

# STATISTICS 536B, Lecture #5

March 10, 2015

# Meta-Analysis: Synthesizing evidence from different studies of the same relationship

$y_i$ : Estimated 'effect size' from  $i$ -th study.

$\sigma_i$ : standard error for this effect size.

# A typical abstract...

Med Sci Monit. 2002 Aug;8(8):CR558-65.

Meta-analysis of 20 case-control studies on the N-acetyltransferase 2 acetylation status and colorectal cancer risk.

## Abstract

**BACKGROUND:** Rapid NAT2 acetylation has been considered as a risk factor for developing colon cancer in a number of studies, however the overall results of such studies are inconsistent. To clarify the influence of NAT2 rapid acetylation status on colon cancer risk, we have performed a meta-analysis of 20 published case-control studies (4431 cases, 4547 controls).

## MATERIAL/METHODS:...

**RESULTS:** The pooling of studies based on phenotyping methods indicated that the overall odds ratio of colon cancer risk associated with rapid acetylator was 1.51 (95%CI: 1.07-2.12). However, the risk of colon cancer associated with rapid acetylator from the studies based on genotyping method was lower with a calculated overall odds ratio of 1.06 (95%CI: 0.971.15)...

# The data

```
> library(mmeta)
> data(colorectal)
> colorectal
```

	y1	y2	n1	n2	studynames
1	10	27	41	49	Ilett
2	19	27	45	49	Ilett1
3	13	23	41	43	Wohlleb
4	40	49	96	109	Ladero
5	13	20	28	44	Rodriguez
6	92	14	205	34	Lang
7	33	33	36	36	Oda
8	151	112	329	234	Shibuta
9	50	96	112	202	Bell
10	34	32	96	103	Spurr
11	140	100	343	275	Hubbard
12	74	73	174	174	Welfare
13	68	44	201	114	Gil
...					
19	119	60	258	120	Agundez
20	162	156	209	200	Butler

E.g., one of the 20 studies

	cases	controls
X=1	23	13
X=0	20	28

## Fixed effect meta-analysis

$$Y_i \sim N(\theta, \sigma_i^2), i = 1, \dots, m.$$

## Random effect meta-analysis

$$Y_i | \theta_i \sim N(\theta_i, \sigma_i^2), \quad i = 1, \dots, m.$$

$$\theta_i \sim N(\mu, \tau^2)$$

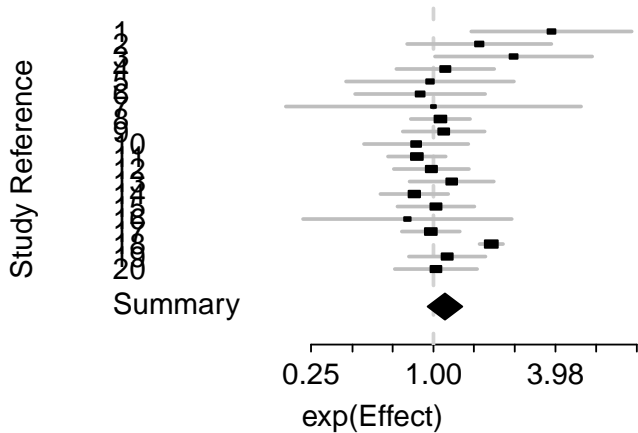




# Implement random effect meta analysis

```
y <- log(colo$y2/(colo$n2-colo$y2))-  
      log(colo$y1/(colo$n1-colo$y1))  
  
sg <- sqrt(1/colo$y1 + 1/(colo$n1-colo$y1) + 1/colo$y2 + 1/(colo$n2-colo$y2))  
  
library(rmeta)  
  
> meta.summaries(y,sg, method="random")  
Random-effects meta-analysis  
Call: meta.summaries(d = y, se = sg, method = "random")  
Summary effect=0.13 95% CI (-0.0661, 0.327)  
Estimated heterogeneity variance: 0.12 p= 0  
  
> plot(meta.summaries(y,sg, method="random"),log=T)
```

# Forest plot



# Quick simulation

```
m.orig <- 100

##### variety of sample sizes
n <- round(1/runif(m.orig, sqrt(1/8000), sqrt(1/100))^2)

##### distribution of actual effects
theta <- rnorm(m.orig, mean=log(1.3), sd=((log(1.3)-log(.9))/2) )

##### simulate the studies
y <- sg <- rep(NA, m.orig); pub <- rep(T, m.orig)
for (i in 1:m.orig) {
  ### simulate exposure status for controls and cases
  tmp.cn <- rbinom(1, size=n[i], prob=.25)
  tmp.cs <- rbinom(1, size=n[i], prob=expit(logit(.25)+theta[i]))
  ### effect estimate
  y[i] <- log(tmp.cs/(n[i]-tmp.cs)) - log(tmp.cn/(n[i]-tmp.cn))
  ### SE
  sg[i] <- sqrt(1/tmp.cs + 1/(n[i]-tmp.cs) + 1/tmp.cn + 1/(n[i]-tmp.cn))
  ### gets published ?
  if (abs(y[i])<(2*sg[i])) { if (runif(1)<.67) {pub[i] <- F} }
}
```

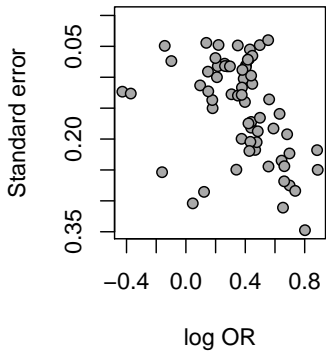
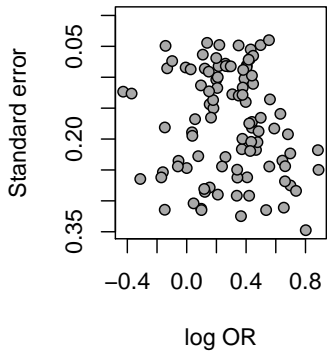
```
> sum(pub)
```

```
[1] 66
```

```
> library("meta")
```

```
> funnel(y, sg, xlab="log OR")
```

```
> funnel(y[pub], sg[pub], xlab="log OR")
```



```
> library(rmeta)
```

```
> meta.summaries(y, sg, method="fixed")
```

```
Fixed-effects meta-analysis
```

```
Call: meta.summaries(d = y, se = sg, method = "fixed")
```

```
Summary effect=0.271 95% CI (0.25, 0.292)
```

```
Estimated heterogeneity variance: 0.039 p= 0
```

```
> meta.summaries(y, sg, method="random")
```

```
Random-effects meta-analysis
```

```
Call: meta.summaries(d = y, se = sg, method = "random")
```

```
Summary effect=0.276 95% CI (0.226, 0.326)
```

```
Estimated heterogeneity variance: 0.039 p= 0
```

```
> meta.summaries(y[pub], sg[pub], method="fixed")
```

```
Fixed-effects meta-analysis
```

```
Call: meta.summaries(d = y[pub], se = sg[pub], method = "fixed")
```

```
Summary effect=0.309 95% CI (0.286, 0.332)
```

```
Estimated heterogeneity variance: 0.04 p= 0
```

```
> meta.summaries(y[pub], sg[pub], method="random")
```

```
Random-effects meta-analysis
```

```
Call: meta.summaries(d = y[pub], se = sg[pub], method = "random")
```

```
Summary effect=0.342 95% CI (0.282, 0.401)
```

```
Estimated heterogeneity variance: 0.04 p= 0
```