

# Lineare Regression: Tests

Statistik (Biol./Pharm./HST) – FS 2015



# Ersatz: Cooper & Shuttle

- 12-Minuten Test nach Cooper (1968)
- 20m-Shuttle-Test nach Leger (1982)

Eur J Appl Physiol (1982) 49: 1–12

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European Journal of  
**Applied  
Physiology**  
and Occupational Physiology  
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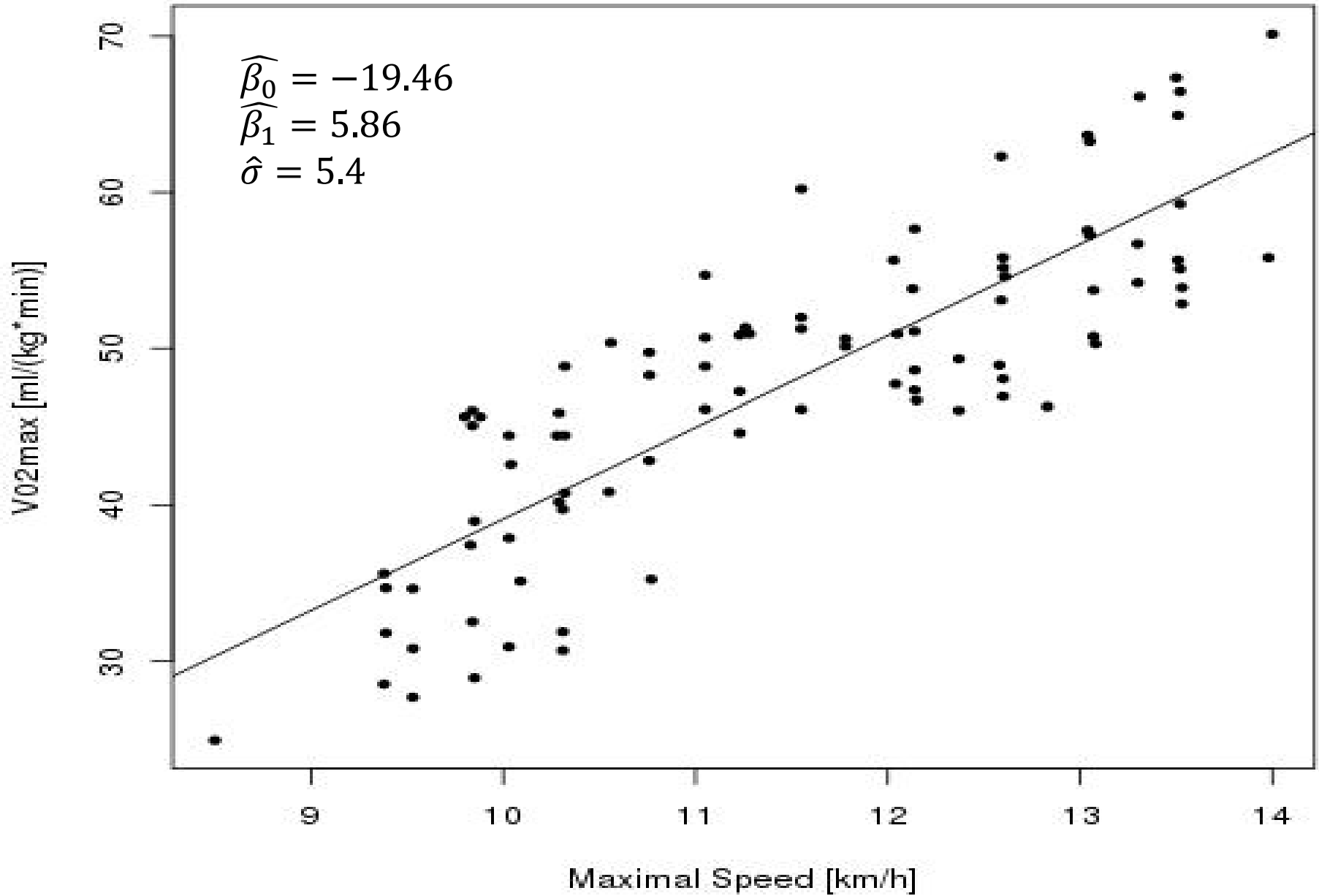
## **A Maximal Multistage 20-m Shuttle Run Test to Predict $\dot{V}O_2 \max^*$**

Luc A. Léger<sup>1</sup> and J. Lambert<sup>2</sup>

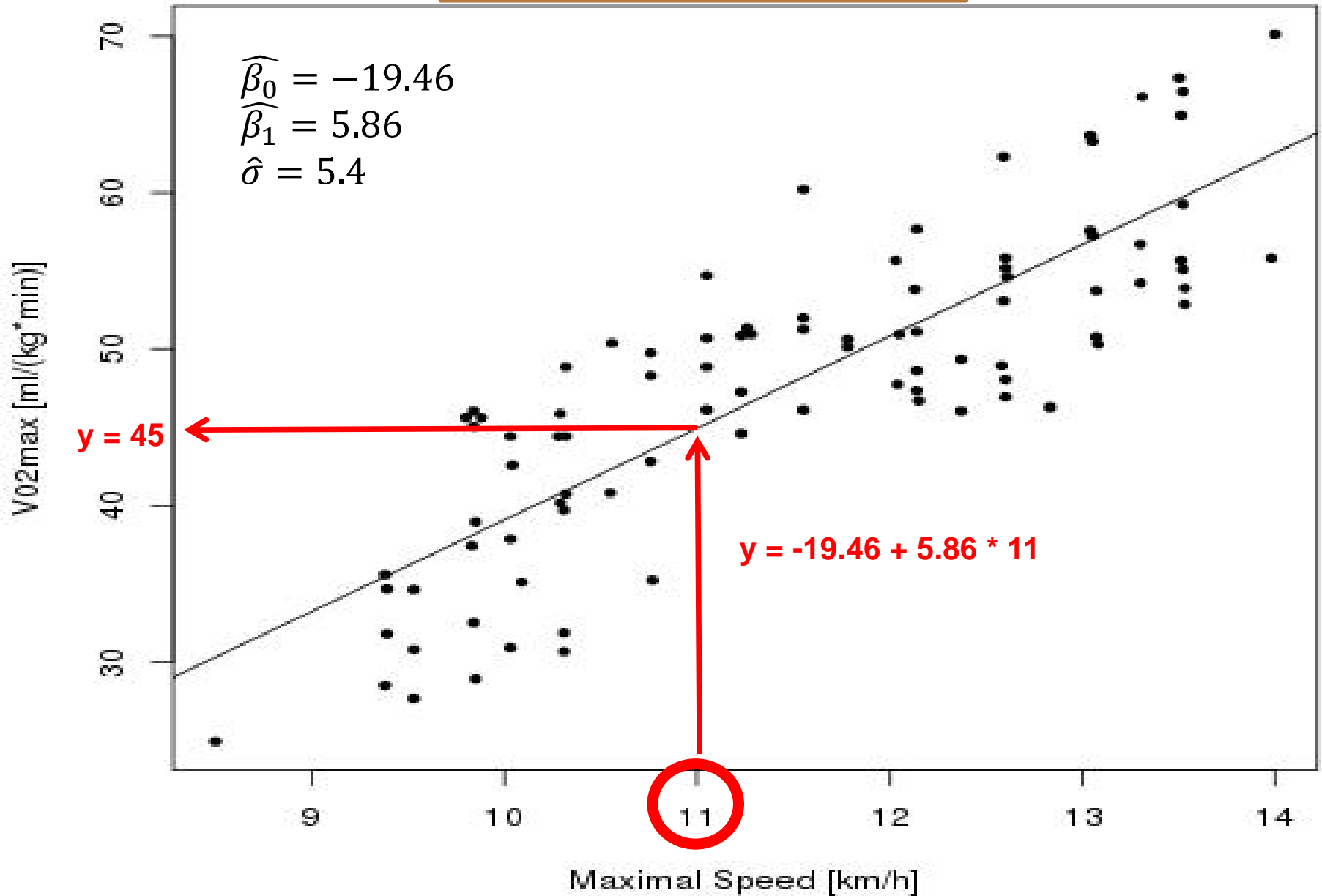
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CEPSUM, C.P. 6128, Succ. "A", Montréal (Québec), Canada, H3C 3J7

<sup>2</sup>Département de Médecine sociale et préventive, Université de Montréal, Canada

# Methode der kleinsten Quadrate



- Wie genau stimmen Parameter?
- Wie genau stimmt Vorhersage?



