

# Pflege Deine Vorurteile!

*A statistician is a person  
who draws a mathematically precise line  
from an unwarranted assumption  
to a foregone conclusion.*

# Statistik als wissenschaftliche Qualitätskontrolle

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**Quo vadis ?**



# Statistik als wissenschaftliche Qualitätskontrolle

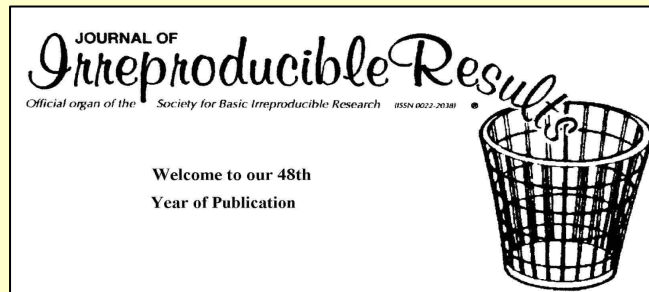
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Quo vadis ?

Science

PNAS  
Proceedings of the National Academy of Sciences  
of the United States of America

nature

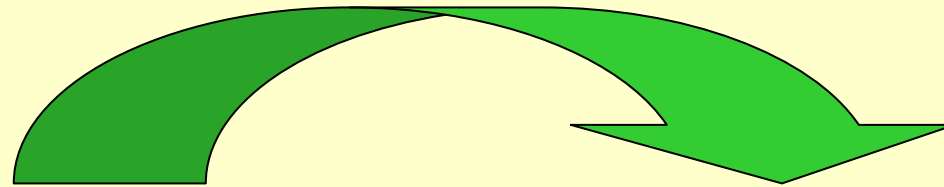


1. Ein p-Wert ist ein p-Wert ist ein p-Wert  
(und kein Test)

# Zwei Arten logischen Schlußfolgerns

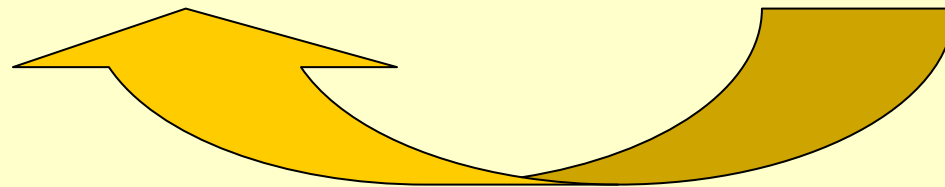
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*deduktiv*



Ursache

Wirkung

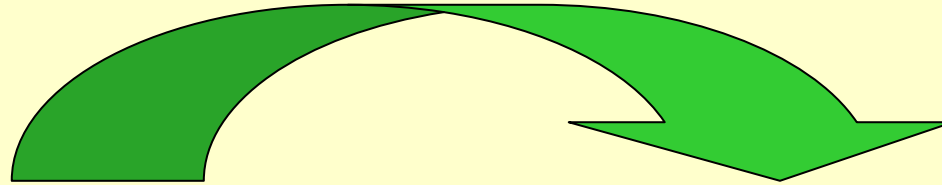


*induktiv*

# Schlußfolgern in der Statistik

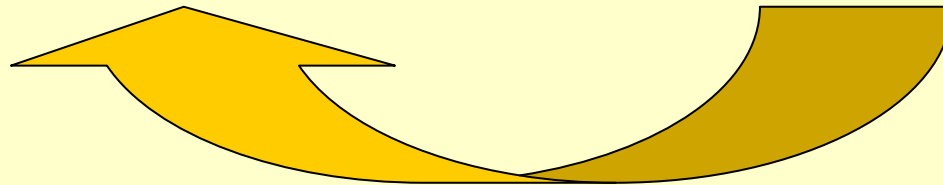
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?



Hypothese

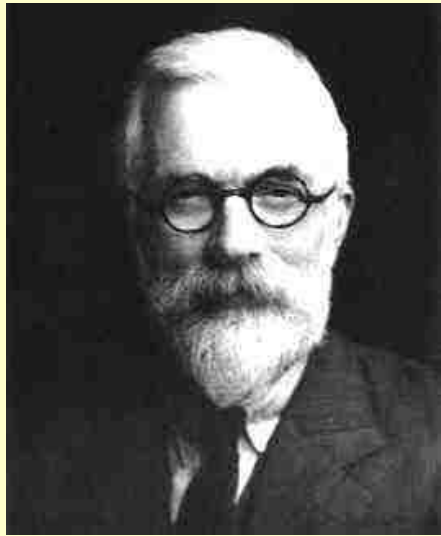
Beobachtung



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# Schlußfolgern in der Statistik (induktiv)

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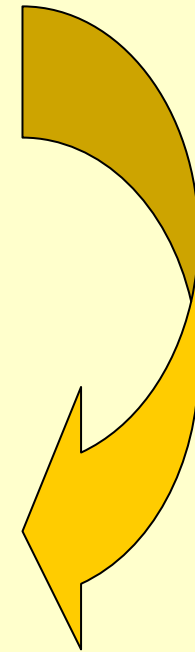


Sir Ronald A. Fisher  
(1890-1962)

Beobachtung

*p-Wert*

Hypothese



# Der $p$ -Wert

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Hypothese	$H_0$
Beobachtung	$x_{\text{obs}}$
Teststatistik	$T: x \rightarrow T(x)$

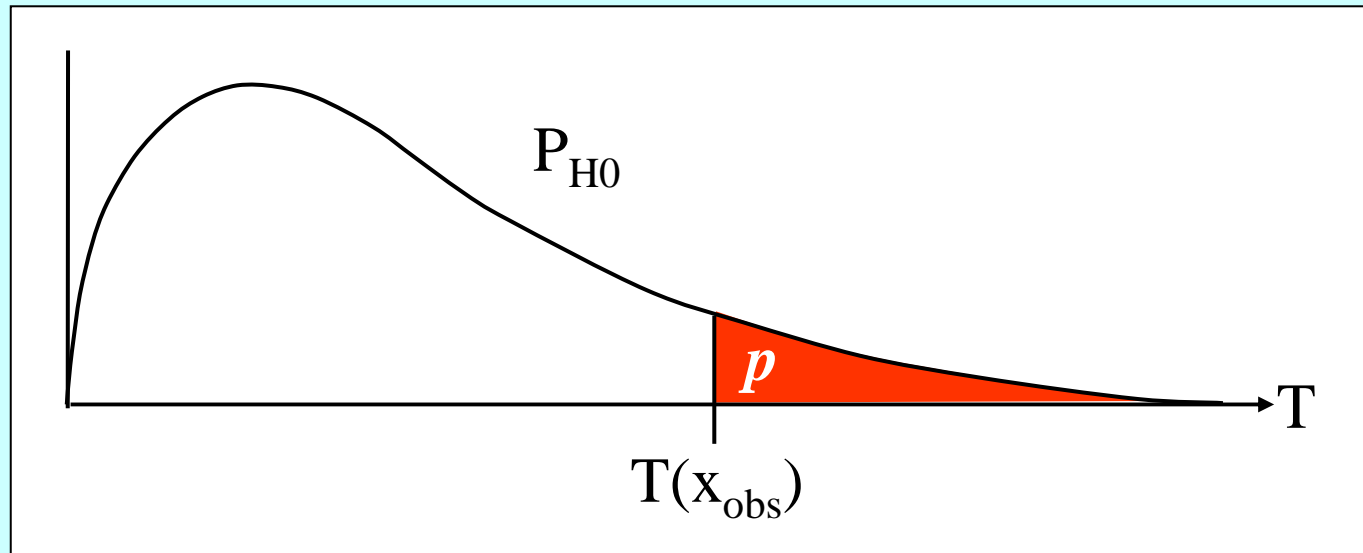
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$$p = P_{H_0}(x: T(x) \geq T(x_{\text{obs}}))$$



