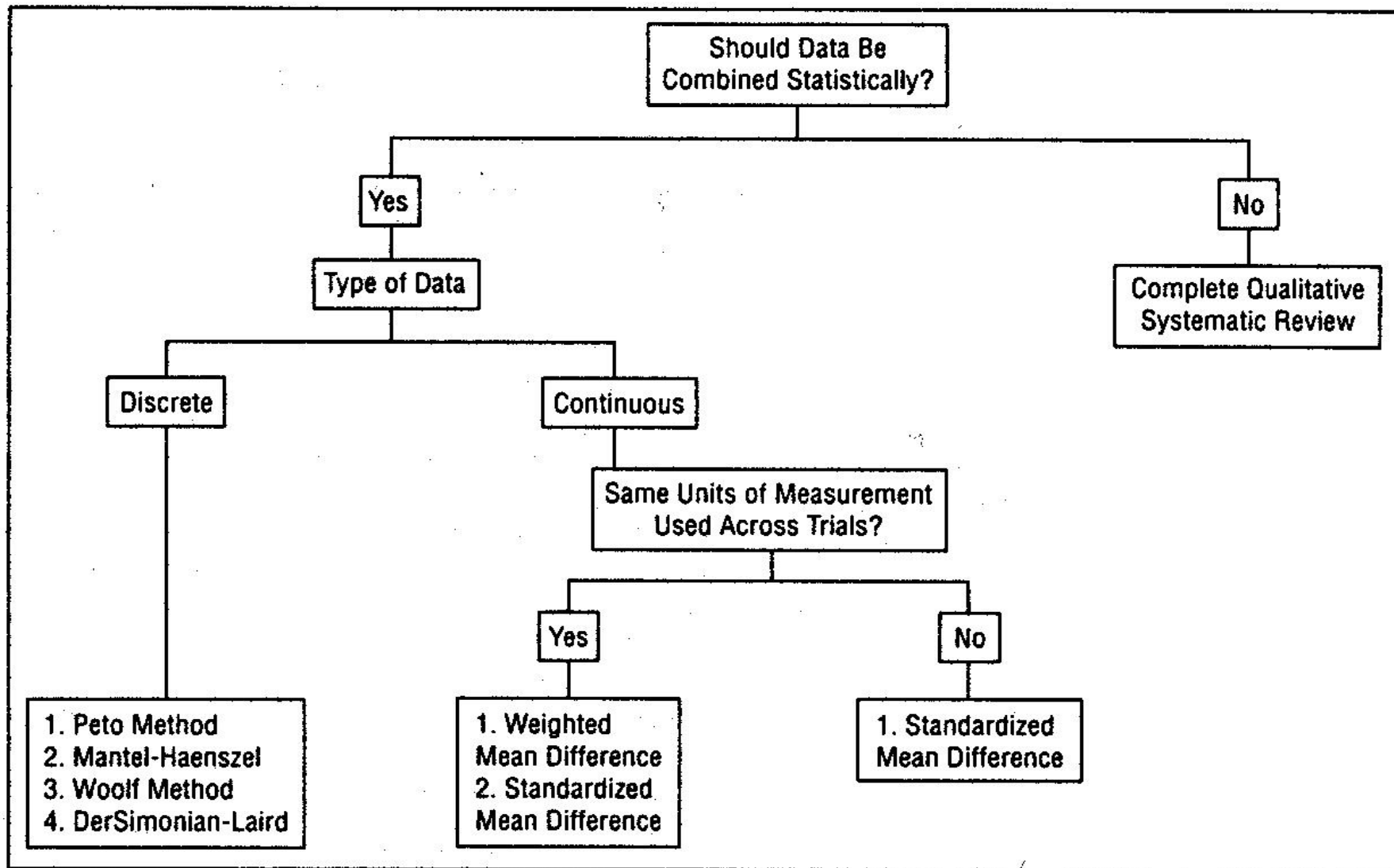
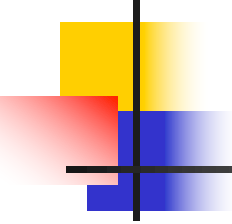


Revisão sistemática e meta-análise

Aula 4

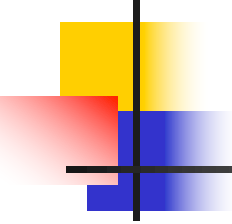


Algorithm of statistical choices available to systematic reviewers.



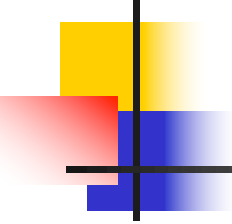
Exemplo

- Ingesta de fibras e incidência de câncer do endométrio
- 7 estudos observacionais
- Compara categorias de maior consumo com as de menor consumo



Ao comparar categorias extremas

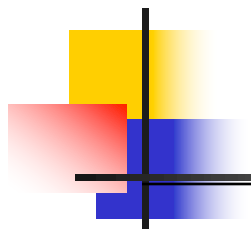
- Os extremos podem não ser similares
 - Um estudo pode comparar ingesta (0) contra a ingesta de uma certa quantia de fibra
 - Outro, compara o primeiro com o quarto quartil



O que fazer

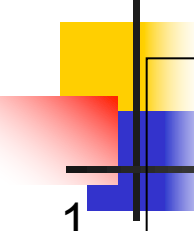
- Se existe algum motivo para comparar o “efeito” de uma certa quantidade de exposição, devemos excluir os estudos que não se enquadram nessa categoria
- Incluímos todos estudos e as categorias comparadas entram como uma variável

Author, year	Study design	Mean age at assessment	Gender	Categorization of breastfeeding	Odds ratio of type-2 diabetes among breastfed subjects
Young, 2002	Case-control	13 years	All	Breastfed $\geq$ 6 months vs. Breastfed $<$ 6 months	0.36 (0.13; 0.99)
Izadi, 2013	Cross-sectional	14 years	All	Breastfed $\geq$ 18 months vs. Breastfed $\leq$ 6 months	0.57 (0.26; 1.22)
Halipchuk, 2018	Case-control	14 years	All	Exclusively breastfed at hospital discharge vs. Formula fed at hospital discharge	0.52 (0.36; 0.74)
Mayer-Davis, 2008	Case-control	15 years	All	Ever breastfed vs. Never breastfed	0.43 (0.19; 0.99)
Evenhouse, 2005	Cross-sectional	15 years	All	Ever breastfed vs. Never breastfed	0.40 (SE: 0.24)
Fall, 2011	Cohort	25 years	All	Ever breastfed vs. Never breastfed	1.26 (0.63; 2.50)
Petit, 1997	Cohort	25 years	All	Exclusively breastfed vs. Exclusively bottle-fed at 2 months	0.41 (0.18; 0.93)
Baldassarre, 2017	Cross-sectional	26 years	All	Exclusively breastfed $\geq$ 6 months vs. Not exclusively breastfed $<$ 6 months	0.63 (0.41; 0.95)
Martens, 2016	Cohort	27 years	All	Breastfed before hospital discharge vs. Not breastfed before hospital discharge	0.83 (0.69; 0.99)
Parikh, 2009	Cohort	41 years	All	Ever breastfed vs. Never breastfed	0.40 (0.09; 1.70)
Ravelli, 2000	Cohort	50 years	All	Exclusively breastfed vs. Bottle-fed during hospital stay	0.51 (0.30; 0.90)
Martin, 2005	Cross-sectional	52 years	Male	Ever breastfed vs. Never breastfed	2.89 (0.65; 12.83)
Rich-Edwards, 2004	Cohort	59 years	Female	Ever breastfed vs. Never breastfed	0.79 (0.74; 0.85)
Martin, 2005	Cohort	71 years	All	Ever breastfed vs. Never breastfed	0.97 (0.41; 2.30)



Estudo	Risco relativo	95% I C
Barbone et al.	.6	(.3, 1.1)
Potischman et al.	.7	(.4, 1.3)
Shu et al.	1.1	-
Goodman et al.	.47	(.25, .86)
⋮	⋮	⋮

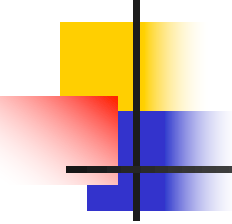
- Para dados dicotômicos, entre com os dados de uma das seguintes formas:

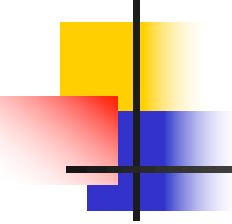


1	No. com desfecho no grupo intervenção	No. sem o desfecho no grupo intervenção	No. com desfecho no grupo controle	No. sem o desfecho no grupo controle
---	--	--	---	---

2	Effect size (Transformado em logaritmo se ratio)	Limite inferior IC	Limite superior IC
---	---	-----------------------	-----------------------

3	Effect size (Transformado em logaritmo se ratio)	Erro padrão
---	---	-------------

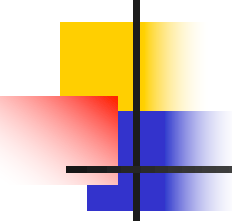
- 
-
- Para a opção 2, metan não aceita valor missing
 - LEMBRANDO: transforma em logaritmo natural razão de odds e risco relativo