# t-Test in der Linearen Regression

1. Modell:  $Y_i = \beta_0 + \beta_1 x_i + E_i, E_1, \dots, E_n \text{ iid } N(0, \sigma^2)$
2. Nullhypothese:  $H_0: \beta_1 = 0$   
Alternative:  $H_A: \beta_1 \neq 0$  (es wird normalerweise ein zweiseitiger Test durchgeführt)

3. Teststatistik:

$$T = \frac{\text{beobachtet} - \text{erwartet}}{\text{geschätzter Standardfehler}} = \frac{\widehat{\beta}_1 - 0}{\widehat{s.e.}(\widehat{\beta}_1)}$$

Dabei ist  $s.e.(\widehat{\beta}_1) = \sqrt{\text{Var}(\widehat{\beta}_1)}$  der "Standard Error" von  $\widehat{\beta}_1$

Verteilung der Teststatistik unter  $H_0: T \sim t_{n-2}$

4. Signifikanzniveau:  $\alpha$
5. Verwerfungsbereich der Teststatistik:

$$K = \left(-\infty, -t_{n-2; 1-\frac{\alpha}{2}}\right] \cup \left[t_{n-2; 1-\frac{\alpha}{2}}, \infty\right)$$

6. Testentscheid: Überprüfe, ob der beobachtete Wert der Teststatistik im Verwerfungsbereich liegt.

# Lineare Regression in R

Modell:  $Y_i = \beta_0 + \beta_1 x_i + E_i$ ,  $E_i \sim N(0, \sigma^2)$  i. i. d

Modell:  $Y_i = -19.46 + 5.86x_i + E_i$ ,  $E_i \sim N(0, 5.43^2)$  i. i. d

```
> fitShuttle <- lm(vo2max ~ vmax, data = dat)
> summary(fitShuttle)
```

```
Call:
lm(formula = vo2max ~ vmax, data = dat)
```

```
Residuals:
    Min       1Q   Median       3Q      Max
-10.2230  -4.3976  -0.2016   4.7026  12.0348
```

```
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -19.4582     4.7239  -4.119  8.5e-05 ***
vmax         5.8566     0.4082  14.347 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 5.433 on 89 degrees of freedom
Multiple R-squared: 0.6981, Adjusted R-squared: 0.6948
F-statistic: 205.8 on 1 and 89 DF, p-value: < 2.2e-16
```

Standardfehler von  $\widehat{\beta}_1 (= \widehat{\sigma}_{\widehat{\beta}_1})$

Approx. 95%-VI:

$$5.86 \pm 2 * 0.41$$

Exaktes 95%-VI:

$t_{89;0.975}$

Beobachtete Teststatistik

im Test  $H_0: \beta_1 = 0$  vs.

$H_A: \beta_1 \neq 0$

P-Wert:

Angenommen  $\beta_1 = 0$ ;  
wie wa. ist Beobachtung  
oder etwas extremere?

Freiheitsgrade:  $n - (\text{Anz. } \beta\text{'s}) = 91 - 2 = 89$

# Bsp Prüfungsfrage



Kann  $H_0: \beta_1 = 0$  auf dem 5%-Signifikanzniveau verworfen werden?

Call:

```
lm(formula = y ~ x)
```

Residuals:

Min	1Q	Median	3Q	Max
-3.9552	-1.3273	-0.0089	1.2986	3.5242

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.0289	0.3302	3.116	0.00385	**
x	1.8859	0.2777			

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.925 on 32 degrees of freedom

Multiple R-squared: 0.5904, Adjusted R-squared: 0.5776

F-statistic: 46.13 on 1 and 32 DF, p-value: 1.119e-07

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x	1.8859	0.2777	6.792	1.12e-07	***

---

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Ja:

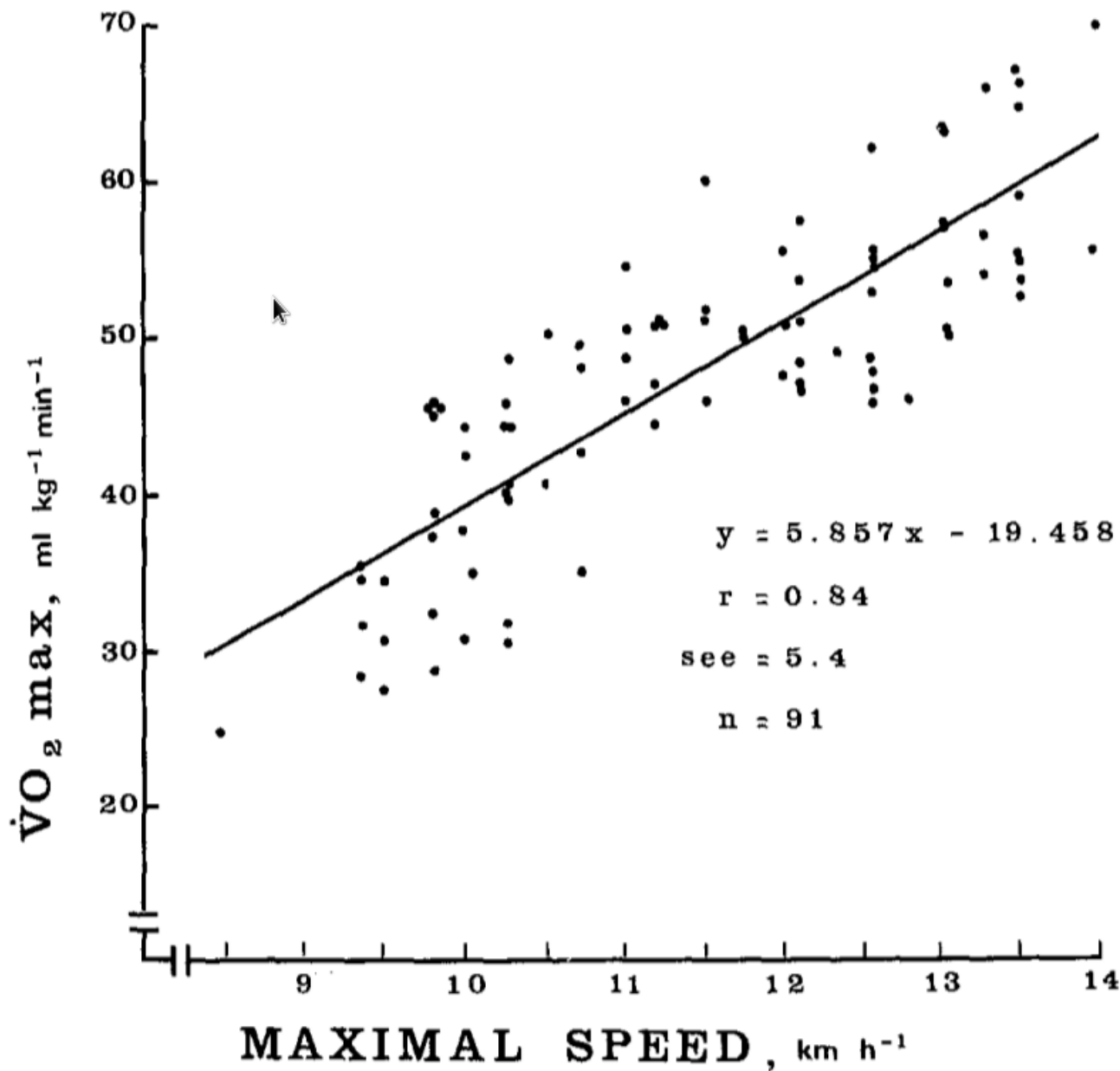
- t value:  $1.88/0.277 = 6.79$

- Verwerfungsbereich:

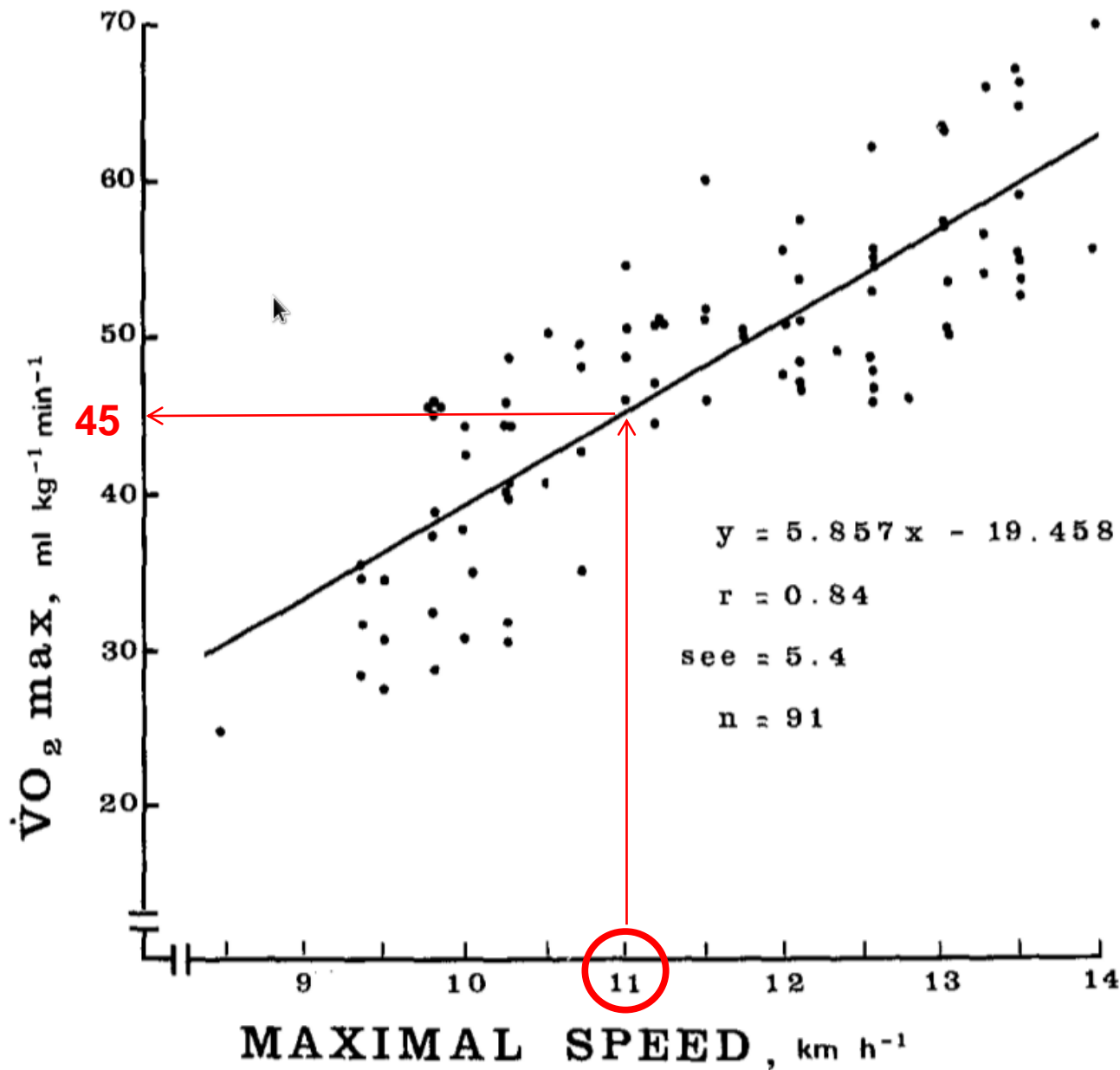
$[-\infty, -t_{32;0.975}] \cup [t_{32;0.975}, \infty]$