# Der $p$ -Wert

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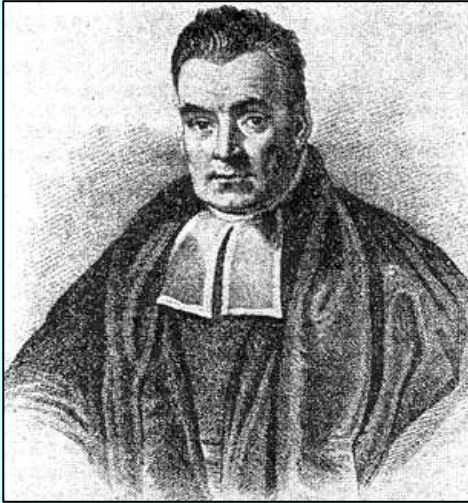


“... an informal index to be used as a measure of discrepancy between the data and the null hypothesis.”

Goodman SN (1999) Ann Intern Med 130: 995-1004

# Das Theorem von Bayes

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$$P(H_0|x_{\text{obs}}) \propto P(H_0) \cdot p$$

Thomas Bayes (1702-1761)

“No test based upon the theory of probability can by itself provide any valuable evidence of the truth or falsehood of a hypothesis”

Neyman J, Pearson E (1933) Phil Trans R Soc A, 231:289-337

# $p$ -Wert als “Entscheidungskriterium“

---

$$p=0.10$$

unerwartet → “shows that the two groups are equivalent“

erwartet → “trend of borderline significance“

“not statistically significant, most probably  
because of small sample size“

# $p$ -Wert als “Entscheidungskriterium“

---

$$p=0.10$$

unerwartet → “shows that the two groups are equivalent“

erwartet → “trend of borderline significance“

“not statistically significant, most probably because of small sample size“

$$p=0.01$$

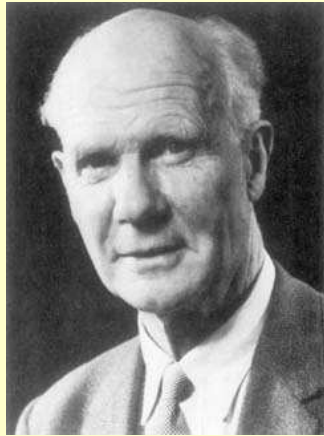
unerwartet → “in all likelihood represents a false positive“

“reflects unknown bias“

erwartet → “clearly demonstrates a treatment effect“

# Schlußfolgern in der Statistik (deduktiv)

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Egon Pearson (1895-1980)



Jerzy Neyman (1894-1981)

Hypothese

*statistisches Testen*

Beobachtung



# Statistisches Testen

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“Without hoping to know whether each separate hypothesis is true or false, we may search for rules to govern our behavior with regard to them, in following which we insure that, in the long run of experience, we shall not often be wrong.”

Neyman J, Pearson E (1933) *Phil Trans R Soc A*, 231:289-337

# Statistisches Testen

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Hypothesen	$H_0, H_1$
Beobachtungen	$x$
Teststatistik	$T: x \rightarrow T(x)$
Fehler 1. Art	$\alpha$
Fehler 2. Art	$\beta$

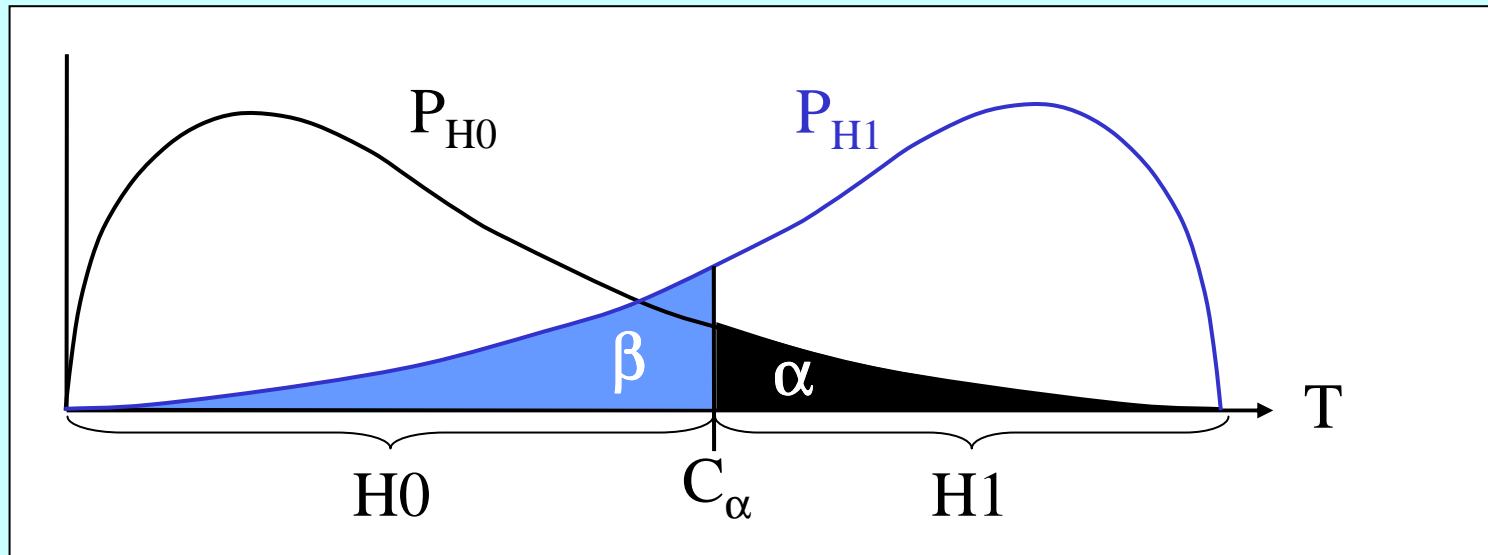
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Wähle  $C_\alpha$  so, daß  $P_{H_0}(x: T(x) > C_\alpha) \leq \alpha$

$$P_{H_1}(x: T(x) \leq C_\alpha) \leq \beta$$

# Statistisches Testen

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$$T(x_{\text{obs}}) \leq C_\alpha \longrightarrow H0$$

$$T(x_{\text{obs}}) > C_\alpha \longrightarrow H1$$



# Das sogenannte „multiple Testproblem“

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	$p$
Placebo <i>vs.</i> Behandlung A ●	0.125
Placebo <i>vs.</i> Behandlung B ●	0.015
Placebo <i>vs.</i> Behandlung C ●	0.045