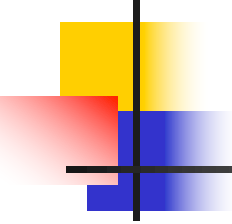
Comandos do STATA

- Excluindo estudos com valores missing
 - `drop if ll==. | ul==.`
- A seguir, transformação em logaritmo

```
gen lnrr = ln(rr)  
gen lnll = ln(ll)  
gen lnul = ln(ul)
```



meta Inrr rrl rrul, **eform**



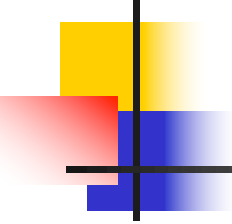
Erro Padrão

O intervalo de confiança (95%) para o risco relativo ou odds ratio θ é calculado através da seguinte fórmula:

$$\exp\left(\ln(\theta) \pm 1.96 \times EP[\ln(\theta)] \right)$$

O erro padrão para $\ln(\theta)$ é:

$$EP[\ln(\theta)] = \frac{\ln(\text{limite superior}) - \ln(\text{limite inferior})}{2 \times 1.96}$$



STATA

- `gen se =(lnul - lnll) / 3.92`
- Outro comando
 - `meta rr se`

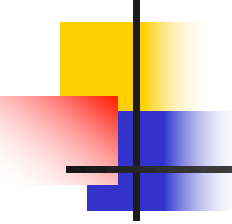
- Meta-analysis of 6 studies
- -----
- Fixed and random effects pooled estimates, lower and upper 95% confidence limits, and asymptotic z-test for null hypothesis that true effect=0
- Fixed effects estimation

- +-----+
- | Est Lower Upper z_value p_value |
- |-----|
- | 0.778 0.573 0.983 7.436 0.000 |
- +-----+

- Test for heterogeneity: Q= 7.085 on 5 degrees of freedom (p= 0.214)
- Der Simonian and Laird estimate of between studies variance = 0.029

- Random effects estimation

- +-----+
- | Est Lower Upper z_value p_value |
- |-----|
- | 0.756 0.505 1.008 5.893 0.000 |
- +-----+



meta Inrr se, ef

- Meta-analysis of 6 studies (exponential form)
- -----
- Fixed and random effects pooled estimates, lower and upper 95% confidence limits, and
- asymptotic z-test for null hypothesis that true effect=0

- Fixed effects estimation

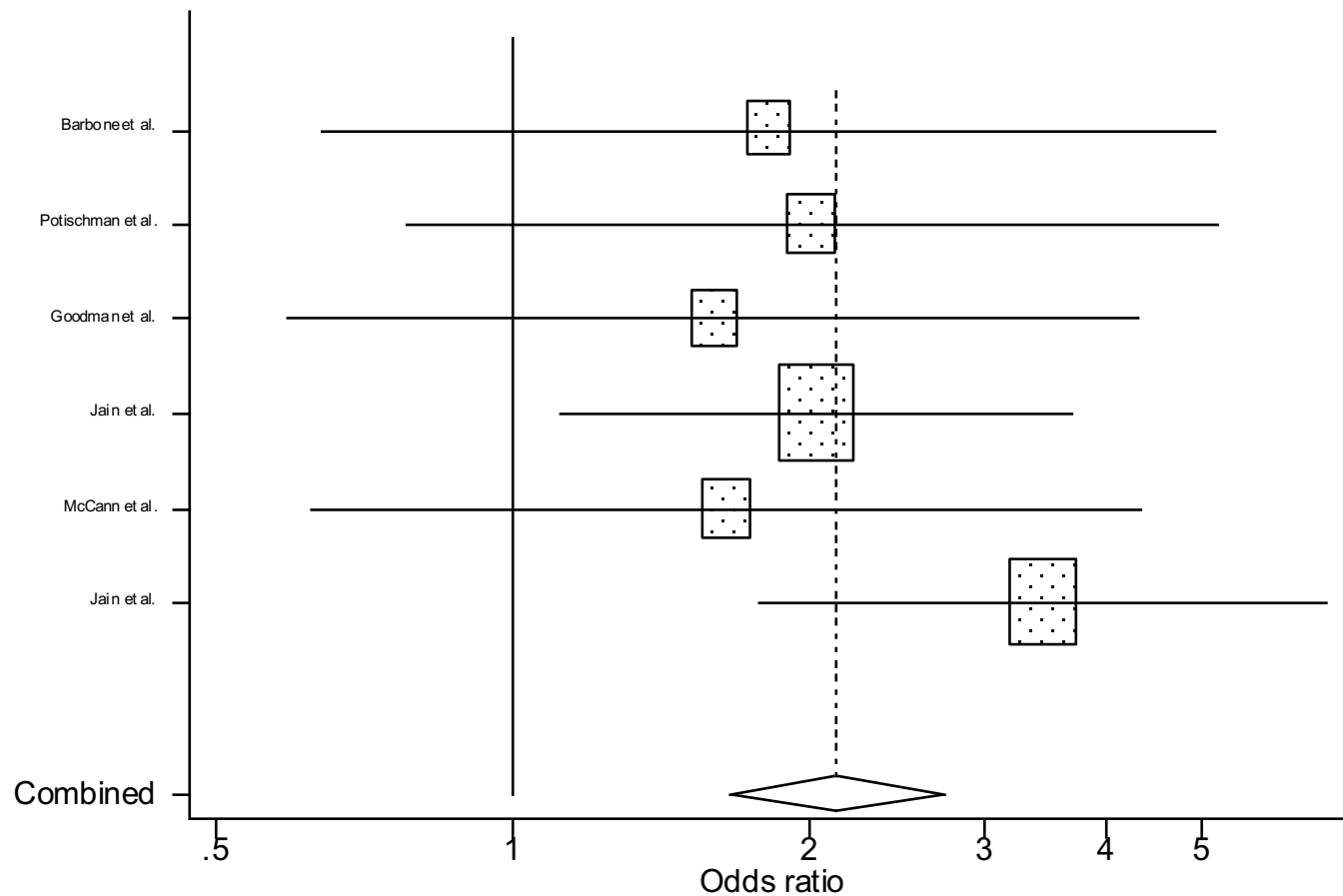
- +-----+
- | Est Lower Upper z_value p_value |
- |-----|
- | 2.178 1.774 2.673 7.436 0.000 |
- +-----+

- Test for heterogeneity: Q= 7.085 on 5 degrees of freedom (p= 0.214)
- Der Simonian and Laird estimate of between studies variance = 0.029

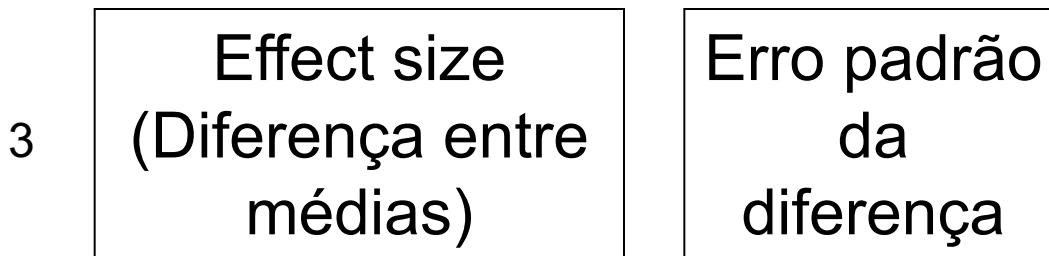
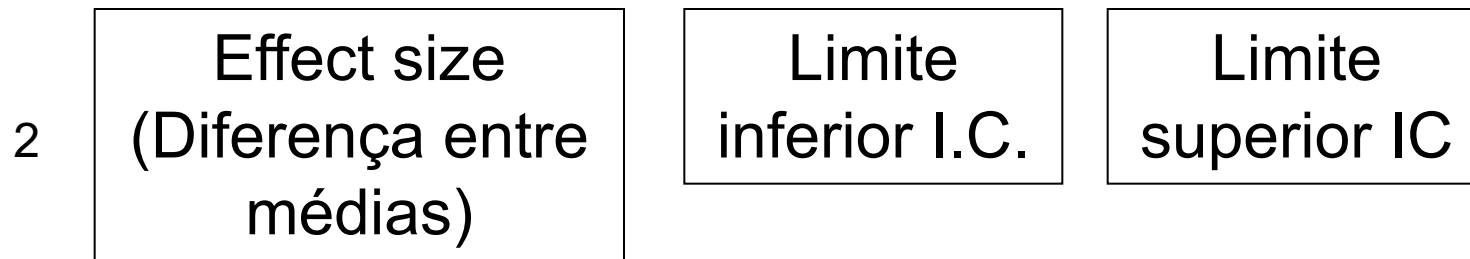
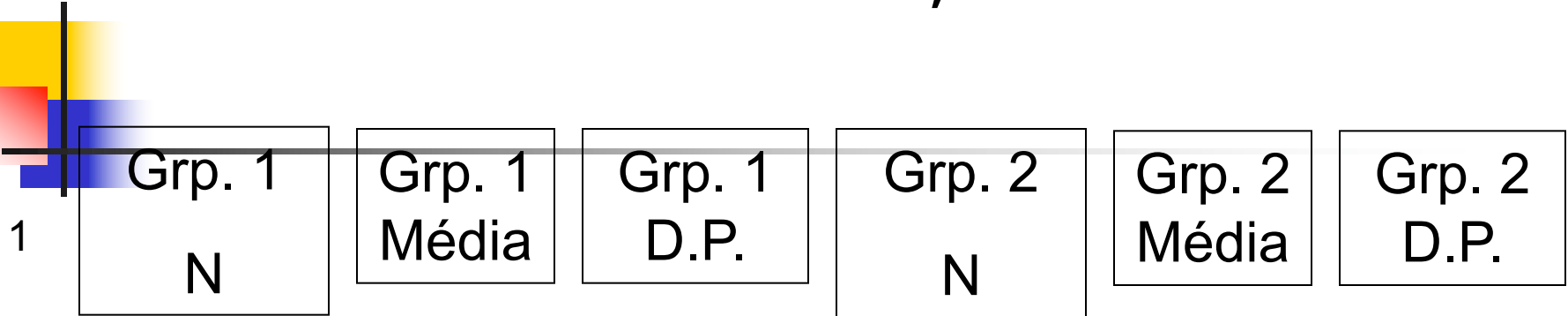
- Random effects estimation