$t_{32;0.975} = 2.037$

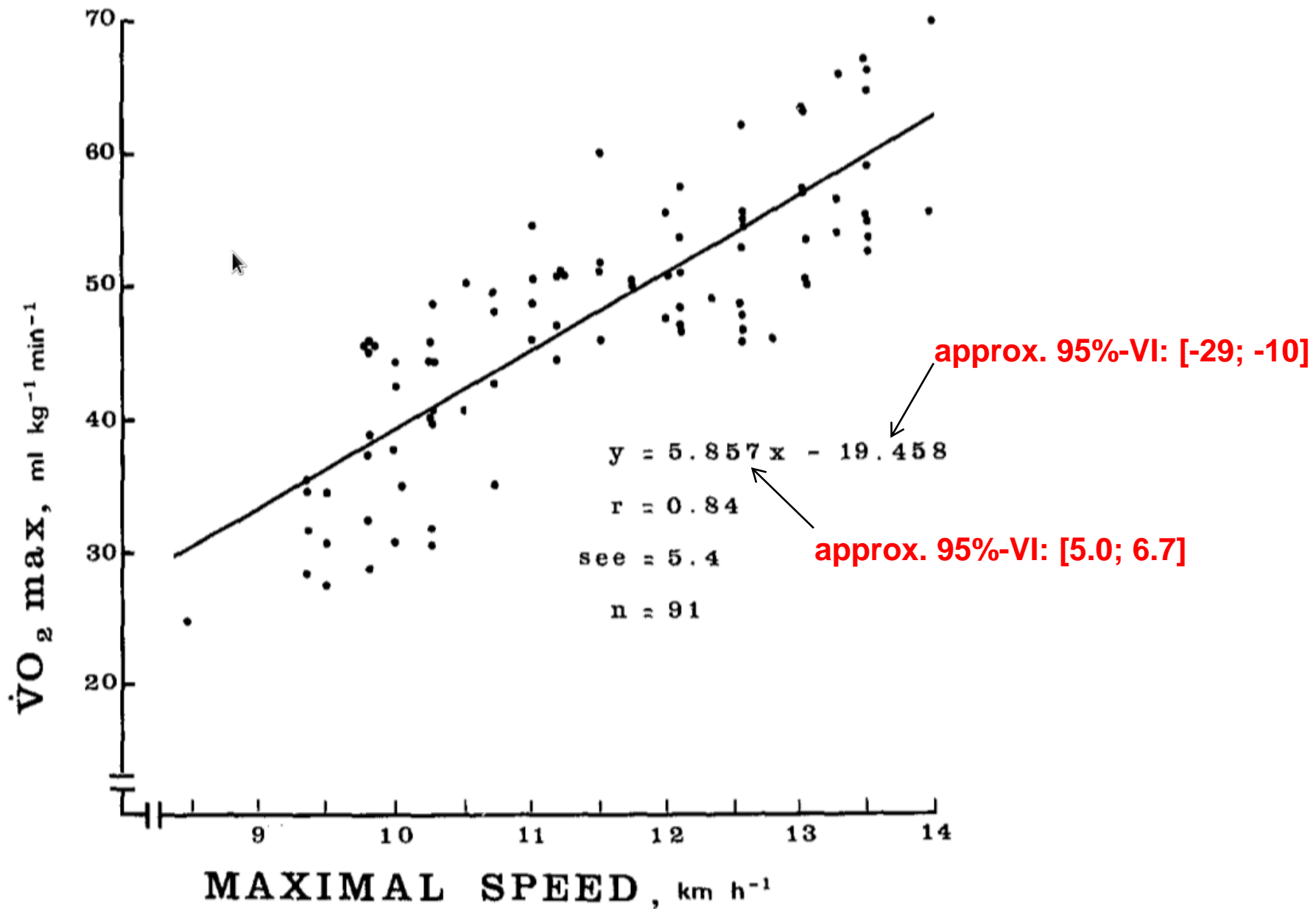




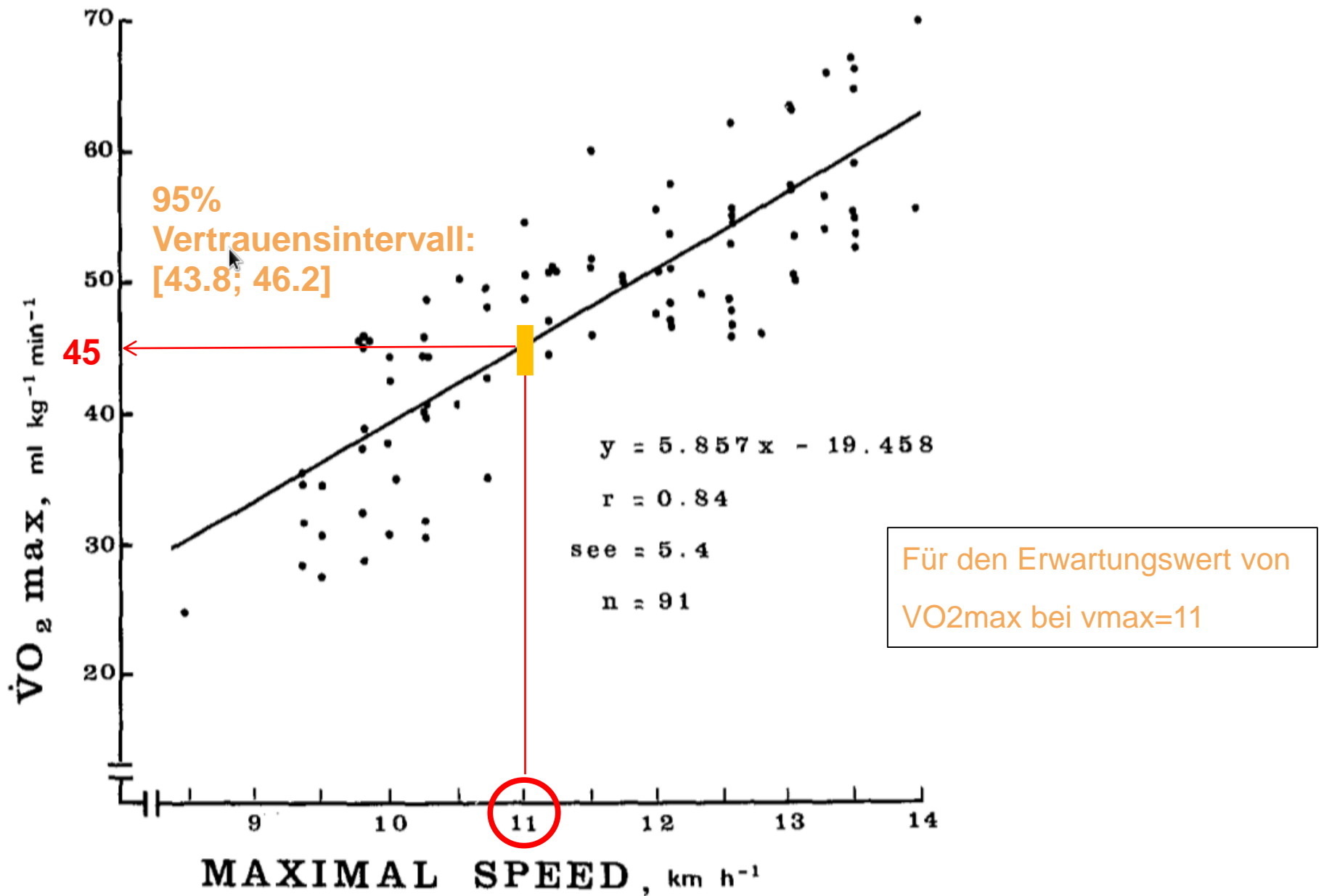
**Fig. 2.**  $\dot{V}O_2$  max as a function of the maximal speed achieved in the 20-m shuttle run test for a total sample of 91 adult subjects. Each point in this figure represents maximal effort



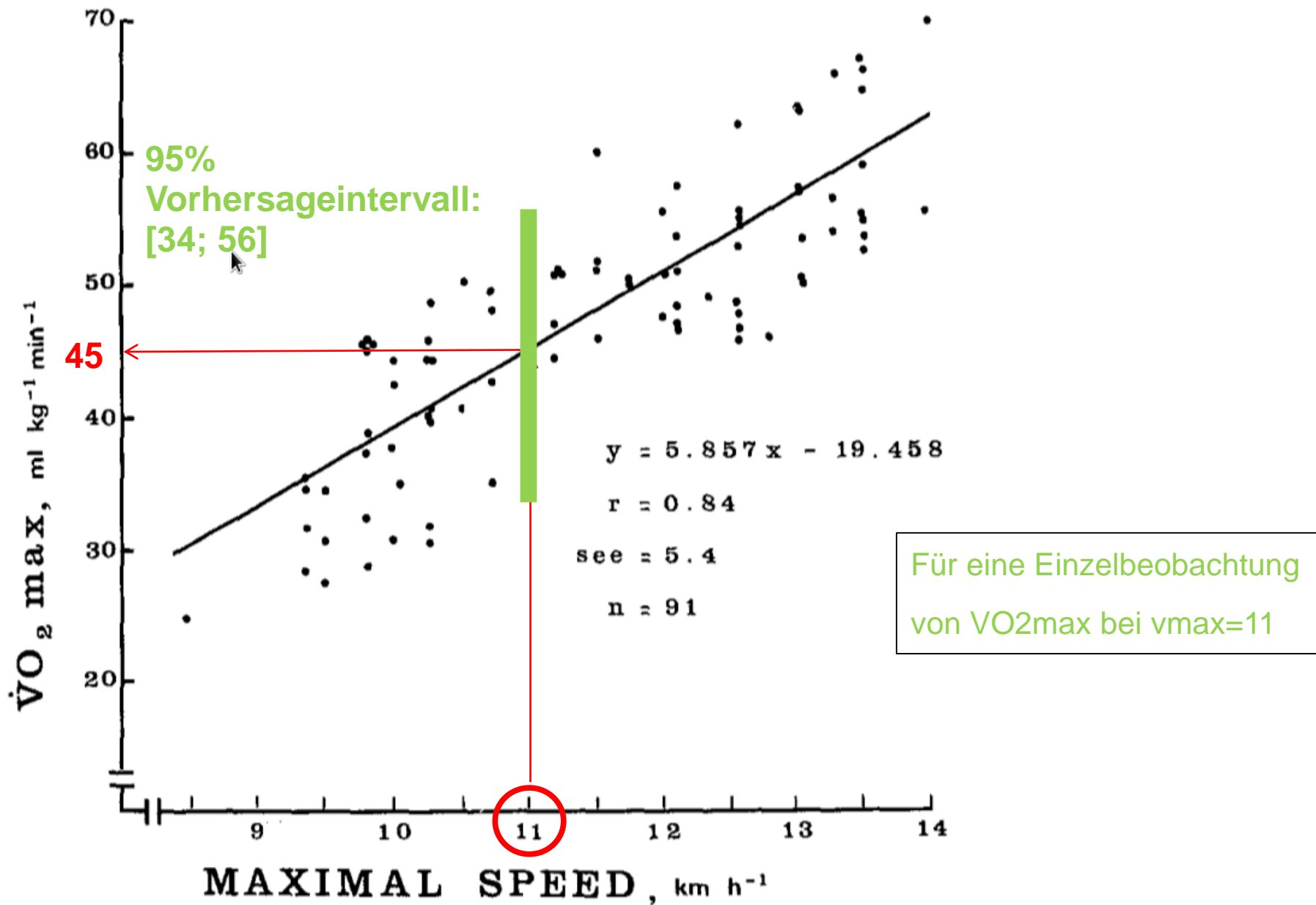
**Fig. 2.**  $\dot{V}O_2$  max as a function of the maximal speed achieved in the 20-m shuttle run test for a total sample of 91 adult subjects. Each point in this figure represents maximal effort