# Das sogenannte „multiple Testproblem“

---

		$p$	$p_{crit}$	
Placebo vs. Behandlung A	●	0.125	} $\frac{0.05}{3}$	●
Placebo vs. Behandlung B	●	0.015		●
Placebo vs. Behandlung C	●	0.045		●

# Das sogenannte „multiple Testproblem“

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		$p$	$p_{crit}$	
Placebo vs. Behandlung A	●	0.125	} $\frac{0.05}{3}$	●
Placebo vs. Behandlung B	●	0.015		●
Placebo vs. Behandlung C	●	0.045		●
Placebo vs. Behandlung D	●	0.020	} $\frac{0.05}{2}$	●
Placebo vs. Behandlung E	●	0.005		●

# Das sogenannte „multiple Testproblem“

		$p$	$p_{crit}$		$p_{crit}$
Placebo vs. Behandlung A	●	0.125	} $\frac{0.05}{3}$	●	} $\frac{0.05}{5}$
Placebo vs. Behandlung B	●	0.015		●	
Placebo vs. Behandlung C	●	0.045		●	
Placebo vs. Behandlung D	●	0.020	} $\frac{0.05}{2}$	●	
Placebo vs. Behandlung E	●	0.005		●	

# Das sogenannte „multiple Testproblem“

		$p$	$p_{crit}$		$p_{crit}$
Placebo vs. Behandlung A	●	0.125	} $\frac{0.05}{3}$	●	} $\frac{0.05}{5}$
Placebo vs. Behandlung B	●	0.015		●	
Placebo vs. Behandlung C	●	0.045		●	
Placebo vs. Behandlung D	●	0.020	} $\frac{0.05}{2}$	●	
Placebo vs. Behandlung E	●	0.005		●	

entweder

$H_{0A}$	<del><math>H_{0B}</math></del>	<del><math>H_{0C}</math></del>	<del><math>H_{0D}</math></del>	<del><math>H_{0E}</math></del>
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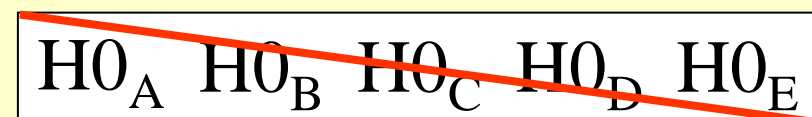
# Das sogenannte „multiple Testproblem“

		$p$	$p_{crit}$		$p_{crit}$
Placebo vs. Behandlung A	●	0.125	} $\frac{0.05}{3}$	●	} $\frac{0.05}{5}$
Placebo vs. Behandlung B	●	0.015		●	
Placebo vs. Behandlung C	●	0.045		●	
Placebo vs. Behandlung D	●	0.020	} $\frac{0.05}{2}$	●	
Placebo vs. Behandlung E	●	0.005		●	

entweder



oder



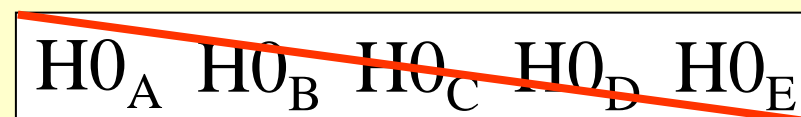
# Das sogenannte „multiple Testproblem“

	$p$	$P_{crit}$	$P_{crit}$
Placebo vs. Behandlung A	0.125	0.05 3	0.05 5
Placebo vs. Behandlung B	0.015		
Placebo vs. Behandlung C	0.045		
Placebo vs. Behandlung D	0.020	0.05 2	0.05 5
Placebo vs. Behandlung E	0.005		

entweder



oder



aber nicht



2. Was ist schon/noch „normal“?



# Verteilung: Fakt und Fiktion

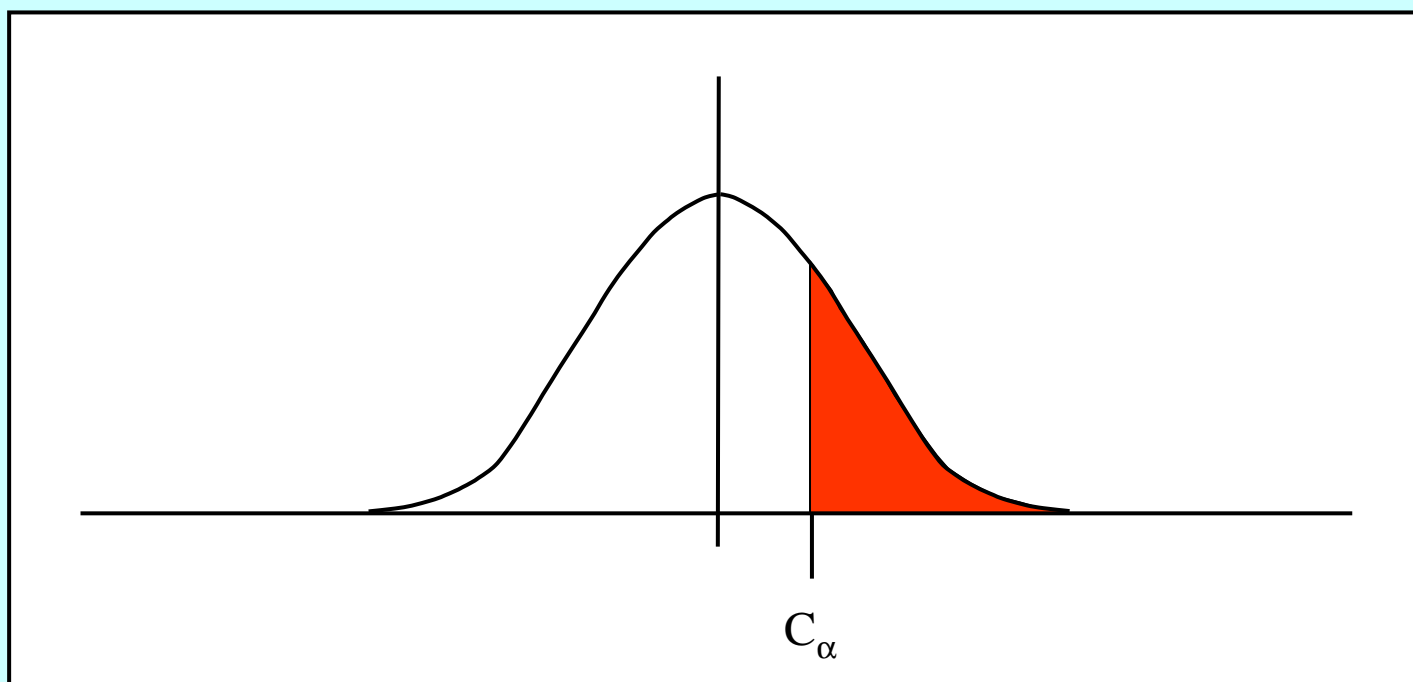
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Test