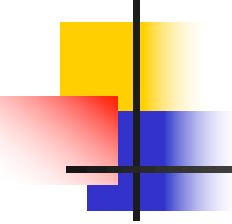
- +-----+
- | Est Lower Upper z_value p_value |
- |-----|
- | 2.130 1.657 2.739 5.893 0.000 |
- +-----+

meta rr se, ef graph(r) cline xlabel (0.5,1,2,3,4,5)
xline(1) id(author) b2title(Odds ratio)



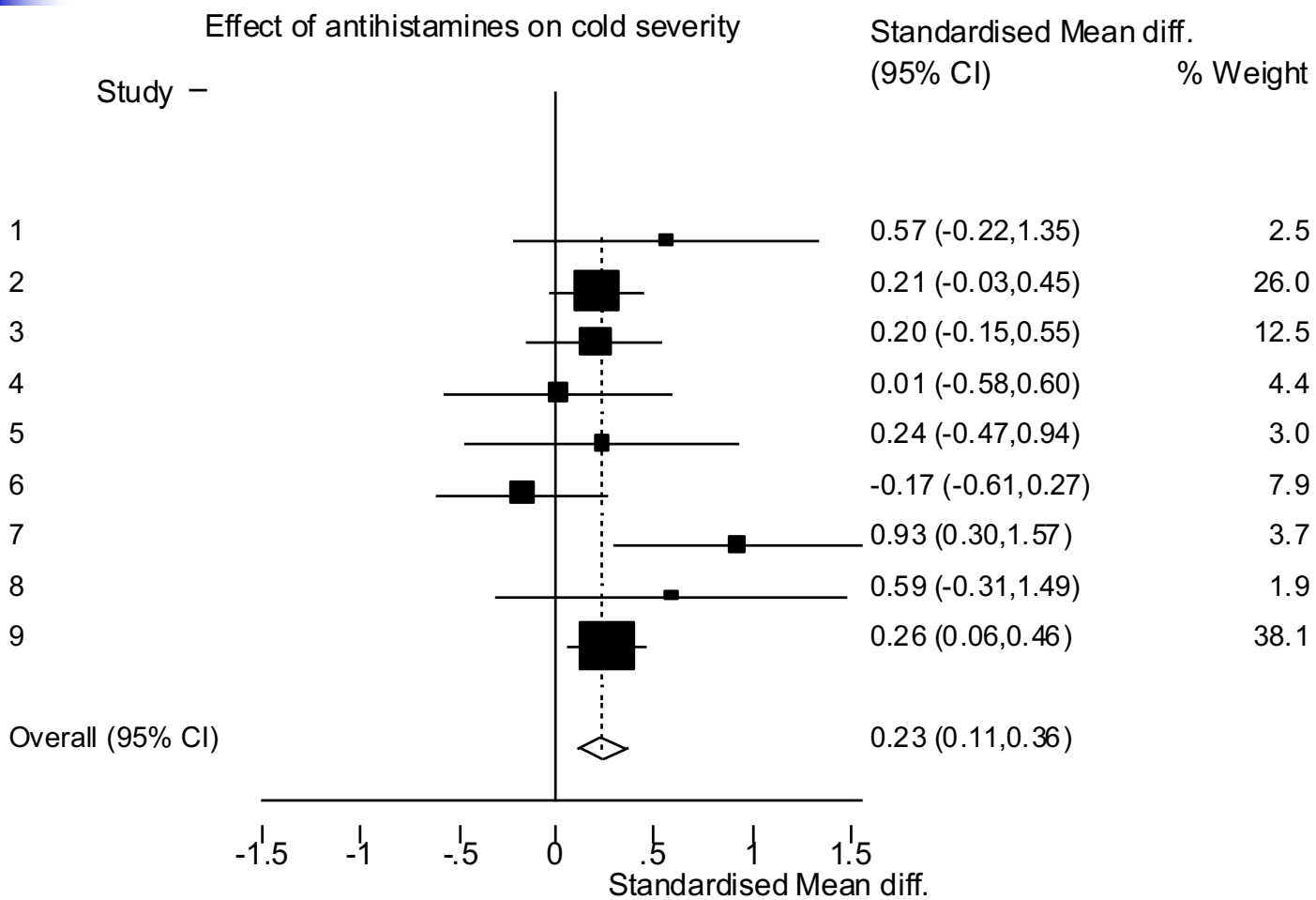
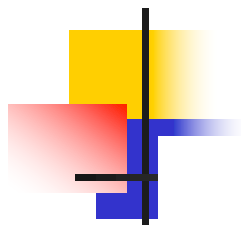
■ Para desfechos contínuos, escolha:

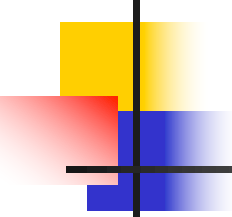




Comandos no Stata

```
metan n1 mean1 sd1 n2 mean2  
sd2, xlabel(-1.5,-1,-  
0.5,0,0.5,1,1.5) t1(Effect of  
antihistamines on cold  
severity)
```





Diferença Média

Todos os desfechos são medidos na mesma escala

$$MD_i = m_{1i} - m_{2i}$$

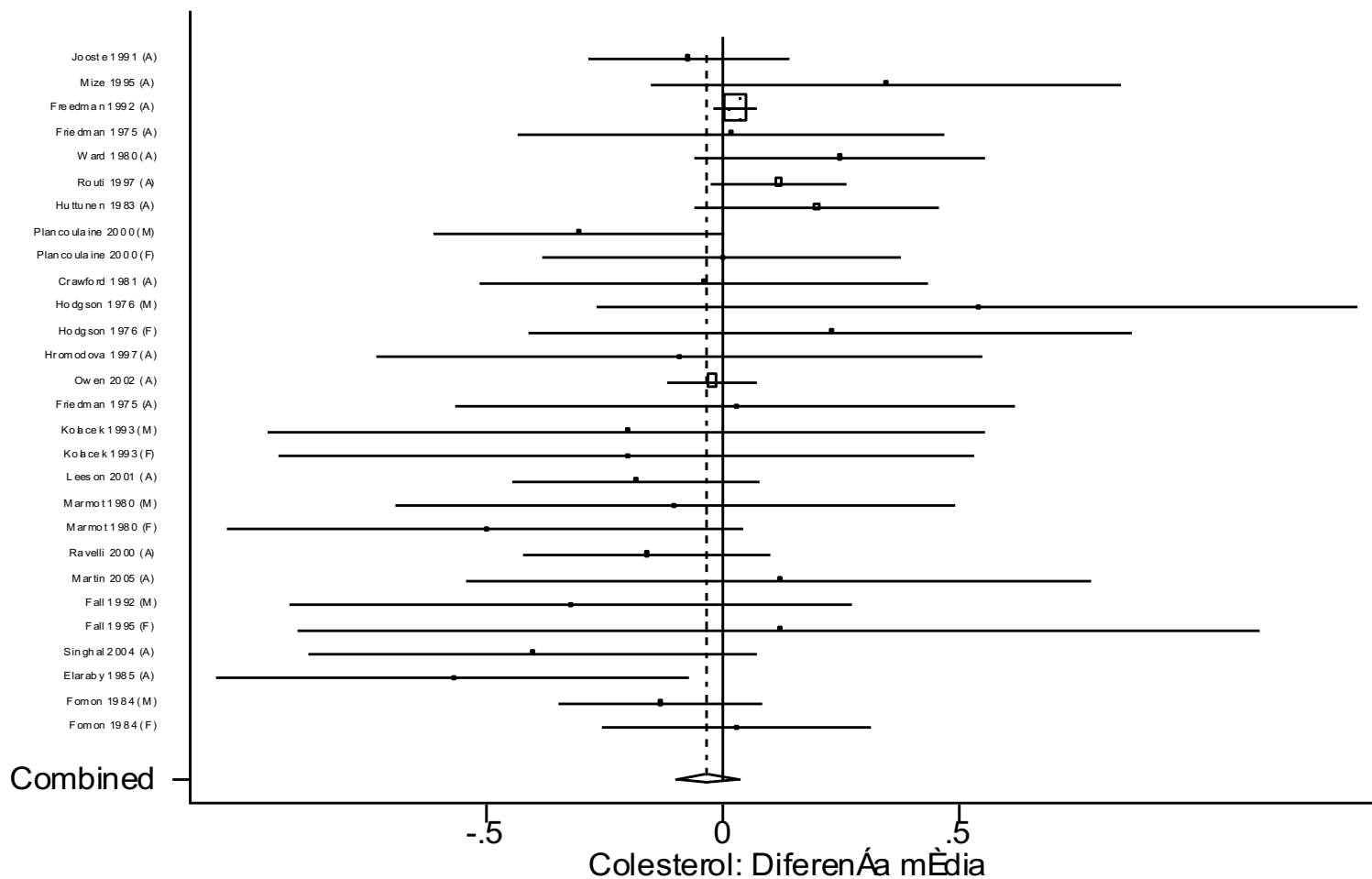
Diferença Média

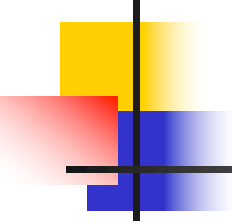
Erro Padrão

$$EP(MD_i) = \sqrt{\frac{SD_{1i}^2}{n_{1i}} + \frac{SD_{2i}^2}{n_{2i}}}$$

meta Mean SD, graph(r) cline xlabel (-0.5,0, 0.5)