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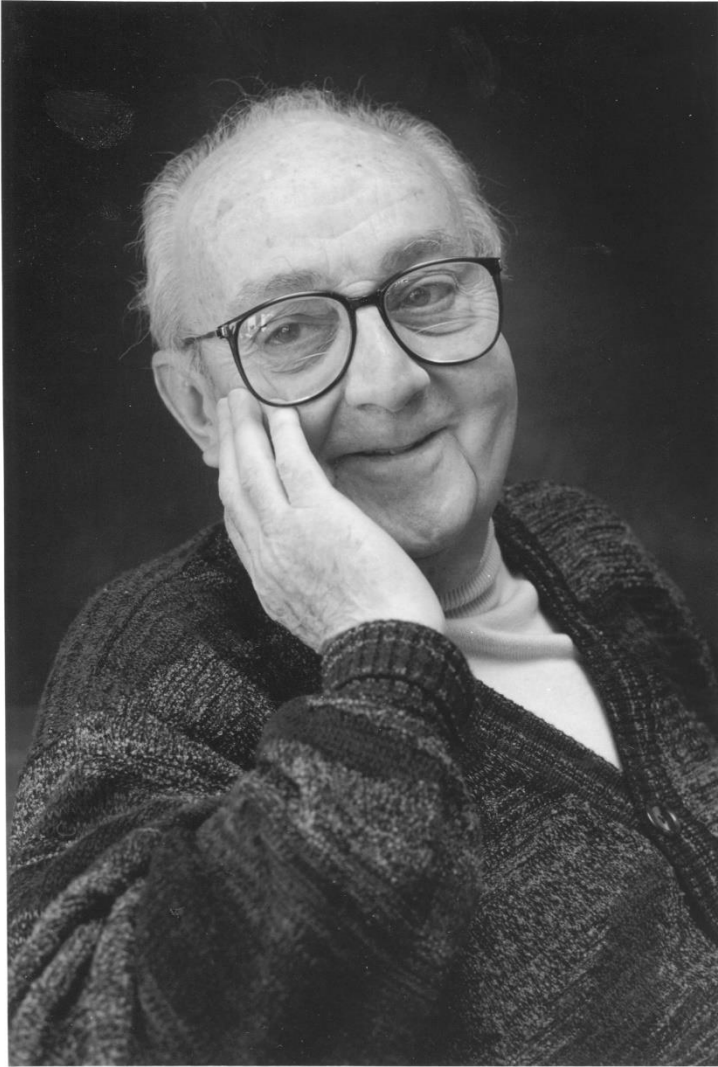


**Fig. 2.**  $\dot{V}O_2 \text{ max}$  as a function of the maximal speed achieved in the 20-m shuttle run test for a total sample of 91 adult subjects. Each point in this figure represents maximal effort



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George E.P. Box



“Essentially,  
all models are  
**wrong**,  
but some are  
**useful.**“

# Residuenanalyse: Wie gut stimmt das Modell ?

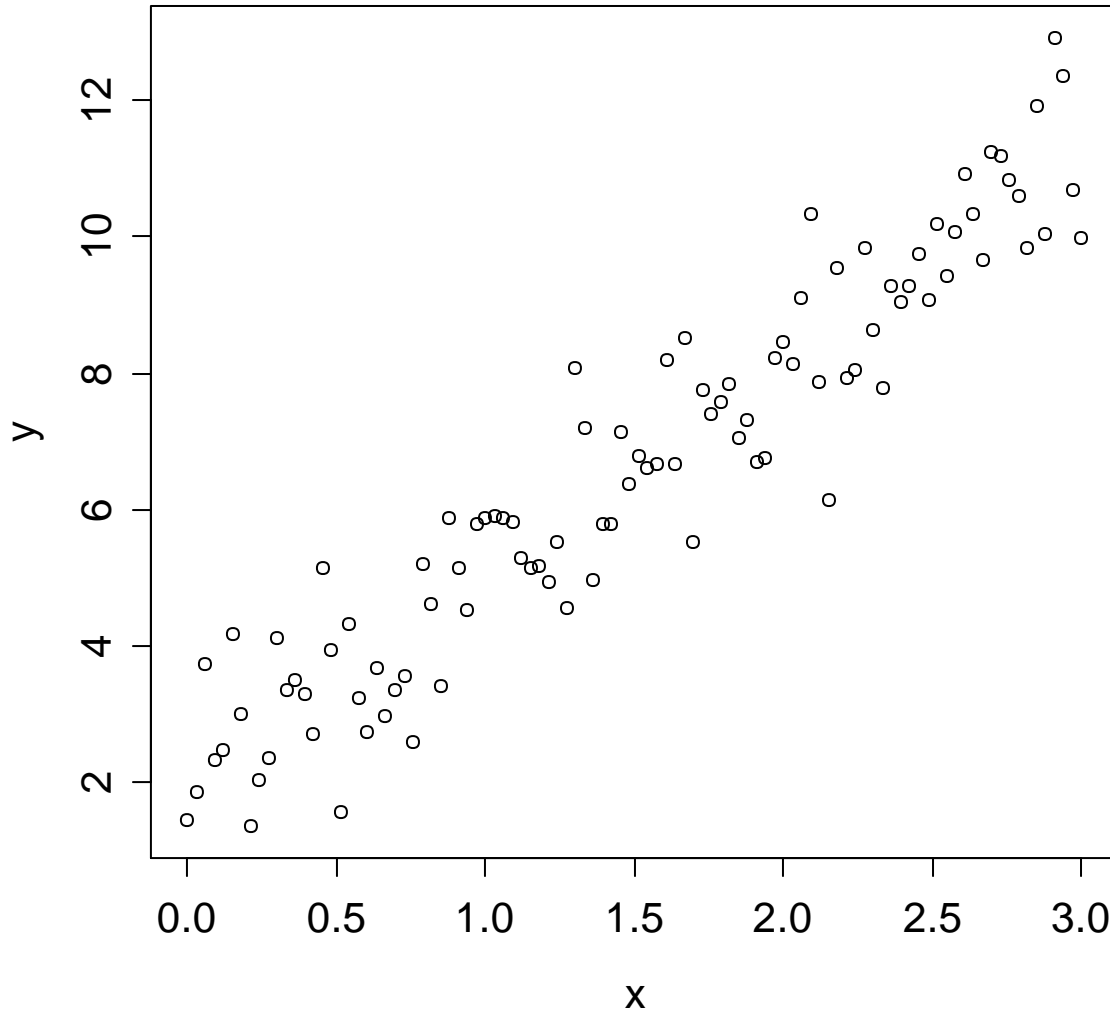
$$\underline{Y_i} = \beta_0 + \beta_1 x_i + \varepsilon_i ; \varepsilon_i \sim \underline{N(0, \sigma^2)} \quad iid$$

- Form des funktionellen Zusammenhangs
- Varianz der Fehler ist konstant
- Fehler sind normalverteilt

Einfache Regression:  
Streudiagramm  
Multiple Regression:  
Tukey-Anscombe Plot

QQ-Plot der  
Residuen

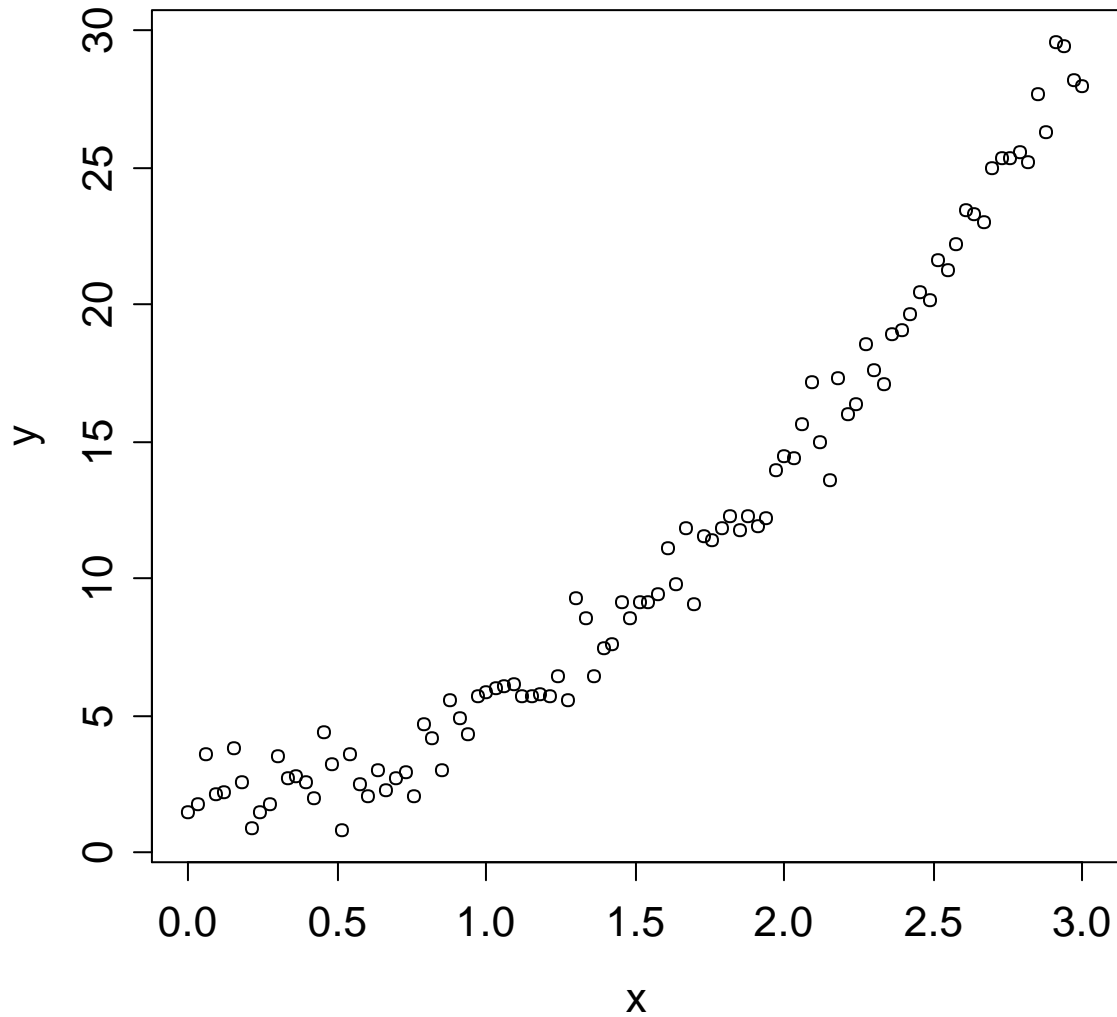
# Streudiagramm bei einfacher linearer Regression