$$P_{H_0}(x:T(x) > C_\alpha)$$

*p*-Wert

$$P_{H_0}(x:T(x) \geq T(x_{\text{obs}}))$$



# Verteilung: Fakt und Fiktion

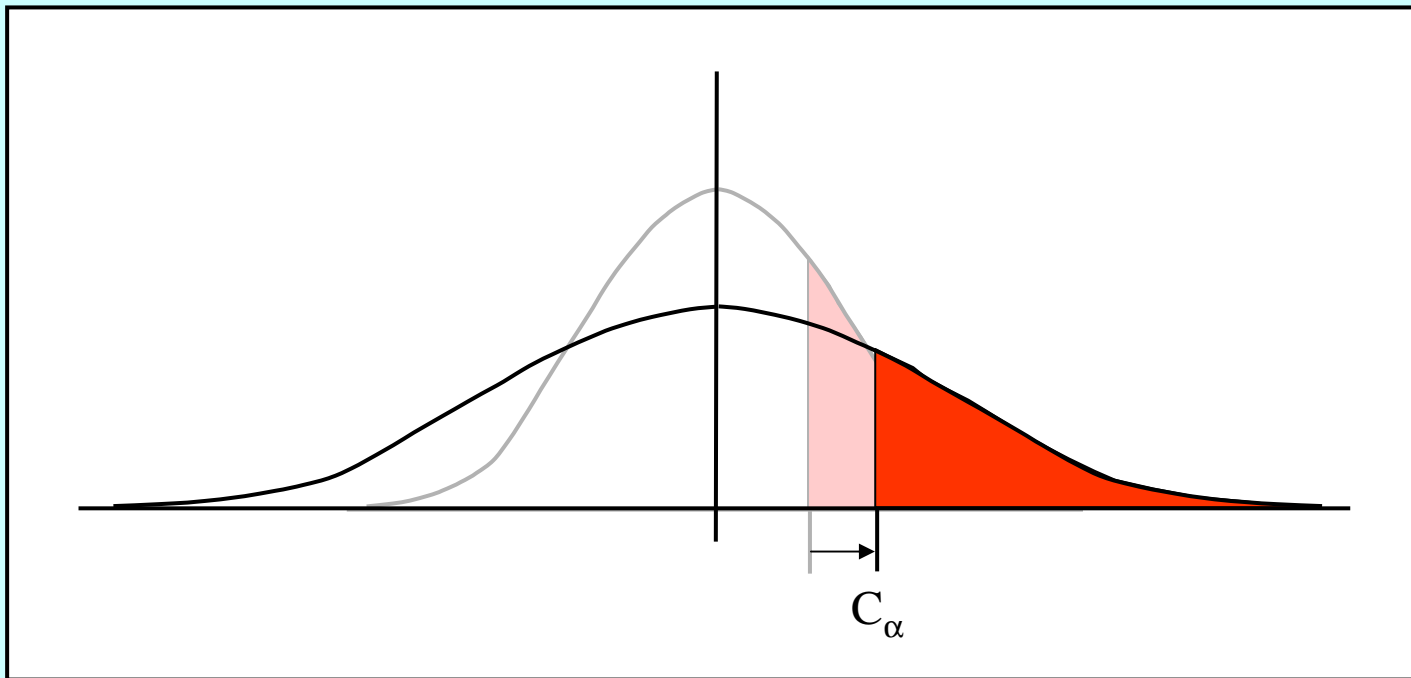
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Test

$$P_{H_0}(x:T(x) > C_\alpha)$$

*p*-Wert

$$P_{H_0}(x:T(x) \geq T(x_{\text{obs}}))$$



# Randomisierungs-Tests

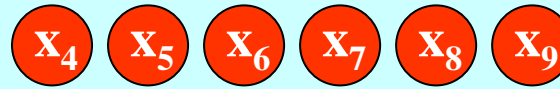
„statistics without tears“

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Fälle

Kontrollen

$X_{\text{obs}}$



$T(x_{\text{obs}})$

# Randomisierungs-Tests

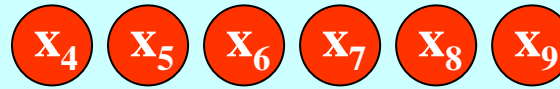
„statistics without tears“

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Fälle

Kontrollen

$X_{\text{obs}}$



$T(x_{\text{obs}})$

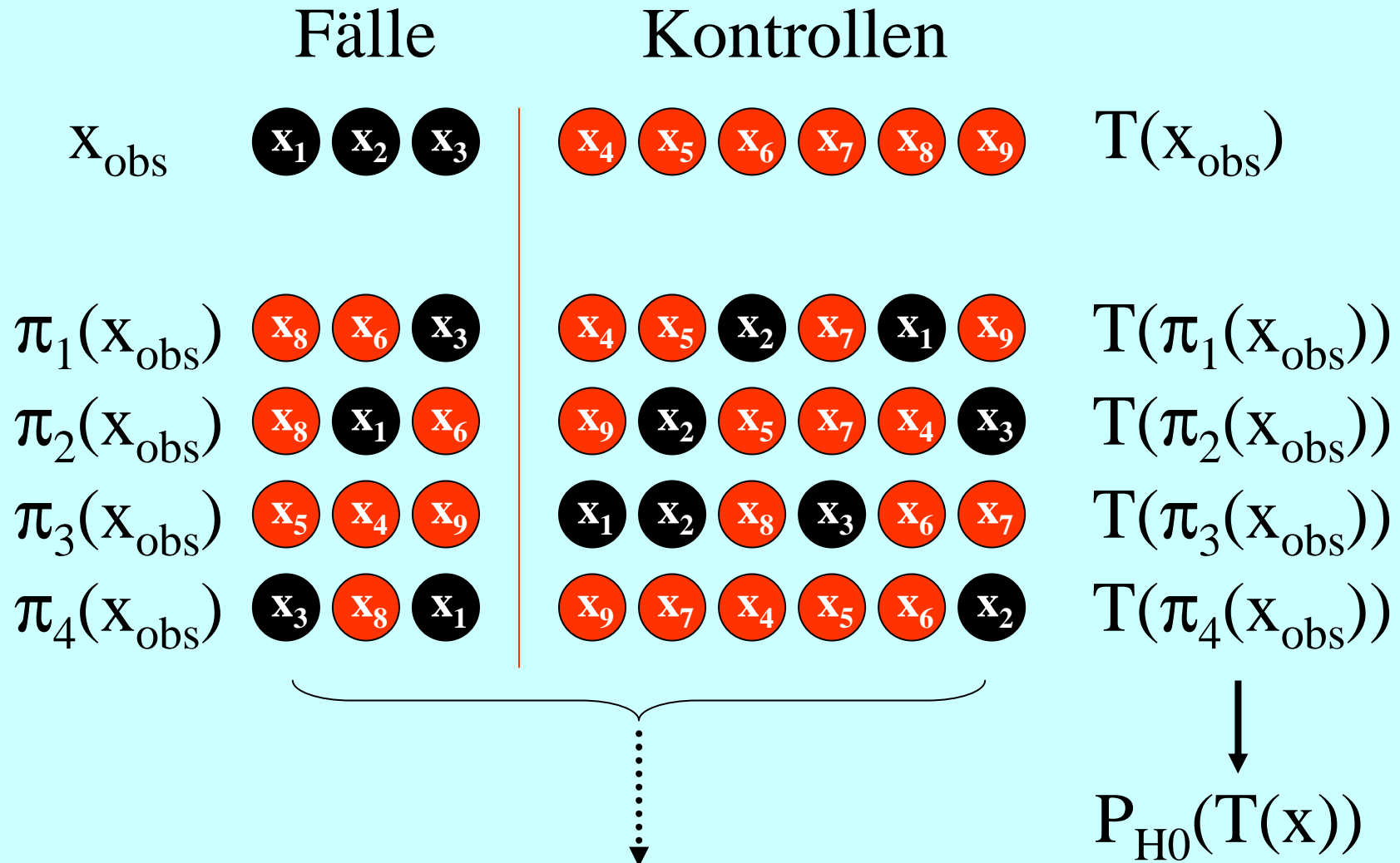
$\pi_1(X_{\text{obs}})$



$T(\pi_1(x_{\text{obs}}))$

# Randomisierungs-Tests

„statistics without tears“



### 3. Ein p-Wert mißt keine Effektgröße (signifikant ist nicht gleich „signifikant“)

# Kleine Studie, großer Effekt ...

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	Erfolg	ØErfolg	$\Sigma$
Verum	40	10	50
Placebo	25	25	50
$\Sigma$	65	35	100