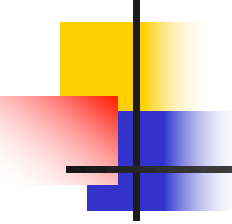
xline(0) id(Study) b2title(Colesterol: Diferença média)





Diferença padronizada

- Quanto os desfechos são medidos em diferentes escalas ou com diferentes instrumentos
- As medidas devem ser padronizadas, permitindo comparações



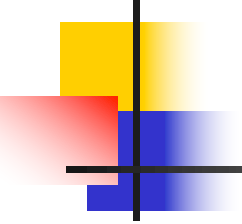
Cohen's d

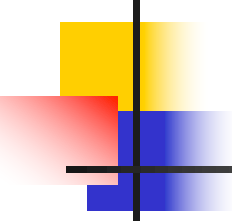
Diferença média

$$d_i = \frac{m_{1i} - m_{2i}}{s_i}$$

Erro padrão

$$SE(d_i) = \sqrt{\frac{N_i}{n_{1i}n_{2i}} + \frac{d_i^2}{2(N_i - 2)}}$$

- 
-
- Superestima o efeito, quando o tamanho da amostra é pequeno
 - Use apenas se todos os *trials* tem $N_i > 10$



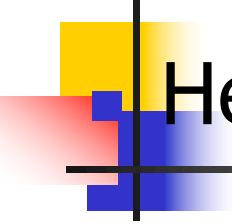
Hedges' g

Diferença média

$$g_i = \frac{m_{1i} - m_{2i}}{s_i} \left(1 - \frac{3}{4N_i - 9} \right)$$

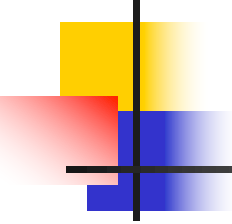
Erro padrão

$$SE(g_i) = \sqrt{\frac{N_i}{n_{1i}n_{2i}} + \frac{g_i^2}{2(N_i - 3.94)}}$$



Hedges' g é uma extensão do Cohen's d

- Usa um pequeno ajuste para produzir melhores estimativas quando o tamanho da amostra é pequeno
- Se N_i é grande, o ajuste é relativamente pequeno e a diferença entre os dois métodos é pequena.



Glass's Δ

Diferença média

$$\Delta_i = \frac{m_{1i} - m_{2i}}{SD_{2i}}$$

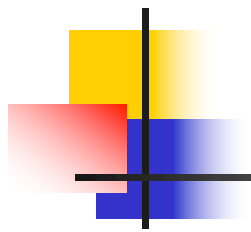
Erro padrão

$$SE(\Delta_i) = \sqrt{\frac{N_i}{n_{1i}n_{2i}} + \frac{\Delta_i^2}{2(N_{2i} - 1)}}$$

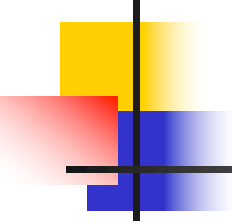


Glass's Δ usa o erro padrão do grupo controle, ao invés de usar o pooled.

- Tem melhor performance quando a intervenção altera a média e o desvio-padrão.

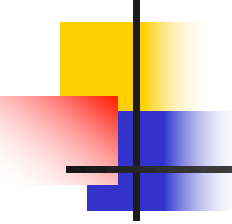


SENSIBILIDADE E ESPECIFICIDADE



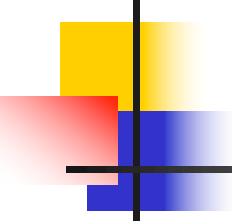
Lembrando

Doença			
Teste		Presente	Ausente
	Positivo	Verdadeiro positivo	Falso positivo
	Negativo	Falso negativo	Verdadeiro negativo



Algumas diferenças na análise

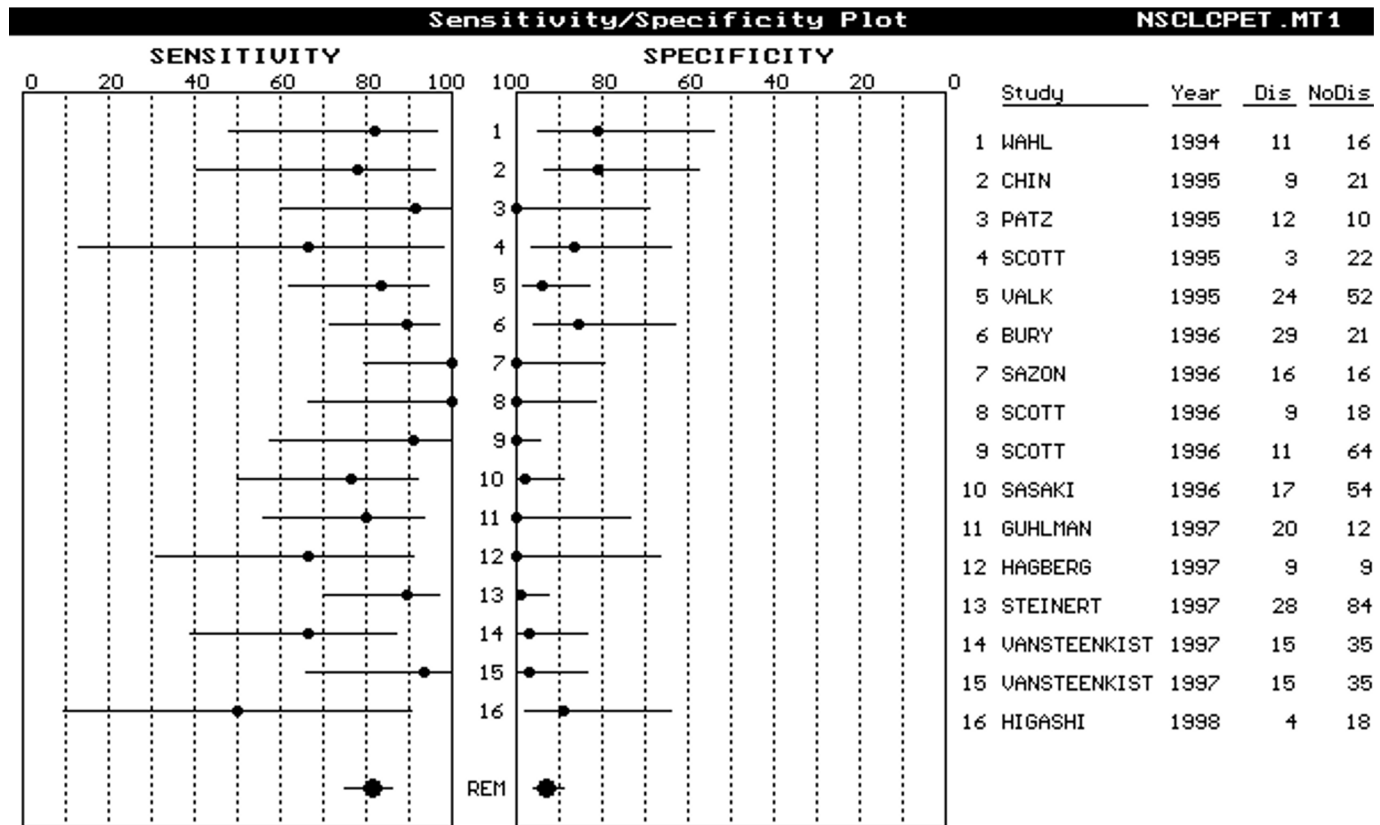
- ❑ Risco relativo no grupo intervenção $\{[a/(a + c)]/[b/(b + d)]\}$ = Likelihood Ratio – Teste Positivo.
- ❑ Risco Relativo no grupo controle = Likelihood Ratio – Teste Negativo.
- ❑ Odds Ratio (OR) $(a \times d)/(b \times c)$.

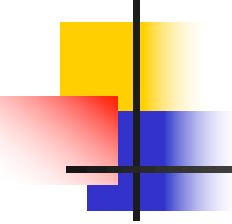


Entrando os dados

FIRST AUTHOR	TP	FN	FP	TN
WAHL	9	2	3	13
CHIN	7	2	4	17
PATZ	11	1	0	10
SCOTT	2	1	3	19
VALK	20	4	3	49
BURY	26	3	3	18
SAZON	16	0	0	16
SCOTT	9	0	0	18
SCOTT	10	1	0	64
SASAKI	13	4	1	53
GUHLMAN	16	4	0	12
HAGBERG	6	3	0	9
STEINERT	25	3	1	83
VANSTEENSKISTE	10	5	1	34
VANSTEENSKISTE	14	1	1	34
HIGASHI	2	2	2	16

Apresentando os resultados

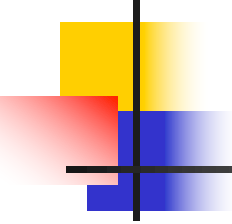




Diagnostic Odds Ratio

- Logit VP (sensibilidade) e FP (1 - especificidade).

$$D = \ln(DOR) = \text{logit}(VP) - \text{logit}(FP)$$



Diagnostic odds ratio

- Descreve em quantas vezes o odds de ter um teste positivo é maior em um indivíduo doente do que não doente.