OK



# Streudiagramm bei einfacher linearer Regression

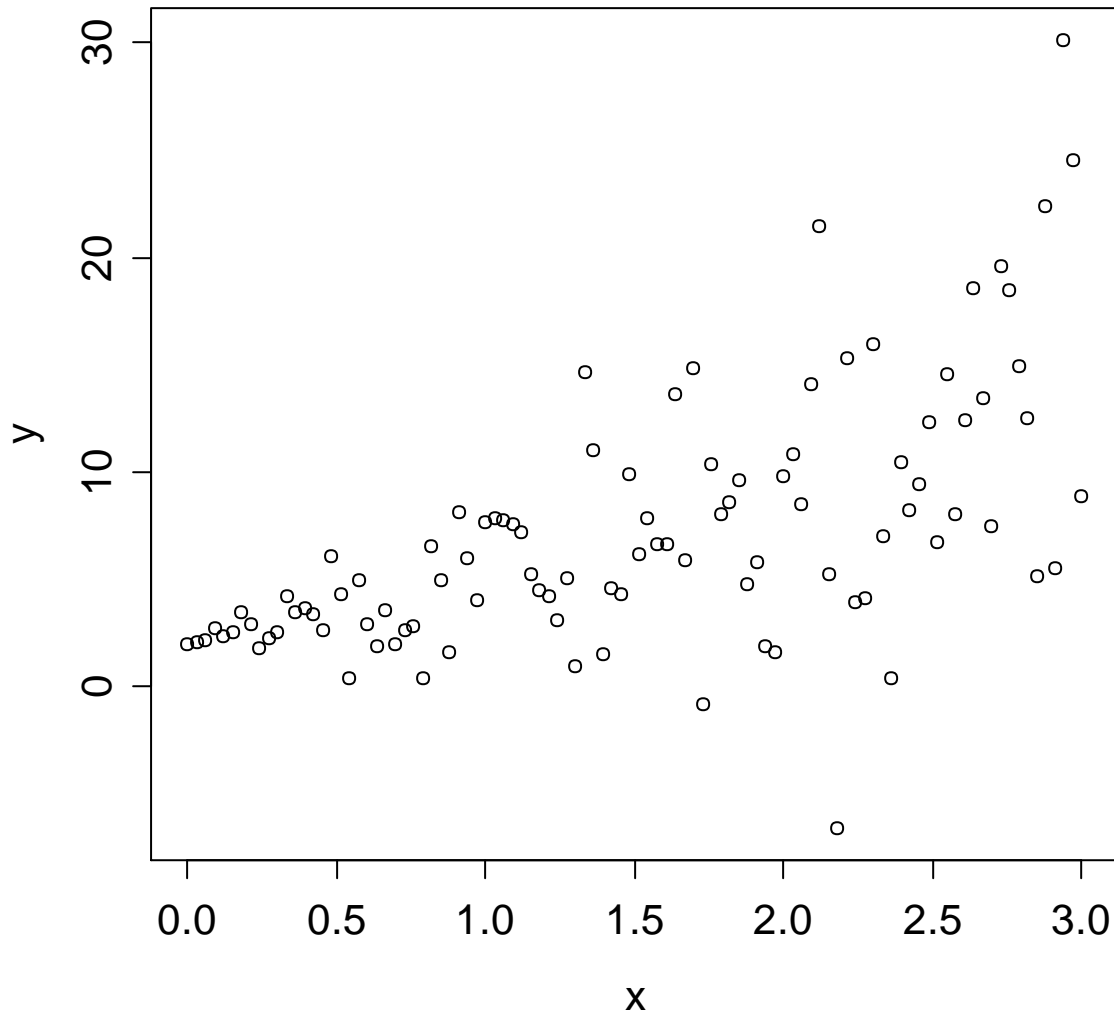


Systematischer Fehler

Krümmung:

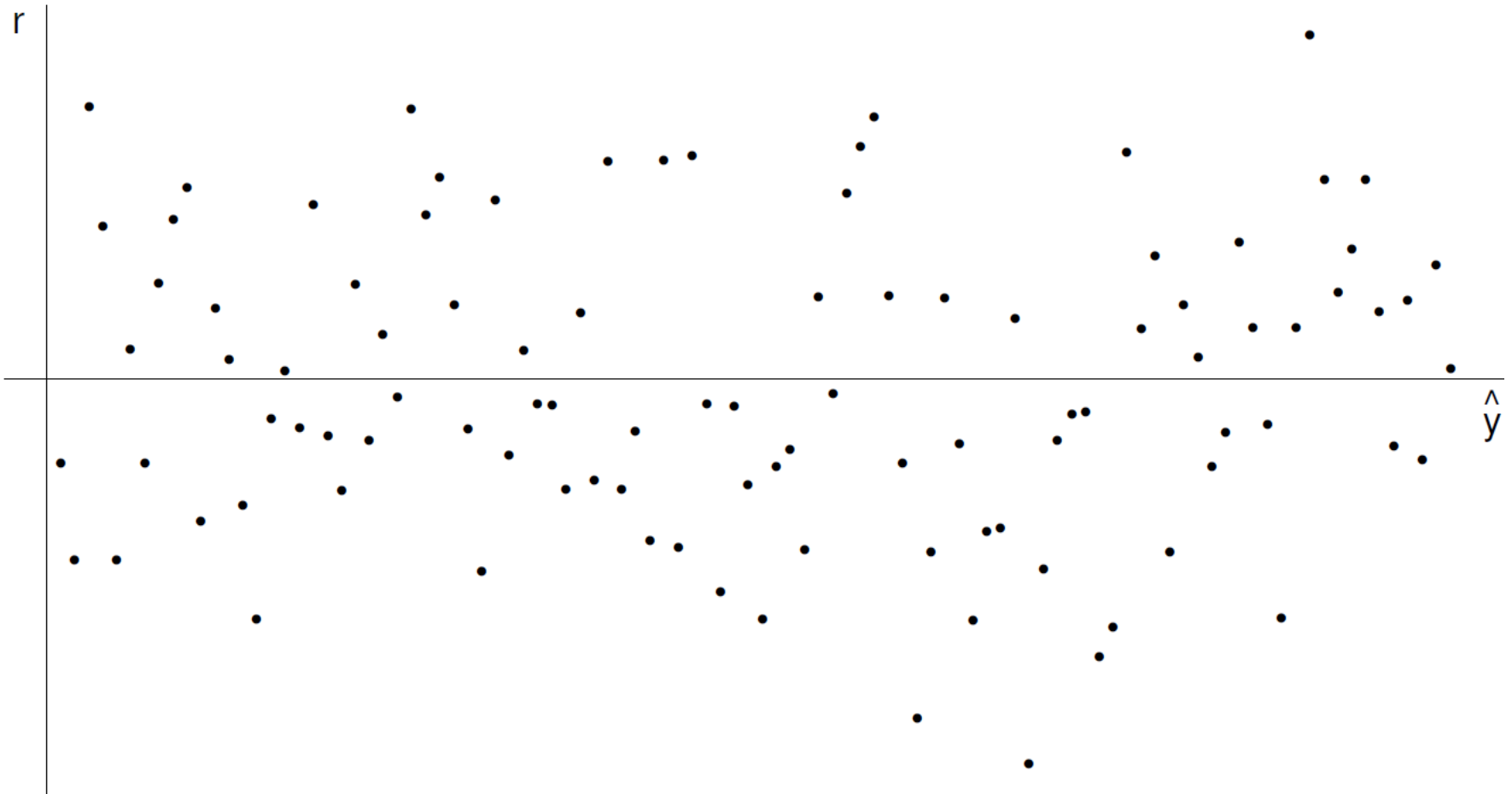
$$y = b_0 + b_1x + b_2x^2$$

# Streudiagramm bei einfacher linearer Regression

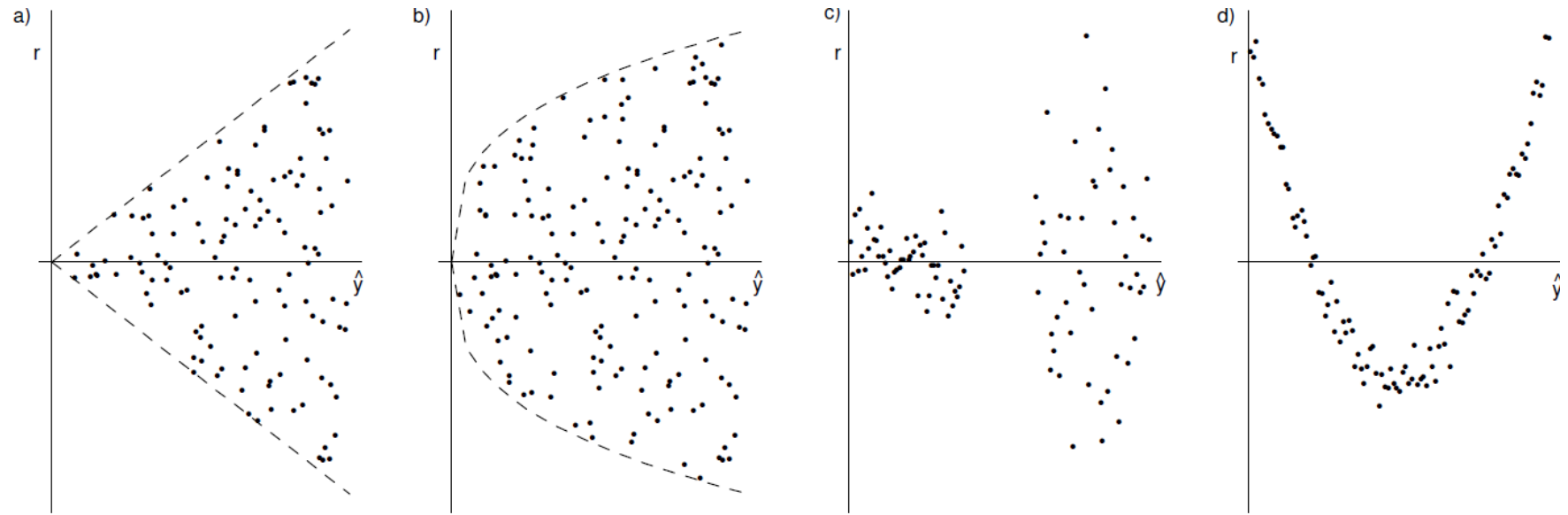


Fehlervarianz  
nicht konstant

# Beispiel für guten Tukey-Anscombe Plot



# Beispiele für schlechte Tukey-Anscombe Plots



Fehlervarianz nicht konstant

Systematischer Fehler

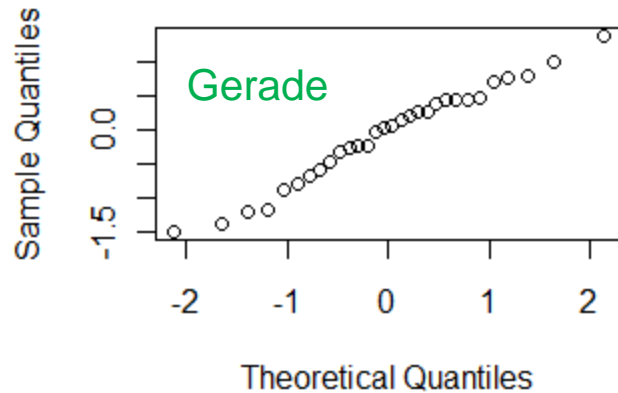
# Residuenanalyse: QQ-Plot

Gerade = "gut"

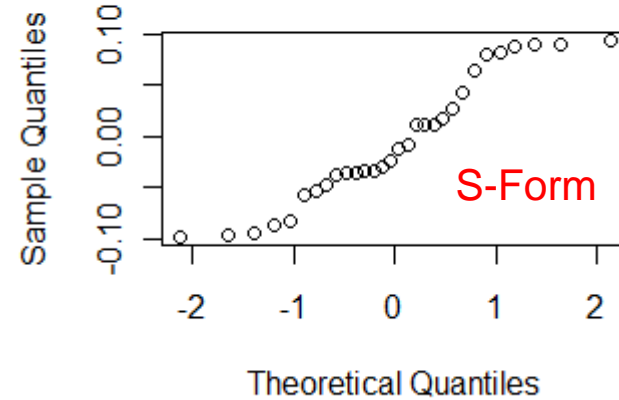
Krümmung = "schlecht"

Normal Q-Q Plot

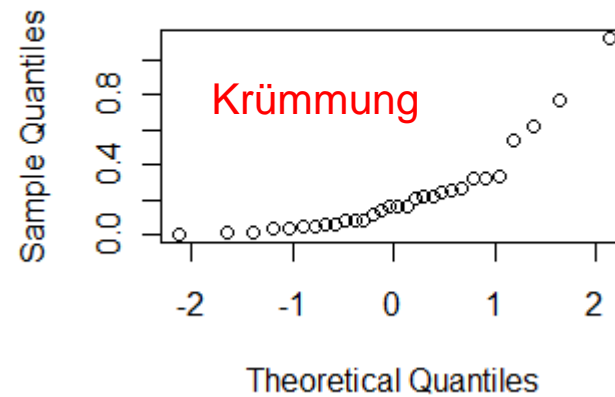
OK



Normal Q-Q Plot



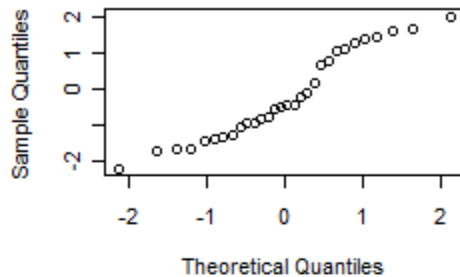
Normal Q-Q Plot



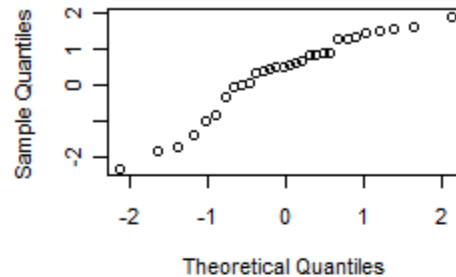
# QQ-Plots: Streuung von "guten" QQ-Plots