$\chi^2=8.62$ , 1 df,  $p=0.004$

OR=4.000 CI: 1.517-10.749

# Kleine Studie, großer Effekt ...

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	Erfolg	ØErfolg	Σ
Verum	40	10	50
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Σ	65	35	100

$\chi^2=8.62$ , 1 df,  $p=0.004$

OR=4.000 CI: 1.517-10.749

	Erfolg	ØErfolg	Σ
Verum	2648	2352	5000
Placebo	2500	2500	5000
Σ	5148	4852	10000

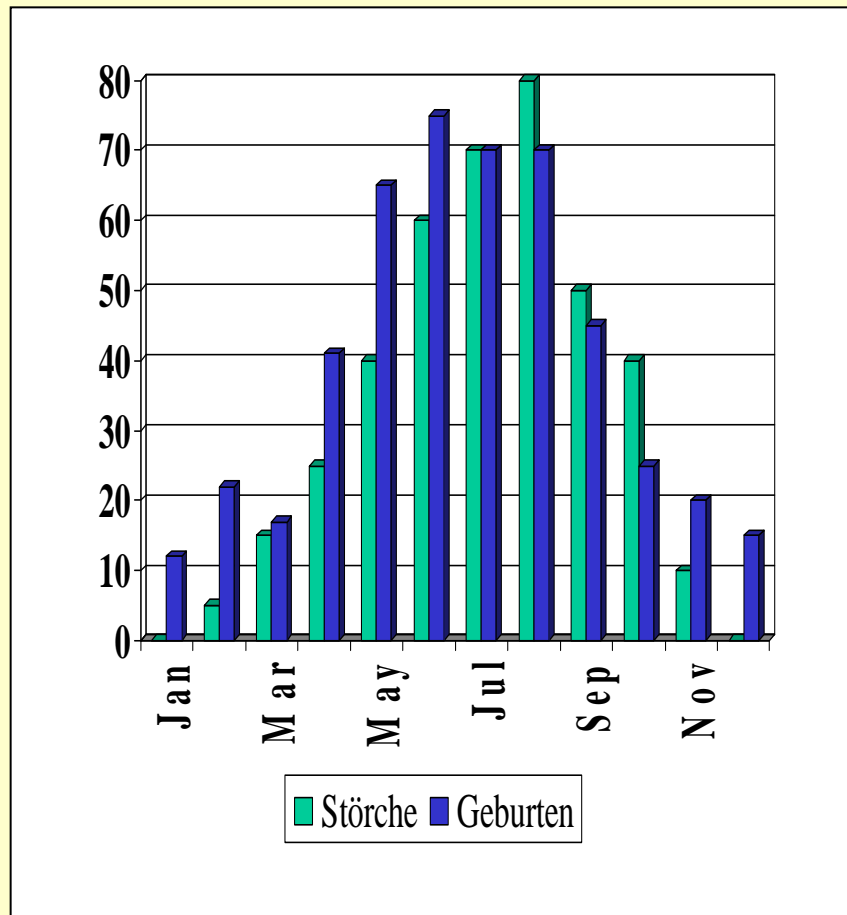
$\chi^2=8.62$ , 1 df,  $p=0.004$

OR=1.126 CI: 1.040-1.219



4. *post hoc ergo propter hoc*  
(die Sache mit dem Klapperstorch)

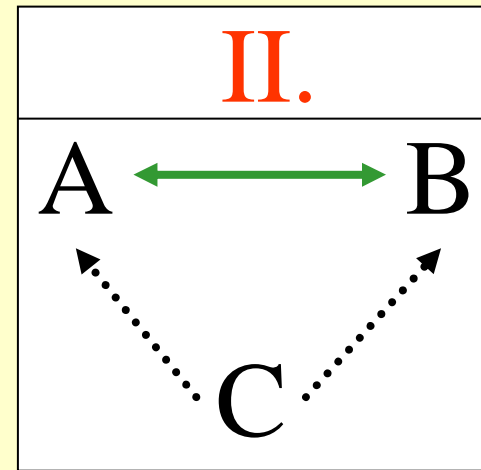
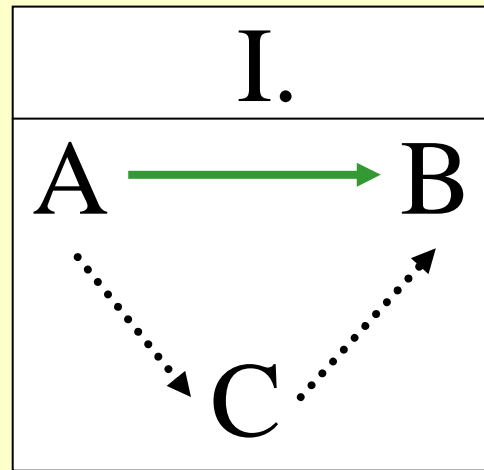
# Von Störchen und Babys



$$r = 0.898$$

# Scheinkorrelation/assoziation

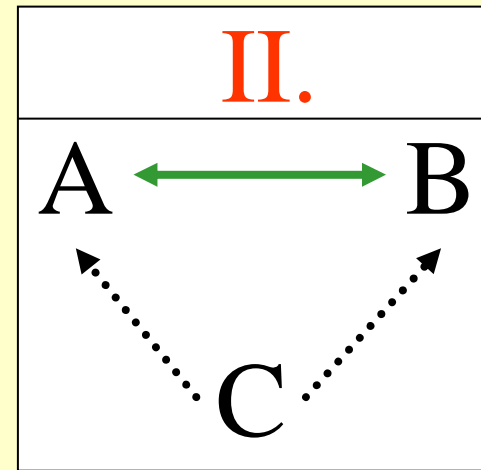
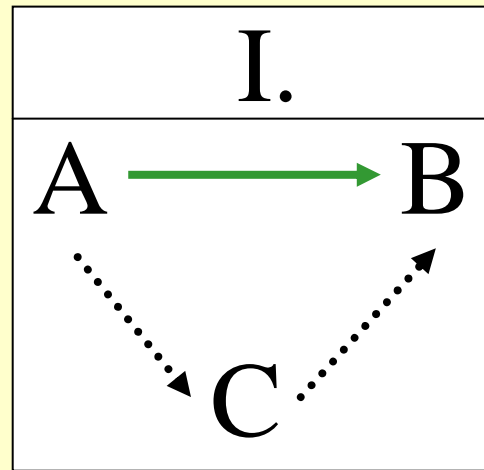
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A: Geschlecht, B: Verhalten, C: Erziehung