. metandi tp fp fn tn

True positives: tp False positives: fp
False negatives: fn True negatives: tn

Refining starting values:

Iteration 0: log likelihood = -87.369886
Iteration 1: log likelihood = -87.286917
Iteration 2: log likelihood = -87.28439
Iteration 3: log likelihood = -87.284364

Performing gradient-based optimization:

Iteration 0: log likelihood = -87.284364
Iteration 1: log likelihood = -87.284283
Iteration 2: log likelihood = -87.284283

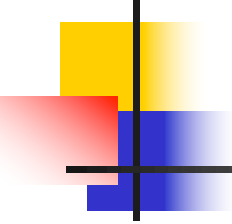
Meta-analysis of diagnostic accuracy

Log likelihood = -87.284283 Number of studies = 14

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Bivariate					
E(logitSe)	1.791051	.2928671			1.217042 2.36506
E(logitSp)	1.978175	.3348528			1.321876 2.634475
Var(logitSe)	.9840405	.4730037			.383585 2.524436
Var(logitSp)	1.229617	.5637082			.5006606 3.019927
Corr(logits)	.4370949	.2810972			-.2092853 .8176519
HSROC					
Lambda	3.764646	.5149784			2.755307 4.773985
Theta	.0113157	.3219068			-.6196101 .6422415
beta	.1113956	.3095365	0.36	0.719	-.4952848 .718076
s2alpha	3.1616	1.381317			1.342806 7.443902
s2theta	.3095969	.1704515			.1052349 .9108222
Summary pt.					
Se	.857056	.0358794			.7715425 .9141238
Sp	.8784865	.0357449			.7894936 .9330476
DOR	43.34649	22.38929			15.75042 119.2932
LR+	7.053175	2.195976			3.831481 12.98383
LR-	.1627162	.0436108			.0962263 .2751487
1/LR-	6.145671	1.647146			3.634398 10.39217

Covariance between estimates of E(logitSe) & E(logitSp) .0344472

Meta análise com dados faltantes

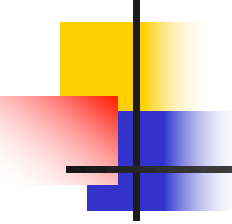


- Perda de informações sobre o desfecho compromete a validade de um ensaio clínico
- Compromete a meta-análise
- Isto é sobre dados missing em estudos, não sobre estudos perdidos

Haloperidol

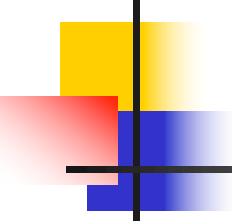
r=success0
f=falha
m=missing
n=total

	Haloperidol				Placebo				% missing
	r1	f1	m1	n1	r2	f2	m2	n2	
Arvanitis	25	25	2	52	18	33	0	51	2%
Beasley	29	18	22	69	20	14	34	68	41%
Bechelli	12	17	1	30	2	28	1	31	3%
Borison	3	9	0	12	0	12	0	12	0%
Chouinard	10	11	0	21	3	19	0	22	0%
Durost	11	8	0	19	1	14	0	15	0%
Garry	7	18	1	26	4	21	1	26	4%
Howard	8	9	0	17	3	10	0	13	0%
Marder	19	45	2	66	14	50	2	66	3%
Nishikawa 82	1	9	0	10	0	10	0	10	0%
Nishikawa 84	11	23	3	37	0	13	0	13	6%
Reschke	20	9	0	29	2	9	0	11	0%
Selman	17	1	11	29	7	4	18	29	50%
Serafetinides	4	10	0	14	0	13	1	14	4%
Simpson	2	14	0	16	0	7	1	8	4%
Spencer	11	1	0	12	1	11	0	12	0%
Vichaiya	9	20	1	30	0	29	1	30	3%



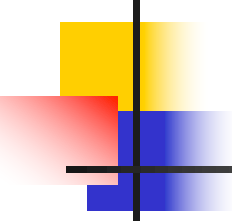
Abordagem padrão para dados faltantes

- Casos disponíveis (com informação completa): dados faltantes são excluídos
 - assume MAR: *Missing at random em cada braço do estudo*
- Assume missing=falha
 - Implausível, mas não é ruim para desfechos em saúde



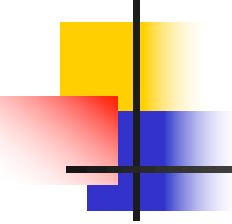
Outras ideias

- Análise de sensibilidade
 - missing=falha
 - Casos com dados completos
- Análise pior e melhor cenário
- Informative missingness (IM)
 - IM: *perda é dependente do desfecho*



metamiss

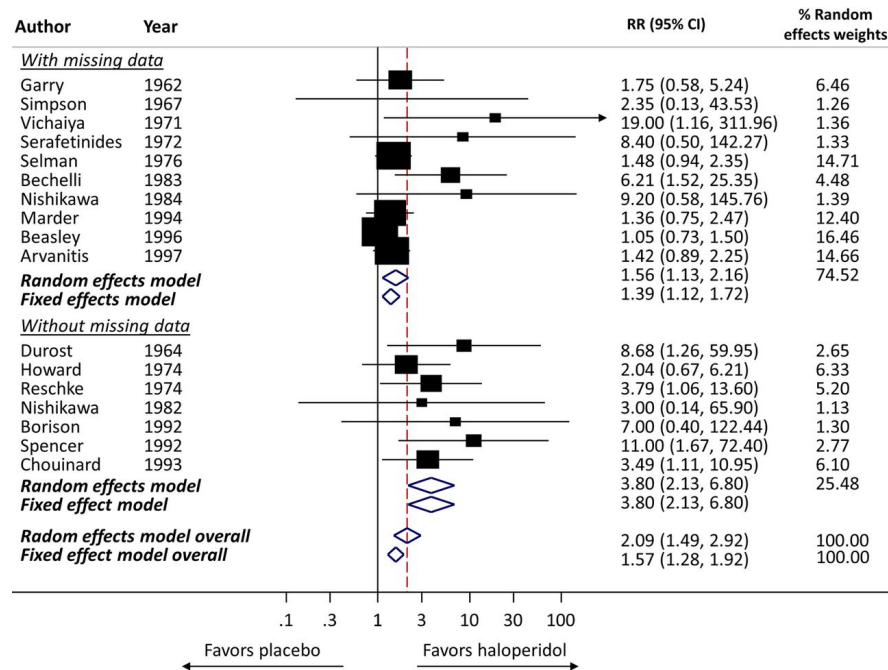
- Análise com casos disponíveis(ACA)
- Análise com casos imputados(ICA):
 - Missing é imputado como falha: ICA-0
 - Missing é imputado como sucesso: ICA-1
 - Melhor cenário: ICA-b (missing=sucesso em E, falha em C)
 - Pior cenário: ICA-w (missing=falha em E, sucesso em C)
 - Usa a probabilidade de desfecho no controle para a imputação : ICA-pC
 - Usa a probabilidade de desfecho no grupo intervenção para a imputação : ICA-pE
 - Usa a probabilidade de cada grupo para a imputação : ICA-p(concorda com ACA)

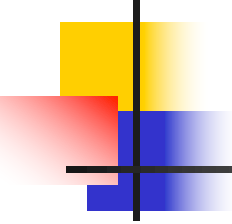


IMORs

- Informative Missing Odds Ratio (IMOR):
 - Odds ratio entre desfecho e perdas em cada grupo
- IMOR
 - 1 – indica que as perdas são totalmente ao acaso
 - 0 – indica que todas as perdas são falha
 - ∞ – indica que todas as perdas são sucesso
- IMOR não é estimado a partir do dado, mas dado um valor de IMOR, é possível analisar os dados
- Por exemplo,
 - ICA-0 usa IMORs 0, 0
 - ICA-1 usa IMORs ∞ , ∞
 - ICA-b usa IMORs ∞ , 0
 - ICA-p usa IMORs 1, 1

Explorando missing





ACA

- Ignora missing
- Assume que os casos missing ocorrem ao acaso
- Resultado igual ao obtido com o comando meta, usando apenas os dados disponíveis

ACA

```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) aca
*****
***** METAMISS: meta-analysis allowing for missing data *****
***** Available cases analysis *****
*****
Measure: RR.
Zero cells detected: adding 1/2 to 6 studies.
```

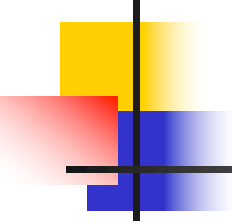
```
(Calling metan with options: label(namevar=author) fixed eform nograph ...)
```

Study	ES	[95% Conf. Interval]		% Weight
Arvanitis	1.417	0.891	2.252	18.86
Beasley	1.049	0.732	1.504	31.22
Bechelli	6.207	1.520	25.353	2.05
Borison	7.000	0.400	122.442	0.49
Chouinard	3.492	1.113	10.955	3.10
Durost	8.684	1.258	59.946	1.09
Garry	1.750	0.585	5.238	3.37
Howard	2.039	0.670	6.208	3.27
Marder	1.357	0.747	2.466	11.37
Nishikawa_82	3.000	0.137	65.903	0.42
Nishikawa_84	9.200	0.581	145.759	0.53
Reschke	3.793	1.058	13.604	2.48
Selman	1.484	0.936	2.352	19.11
Serafetinides	8.400	0.496	142.271	0.51
Simpson	2.353	0.127	43.529	0.48
Spencer	11.000	1.671	72.396	1.14
Vichaiya	19.000	1.157	311.957	0.52
I-V pooled ES	1.567	1.281	1.916	100.00

```
Heterogeneity chi-squared = 27.29 (d.f. = 16) p = 0.038
I-squared (variation in ES attributable to heterogeneity) = 41.4%
```

```
Test of ES=1 : z= 4.37 p = 0.000
```

```
.
end of do-file
```



ICA-0

- Todas as perdas são consideradas como sendo falha

```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) w4 ica0
```

```
*****
```

```
***** METAMISS: meta-analysis allowing for missing data *****
```

```
***** Simple imputation *****
```

```
*****
```

```
Measure: RR.
```

```
Method: ICA-0 (impute zeros).
```

```
Weighting scheme: w4.
```

```
Zero cells detected: adding 1/2 to 6 studies.
```

```
(Calling metan with options: label(namevar=author) fixed eform nograph ...)
```

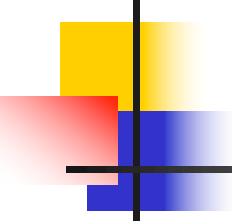
Study	ES	[95% Conf. Interval]		% Weight
Arvanitis	1.362	0.854	2.172	24.38
Beasley	1.429	0.901	2.266	25.01
Bechelli	6.200	1.513	25.402	2.67
Borison	7.000	0.400	122.442	0.65
Chouinard	3.492	1.113	10.955	4.06
Durost	8.684	1.258	59.946	1.42
Garry	1.750	0.582	5.266	4.38
Howard	2.039	0.670	6.208	4.29
Marder	1.357	0.745	2.473	14.75
Nishikawa_82	3.000	0.137	65.903	0.56
Nishikawa_84	8.474	0.534	134.463	0.70
Reschke	3.793	1.058	13.604	3.26
Selman	2.429	1.189	4.960	10.42
Serafetinides	9.000	0.530	152.927	0.66
Simpson	2.647	0.142	49.419	0.62
Spencer	11.000	1.671	72.396	1.50
Vichaiya	19.000	1.156	312.417	0.68
I-V pooled ES	1.898	1.507	2.390	100.00

```
Heterogeneity chi-squared = 21.56 (d.f. = 16) p = 0.158
```

```
I-squared (variation in ES attributable to heterogeneity) = 25.8%
```

```
Test of ES=1 : z= 5.45 p = 0.000
```

```
.  
end of do-file
```



ICA-1

- Perdas são consideradas como sendo sucesso


```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) w4 ical
```

```
*****
```

```
***** METAMISS: meta-analysis allowing for missing data *****
```

```
***** Simple imputation *****
```

```
*****
```

```
Measure: RR.
```

```
Method: ICA-1 (impute ones).
```

```
Weighting scheme: w4.
```

```
Zero cells detected: adding 1/2 to 3 studies.
```

```
(Calling metan with options: label(namevar=author) fixed eform nograph ...)
```

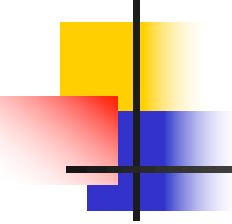
Study		ES	[95% Conf. Interval]		% Weight
Arvanitis		1.471	0.934	2.317	5.95
Beasley		0.931	0.773	1.120	35.81
Bechelli		4.478	1.417	14.151	0.93
Borison		7.000	0.400	122.442	0.15
Chouinard		3.492	1.113	10.955	0.94
Durost		8.684	1.258	59.946	0.33
Garry		1.600	0.603	4.247	1.29
Howard		2.039	0.670	6.208	0.99
Marder		1.312	0.754	2.283	4.01
Nishikawa_82		3.000	0.137	65.903	0.13
Nishikawa_84		10.684	0.682	167.429	0.16
Reschke		3.793	1.058	13.604	0.75
Selman		1.120	0.953	1.316	47.38
Serafetinides		4.000	0.509	31.456	0.29
Simpson		1.000	0.106	9.444	0.24
Spencer		11.000	1.671	72.396	0.35
Vichaiya		10.000	1.364	73.328	0.31
I-V pooled ES		1.156	1.035	1.292	100.00

```
Heterogeneity chi-squared = 40.34 (d.f. = 16) p = 0.001
```

```
I-squared (variation in ES attributable to heterogeneity) = 60.3%
```

```
Test of ES=1 : z= 2.57 p = 0.010
```

```
.  
end of do-file
```



Melhor cenário

- missing=sucesso em E, falha em C

```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) w4 icapc
*****
***** METAMISS: meta-analysis allowing for missing data *****
***** Simple imputation *****
*****
Measure: RR.
Method: ICA-pC (impute control group risk).
Weighting scheme: w4.
Zero cells detected: adding 1/2 to 6 studies.
```

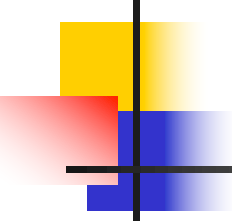
(Calling metan with options: label(namevar=author) fixed eform nograph ...)

Study	ES	[95% Conf. Interval]		% Weight
Arvanitis	1.401	0.880	2.230	19.89
Beasley	1.033	0.718	1.486	32.56
Bechelli	6.033	1.475	24.682	2.17
Borison	7.000	0.400	122.442	0.53
Chouinard	3.492	1.113	10.955	3.29
Durost	8.684	1.258	59.946	1.15
Garry	1.721	0.574	5.161	3.57
Howard	2.039	0.670	6.208	3.47
Marder	1.346	0.741	2.448	12.05
Nishikawa_82	3.000	0.137	65.903	0.45
Nishikawa_84	8.553	0.539	135.713	0.56
Reschke	3.793	1.058	13.604	2.64
Selman	1.300	0.759	2.228	14.85
Serafetinides	8.400	0.496	142.271	0.54
Simpson	2.353	0.127	43.529	0.51
Spencer	11.000	1.671	72.396	1.21
Vichaiya	18.419	1.121	302.646	0.55
I-V pooled ES	1.530	1.243	1.883	100.00

Heterogeneity chi-squared = 27.61 (d.f. = 16) p = 0.035
I-squared (variation in ES attributable to heterogeneity) = 42.0%

Test of ES=1 : z= 4.02 p = 0.000

```
.
end of do-file
```



Pior cenário

```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) w4 icab
*****
***** METAMISS: meta-analysis allowing for missing data *****
***** Simple imputation *****
*****
```

Measure: RR.
Method: ICA-b ('best'-case [0 in exp, 1 in control]).
Weighting scheme: w4.
Zero cells detected: adding 1/2 to 6 studies.

(Calling metan with options: label(namevar=author) fixed eform nograph ...)

Study	ES	[95% Conf. Interval]		% Weight
Arvanitis	1.471	0.934	2.317	22.59
Beasley	2.513	1.695	3.727	30.05
Bechelli	6.717	1.654	27.279	2.37
Borison	7.000	0.400	122.442	0.57
Chouinard	3.492	1.113	10.955	3.57
Durost	8.684	1.258	59.946	1.25
Garry	2.000	0.686	5.831	4.07
Howard	2.039	0.670	6.208	3.76
Marder	1.500	0.837	2.689	13.68
Nishikawa_82	3.000	0.137	65.903	0.49
Nishikawa_84	10.684	0.682	167.429	0.62
Reschke	3.793	1.058	13.604	2.86
Selman	4.000	2.091	7.654	11.08
Serafetinides	9.000	0.530	152.927	0.58
Simpson	2.647	0.142	49.419	0.54
Spencer	11.000	1.671	72.396	1.31
Vichaiya	21.000	1.286	342.934	0.60
I-V pooled ES	2.416	1.947	2.998	100.00

Heterogeneity chi-squared = 21.60 (d.f. = 16) p = 0.156
I-squared (variation in ES attributable to heterogeneity) = 25.9%

Test of ES=1 : z= 8.00 p = 0.000

end of do-file

```
. metamiss r1 f1 m1 r2 f2 m2, fixed nograph id(author) w4 icaw
*****
***** METAMISS: meta-analysis allowing for missing data *****
***** Simple imputation *****
*****
```

Measure: RR.
Method: ICA-w ('worst'-case [1 in exp, 0 in control]).
Weighting scheme: w4.
Zero cells detected: adding 1/2 to 3 studies.

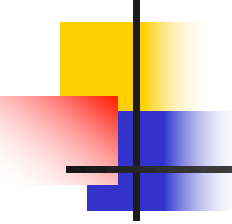
(Calling metan with options: label(namevar=author) fixed eform nograph ...)

Study	ES	[95% Conf. Interval]		% Weight
Arvanitis	1.362	0.854	2.172	13.99
Beasley	0.529	0.391	0.716	33.33
Bechelli	4.133	1.294	13.202	2.26
Borison	7.000	0.400	122.442	0.37
Chouinard	3.492	1.113	10.955	2.33
Durost	8.684	1.258	59.946	0.82
Garry	1.400	0.510	3.847	2.98
Howard	2.039	0.670	6.208	2.46
Marder	1.187	0.671	2.102	9.35
Nishikawa_82	3.000	0.137	65.903	0.32
Nishikawa_84	8.474	0.534	134.463	0.40
Reschke	3.793	1.058	13.604	1.87
Selman	0.680	0.485	0.954	26.57
Serafetinides	4.000	0.509	31.456	0.72
Simpson	1.000	0.106	9.444	0.60
Spencer	11.000	1.671	72.396	0.86
Vichaiya	9.000	1.214	66.704	0.76
I-V pooled ES	0.944	0.793	1.124	100.00

Heterogeneity chi-squared = 62.07 (d.f. = 16) p = 0.000
I-squared (variation in ES attributable to heterogeneity) = 74.2%