# Scheinkorrelation/assoziation

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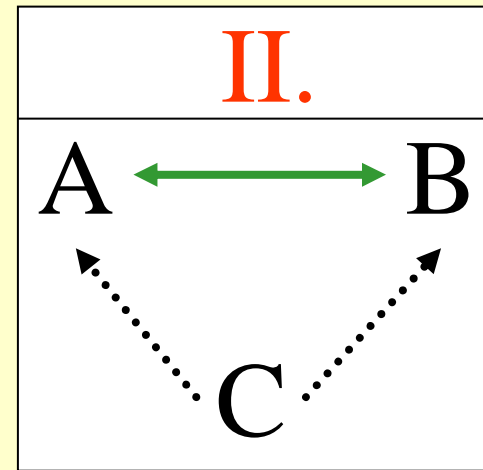
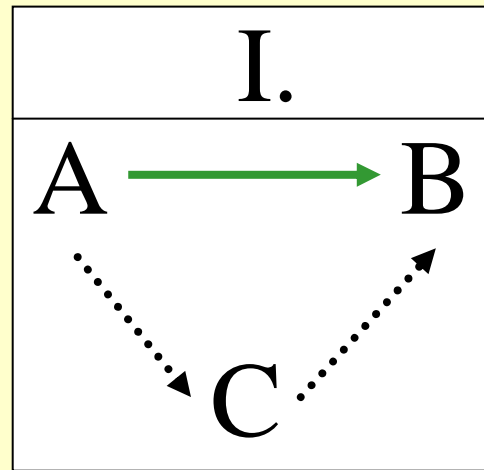


A: Geschlecht, B: Verhalten, C: Erziehung

A: Therapieform, B: Morbidität, C: Mobilität

# Scheinkorrelation/assoziation

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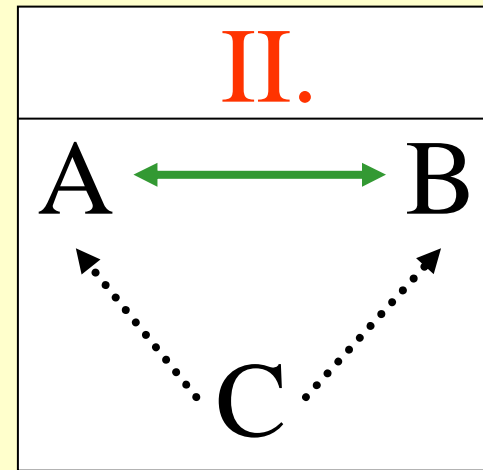
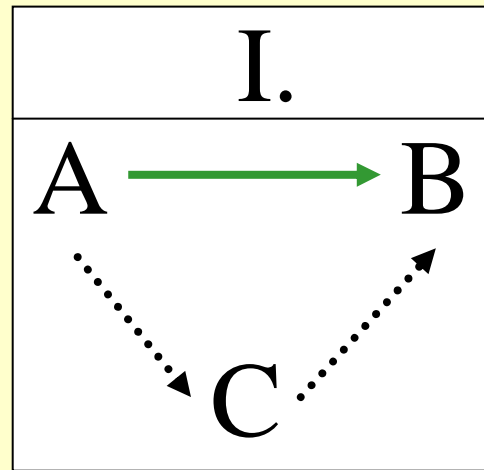
A: Geschlecht, B: Verhalten, C: Erziehung

A: Therapieform, B: Morbidität, C: Mobilität

A: Ernährung, B: Lebensdauer, C: Sozialisation

# Scheinkorrelation/assoziation

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A: Geschlecht, B: Verhalten, C: Erziehung

A: Therapieform, B: Morbidität, C: Mobilität

A: Ernährung, B: Lebensdauer, C: Sozialisation

A: Mobiltelefonieren, B: Schlafstörungen, C: Lebensweise

# Alle Confounder bedacht?

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# Alle Confounder bedacht?

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# Alle Confounder bedacht?

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... Augen links



# Alle Confounder bedacht?

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... Augen rechts



# „The DDT ban myth“

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Malaria-Prävalenz in Sri Lanka

1948	2,800,000
1958	Beginn des DDT Einsatzes
1962	<i>Silent Spring</i> (Rachel Carson)
1963	17
1964	Verbot von DDT
1968	1,000,000
1969	2,500,000

# „The DDT ban myth“

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Malaria-Prävalenz in Sri Lanka

1948

1958

1962

1963

1964

1968

1969

**Resistenzbildung!**

Beitrag zum DDT Einsatzes

Spring (Rachel Carson)