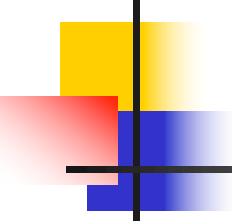
Test of ES=1 : z= 0.64 p = 0.519

end of do-file



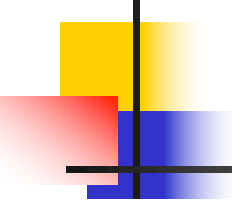
Meta análise em rede

- Tradicional
 - Compara dois tratamentos
- Se temos vários tratamentos
 - Estudos comparam A com B
 - Outros comparam B com C
- Meta análise tradicional
 - Compara A com B
 - B com C
 - Mas não A com C
- A com B e B com C são comparações diretas



Meta-análise em rede

- Permite
 - Comparar A com B
 - B com C
 - Também A com C – comparação indireta



Fazer o set – up do banco

```
. network setup r n, studyvar(study) trtvar(treat)
```

Treatments used

A (reference):	A
B:	B
C:	C
D:	D
E:	E
F:	F
G:	G
H:	H

Measure	Log odds ratio
---------	----------------

Studies

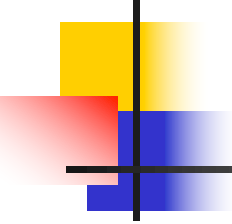
ID variable:	study
Number used:	28
IDs with zero cells:	[none]
IDs with augmented reference arm:	17 18 19 20 21 22 23 24 25 26 27 28
– observations added:	0.001
– mean in augmented observations:	study-specific mean

Network information

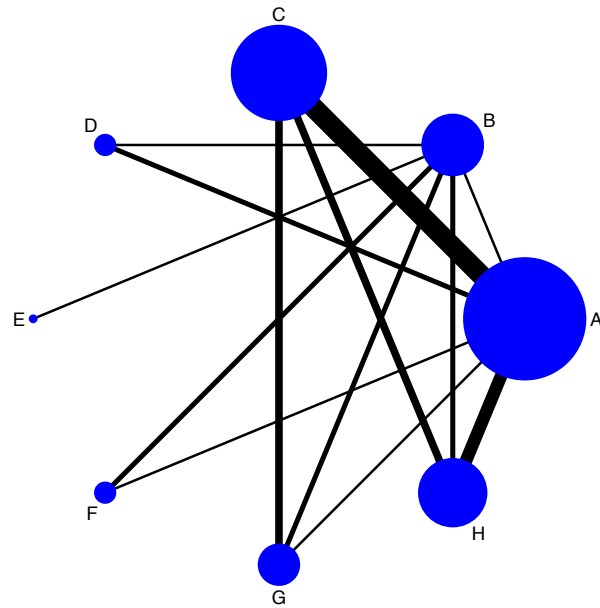
Components:	1 (connected)
D.f. for inconsistency:	8
D.f. for heterogeneity:	15

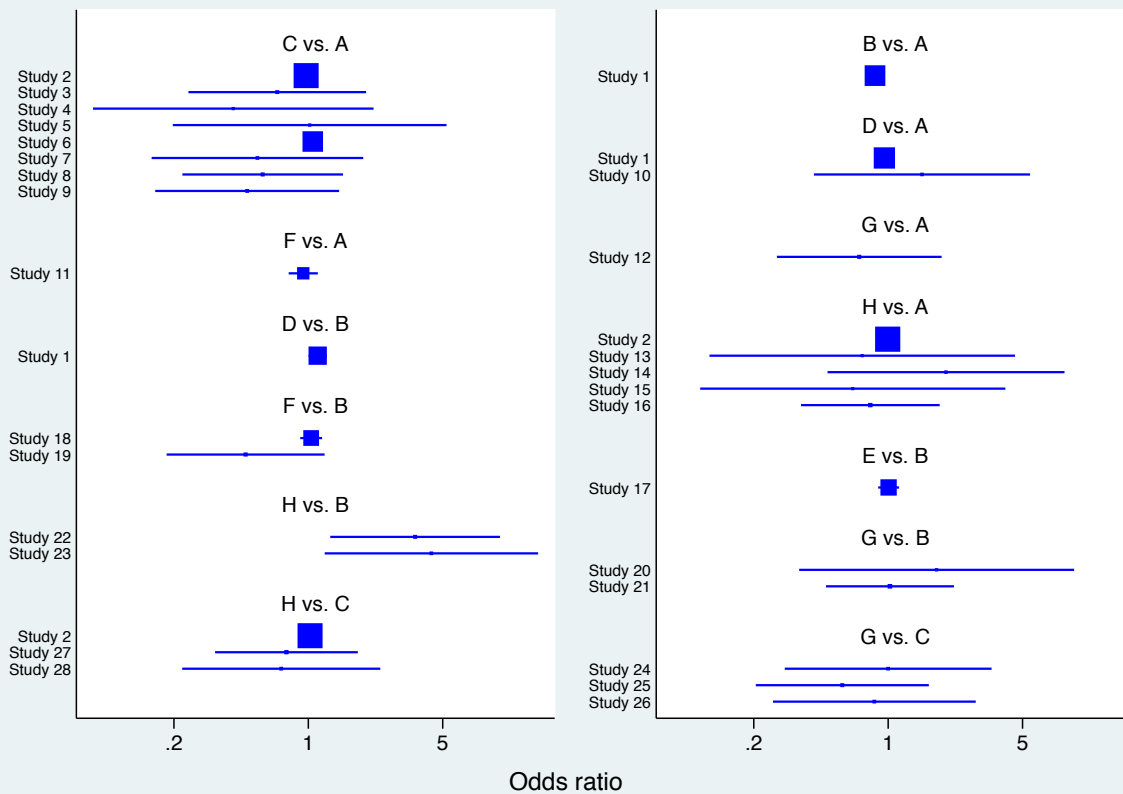
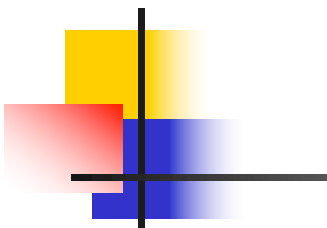
Current data

Data format:	augmented
Design variable:	_design
Estimate variables:	_y*
Variance variables:	_S*
Command to list the data:	list study _y* _S*, noo sepby(_design)

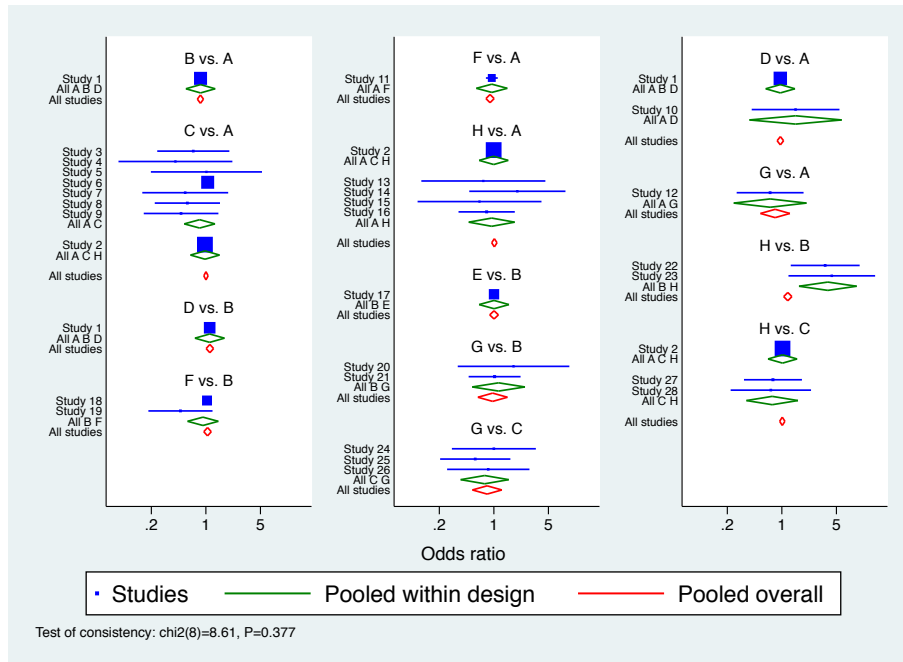


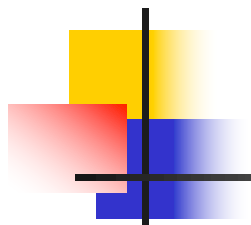
network mapgraph export thrombmap1.eps,
replace





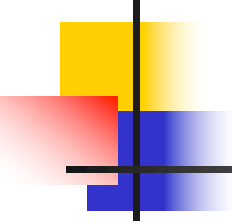
Após estimar todas as comparações





. network sidesplit all

Side	Direct		Indirect		Difference		
	Coef.	Std. Err.	Coef.	Std. Err.	Coef.	Std. Err.	P> z
A B	-.1575349	.062153	-.1895708	.1255688	.0320359	.1400551	0.819
A C	-.0030623	.0320637	.4616848	.3032621	-.4647471	.3049371	0.127
A D *	-.0406997	.0624884	-.1265838	.2884612	.085884	.2952869	0.771
A F	-.0603276	.0972427	-.1720842	.1135808	.1117566	.1495217	0.455
A G	-.3462582	.5035061	-.1614462	.2467496	-.184812	.5607172	0.742
A H	-.0058175	.0389527	.1694457	.1107251	-.1752632	.1173529	0.135
B D	.1122506	.0713263	.2353325	.3053638	-.1230818	.3140009	0.695
B E *	.005435	.0656528	.322518	238.4321	-.317083	238.4321	0.999
B F	.0146997	.0782736	.1264566	.1216721	-.1117568	.1495227	0.455
B G	.1212546	.3545816	-.1382147	.2860893	.2594694	.4555891	0.569
B H	1.354421	.4059271	.1496602	.0606317	1.204761	.4104342	0.003
C G	-.2756473	.339353	-.1424003	.2933971	-.1332471	.4486046	0.766
C H	-.017902	.1392018	.3408459	.3755195	-.3587479	.4313328	0.406



Meta p

■ Combina valores de p

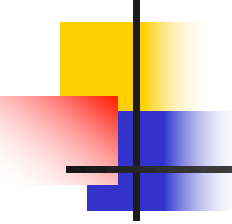
```
. metap pvar
```

Meta-analysis of p-values

Method		chi2	p_value	studies

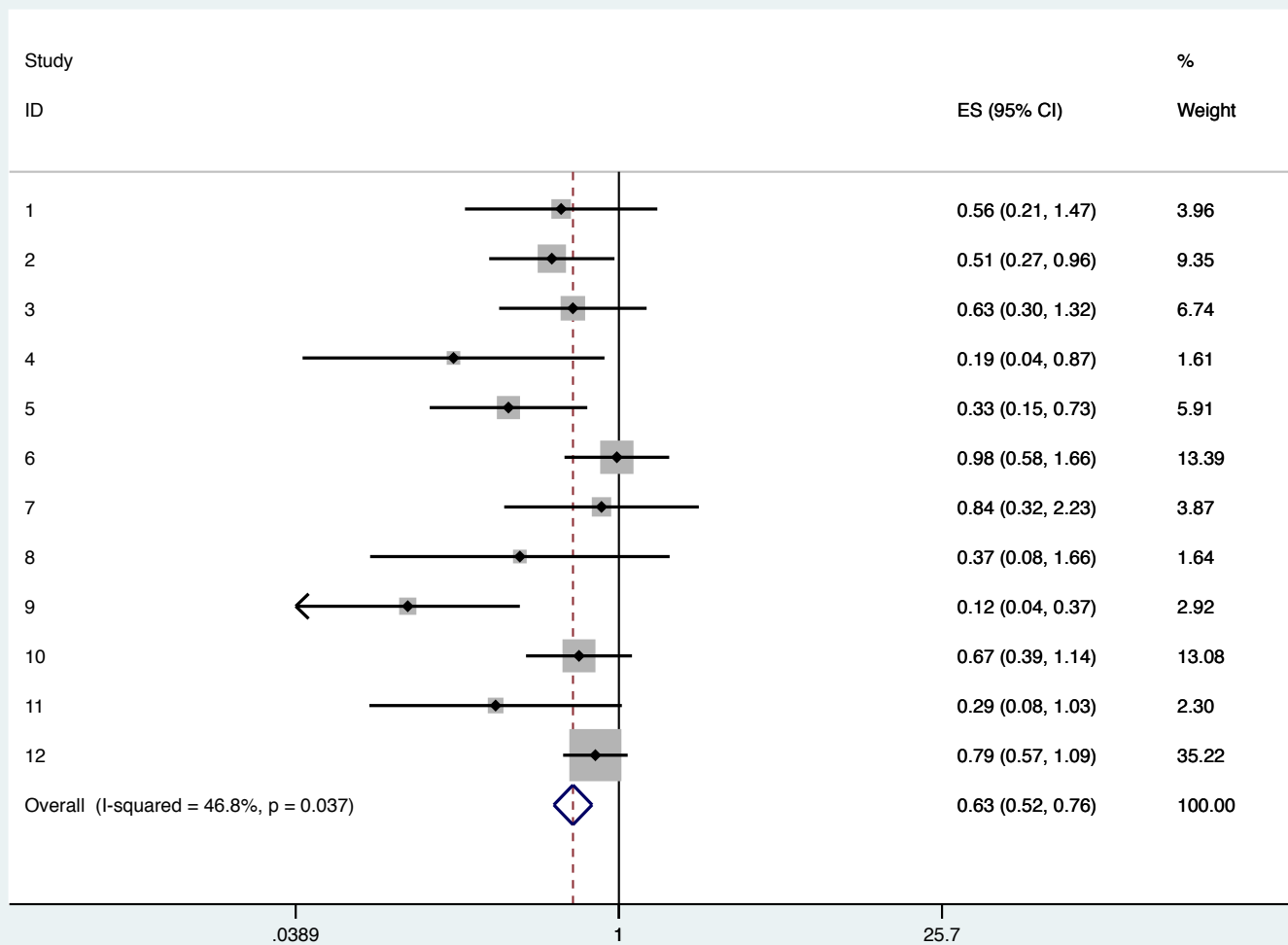
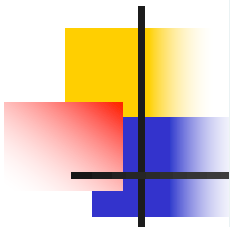
Fisher		38.938246	.00037283	7

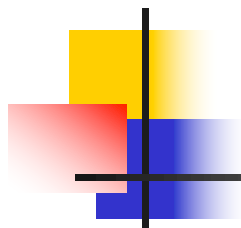
```
.
```



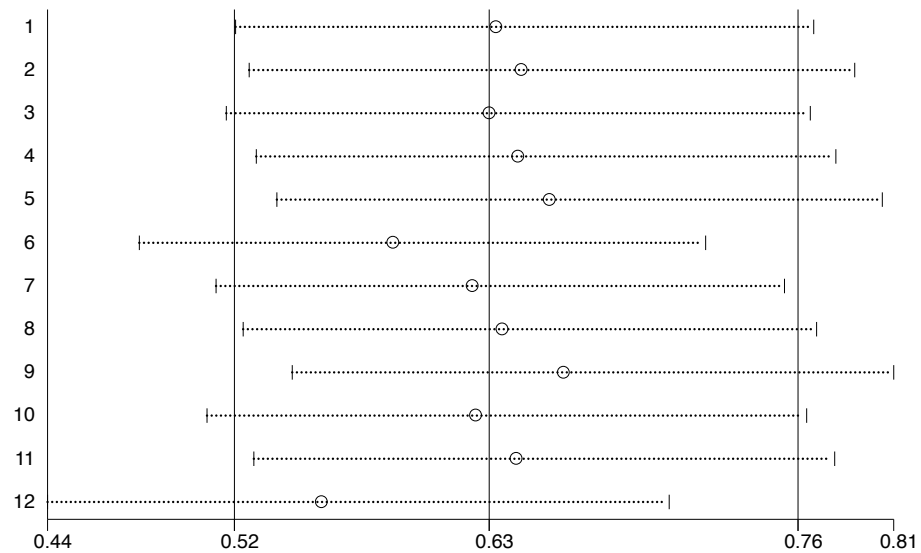
metainf

- Retira um estudo de cada vez e faz uma nova meta-análise
- Serve para avaliar se algum estudo está influenciando o pooled effect

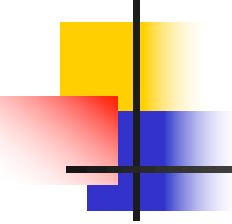




Meta-analysis fixed-effects estimates (exponential form)
Study omitted

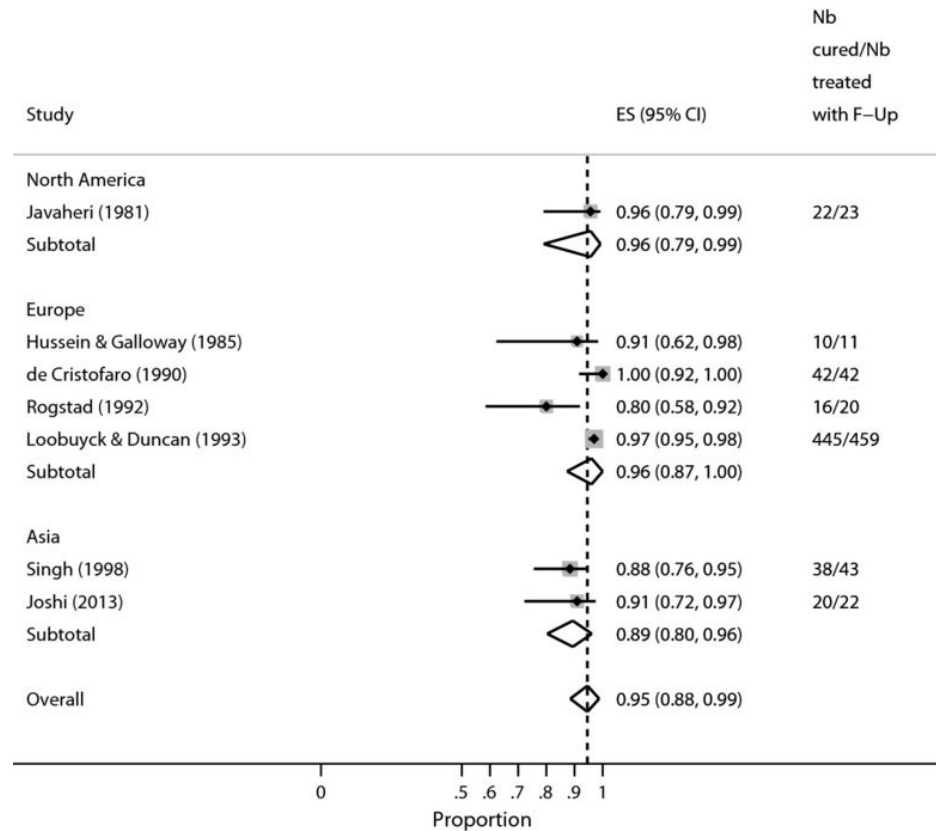


Meta análise – prevalência / incidência



- Prevalência – varia de 0 a 1
- $V_{\hat{p}} = \text{Var}(\hat{p}) = \frac{p(1-p)}{N}$
- Distribuição binomial

Ao usar meta ou metan, assimetria no IC



metaprop