17

Verbot von DDT

1,000,000

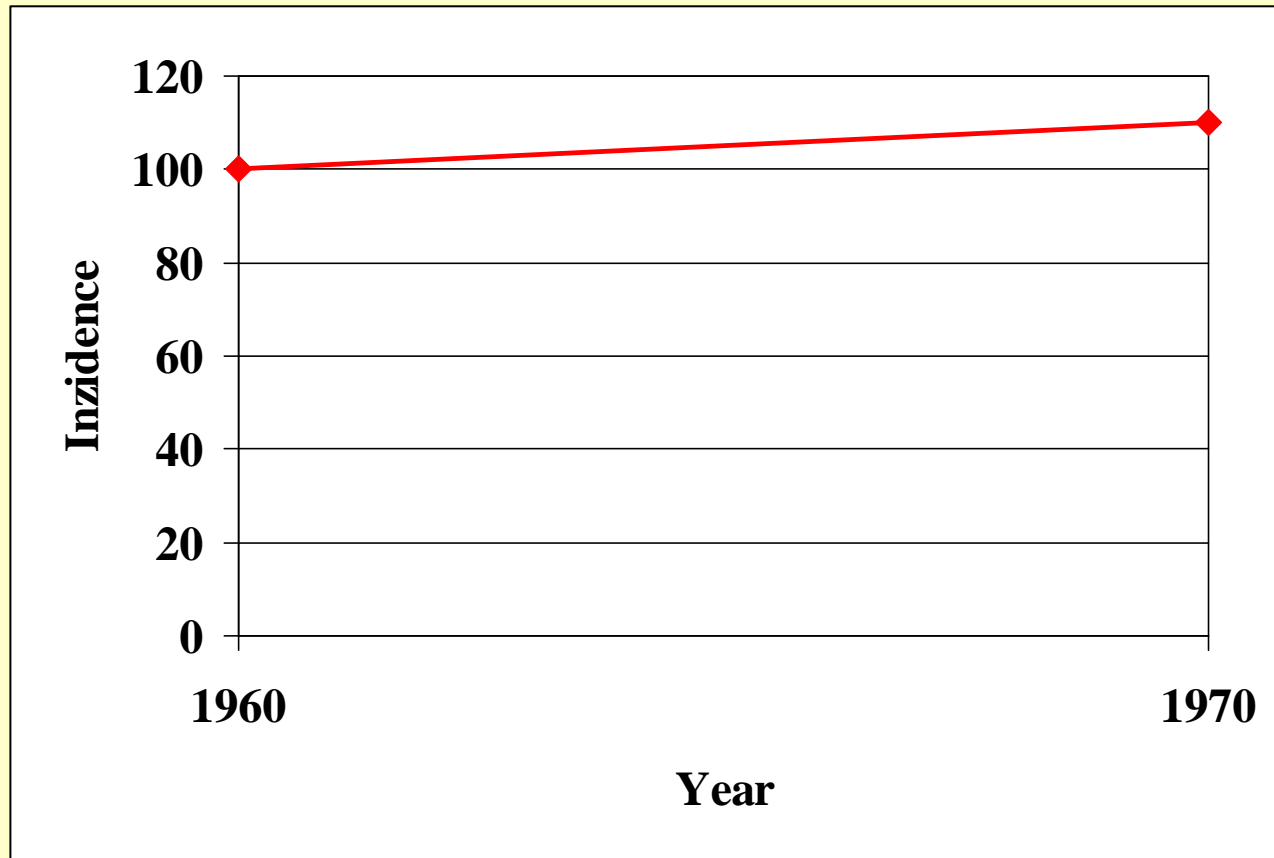
2,500,000

# 5. Skalieren, Normieren, Standardisieren

(Bilder sagen mehr als tausend Worte)

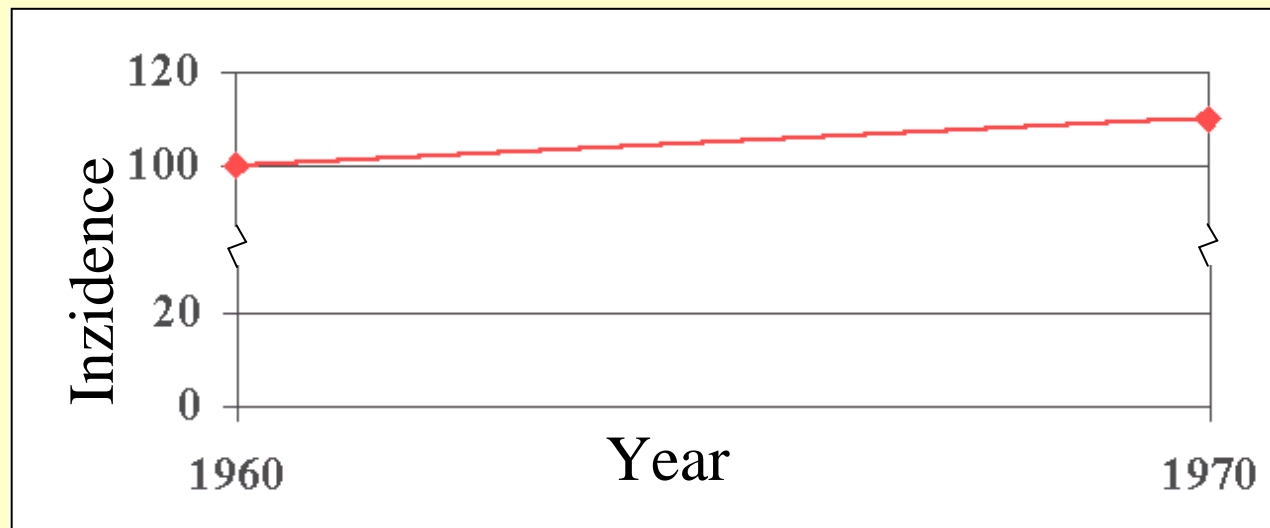
# Schwache Trends, starke Trends

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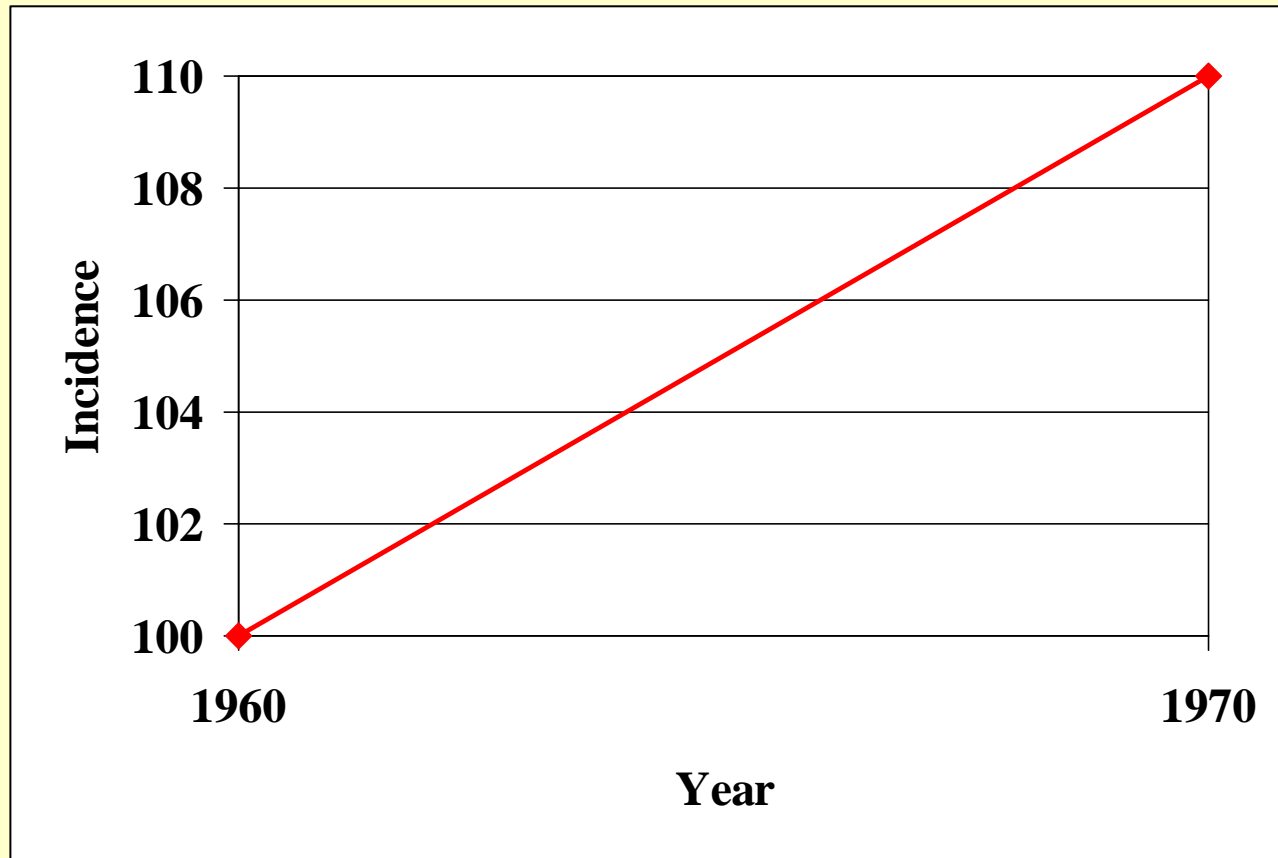
# Schwache Trends, starke Trends

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# Schwache Trends, starke Trends

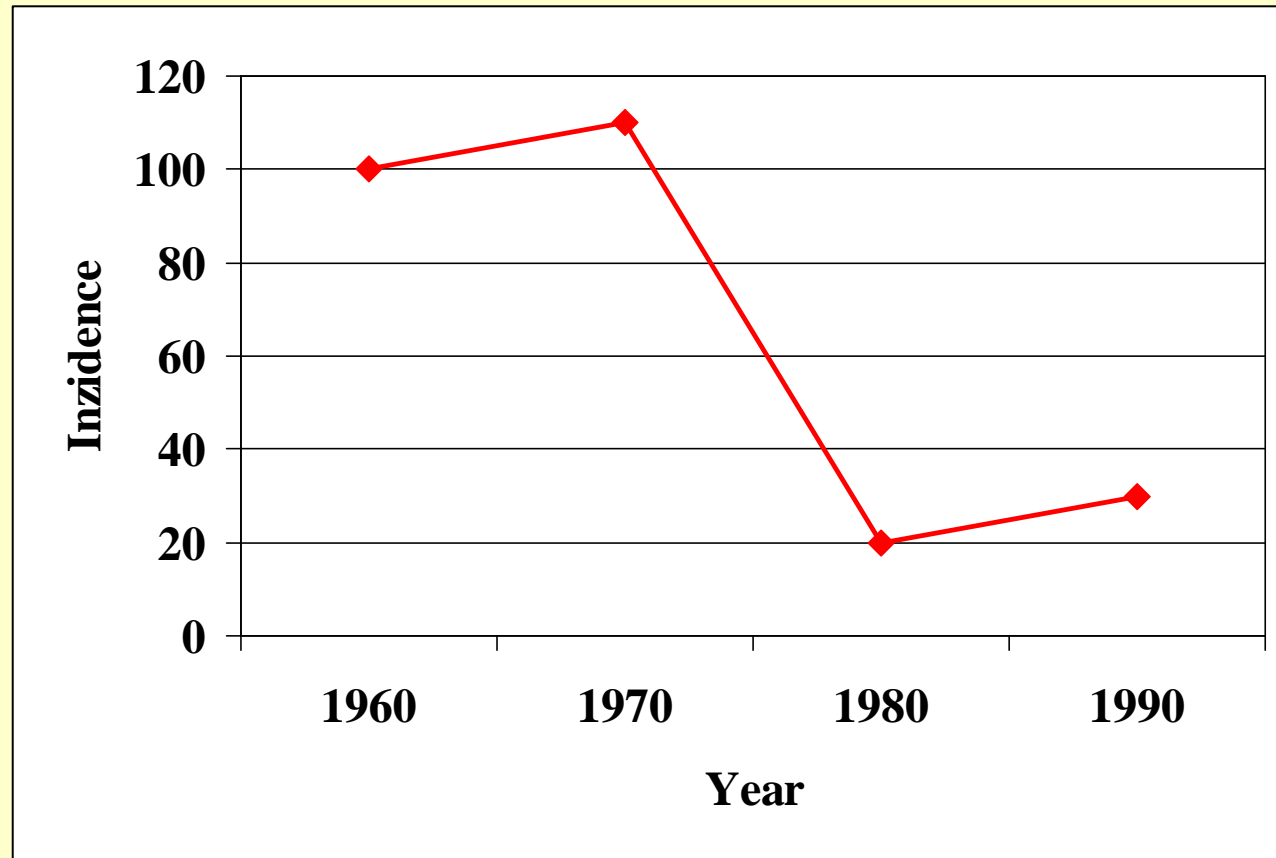
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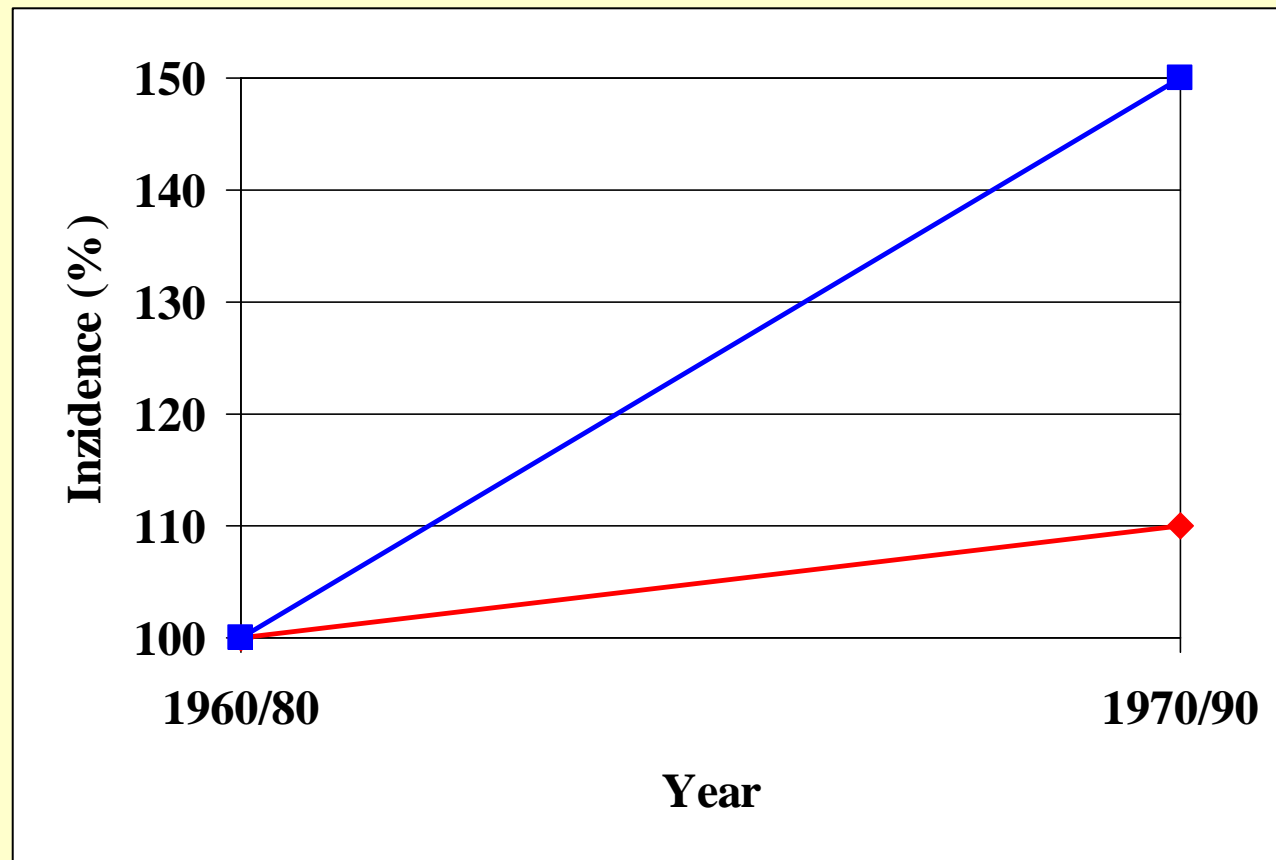
# Schwache Trends, starke Trends

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# Schwache Trends, starke Trends

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# The End

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