$(n = 30, R_i \sim N(0, 1))$

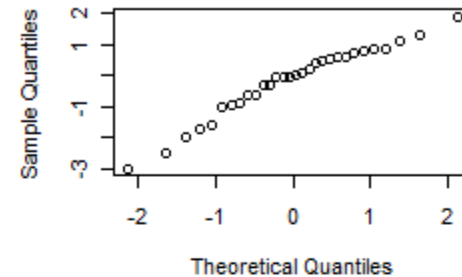
Normal Q-Q Plot



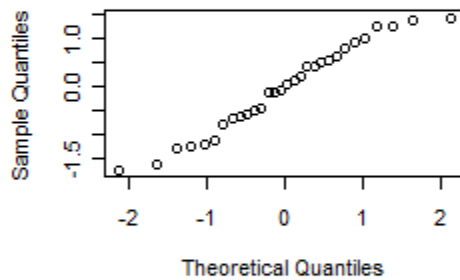
Normal Q-Q Plot



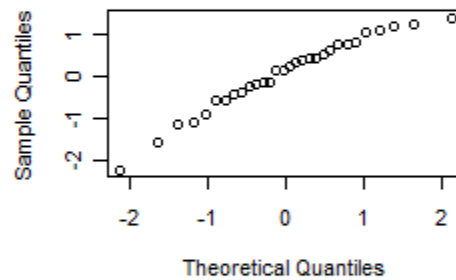
Normal Q-Q Plot



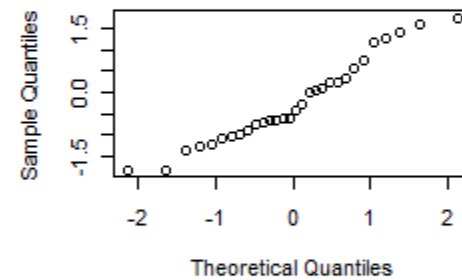
Normal Q-Q Plot



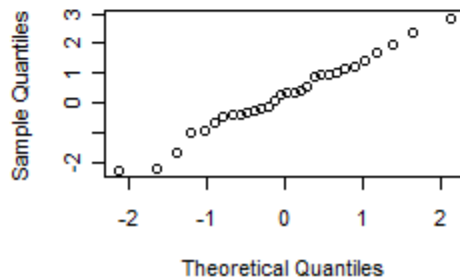
Normal Q-Q Plot



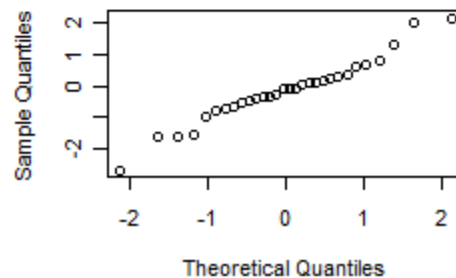
Normal Q-Q Plot



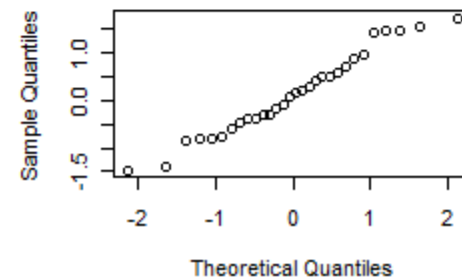
Normal Q-Q Plot



Normal Q-Q Plot



Normal Q-Q Plot



# Falls Residuenplots schlecht

- Oft helfen Transformationen von  $x$  oder  $y$
- Achtung: Vorsicht beim Interpretieren der neuen Parameter
- Bsp:  $\log(y)$  statt  $y$

Vorher:  $Y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$

Wenn  $x$  durch  $x+1$  ersetzt wird, ändert sich  $Y$  im Mittel zu  $Y + \beta_1$

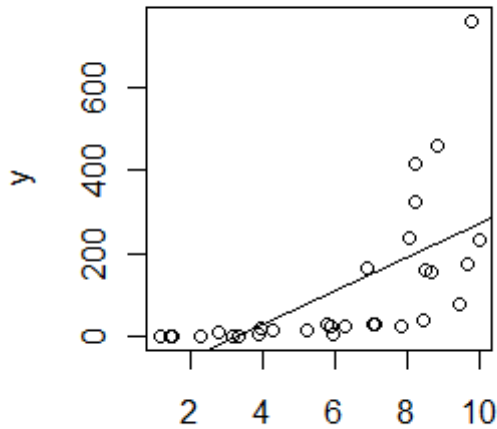
Nachher:

$$\log(Y_i) = \beta_0 + \beta_1 x_i + \varepsilon_i \leftrightarrow Y_i = \exp(\beta_0 + \beta_1 x_i + \varepsilon_i)$$

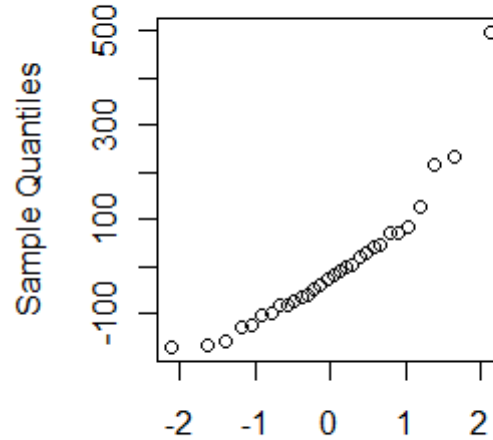
Wenn  $x$  durch  $x+1$  ersetzt wird, ändert sich  $Y$  "im Mittel" zu  $Y * \exp(\beta_1)$

# Bsp: Ohne Log-Transformation

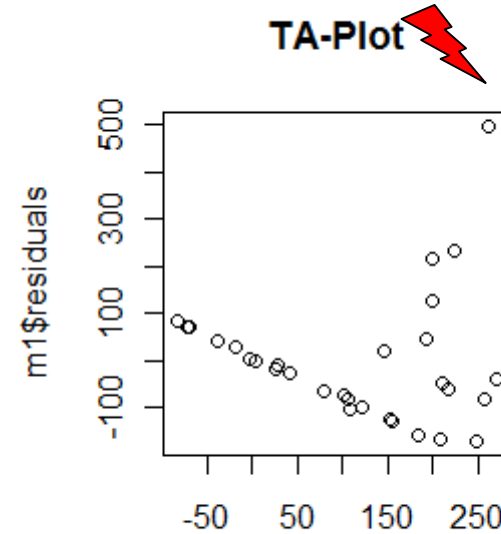
Streudiagramm



Normal Q-Q Plot

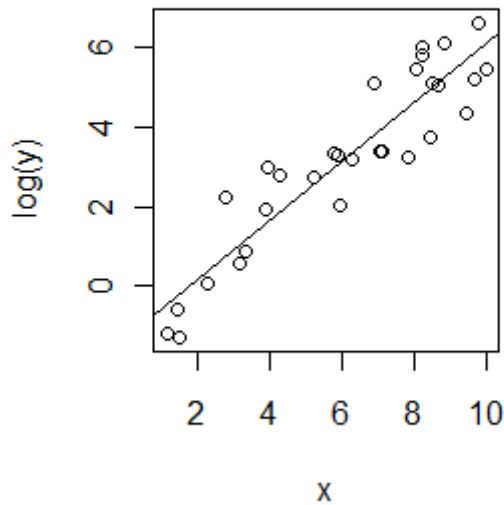


TA-Plot

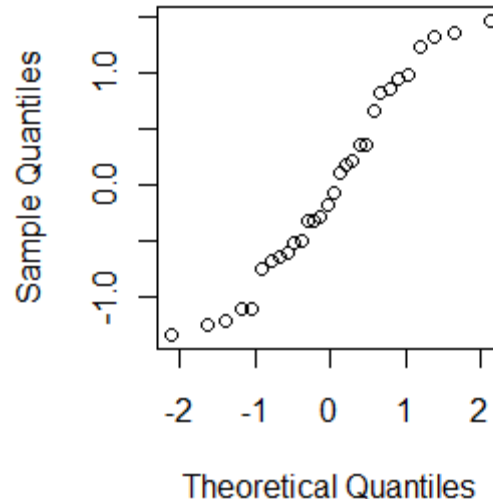


y

Streudiagramm (log(y))

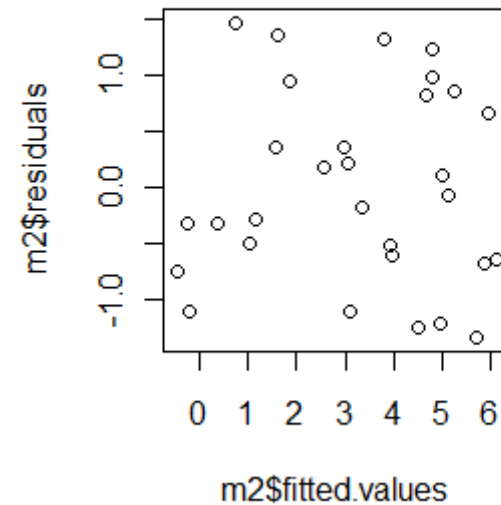


Normal Q-Q Plot



TA-Plot

OK

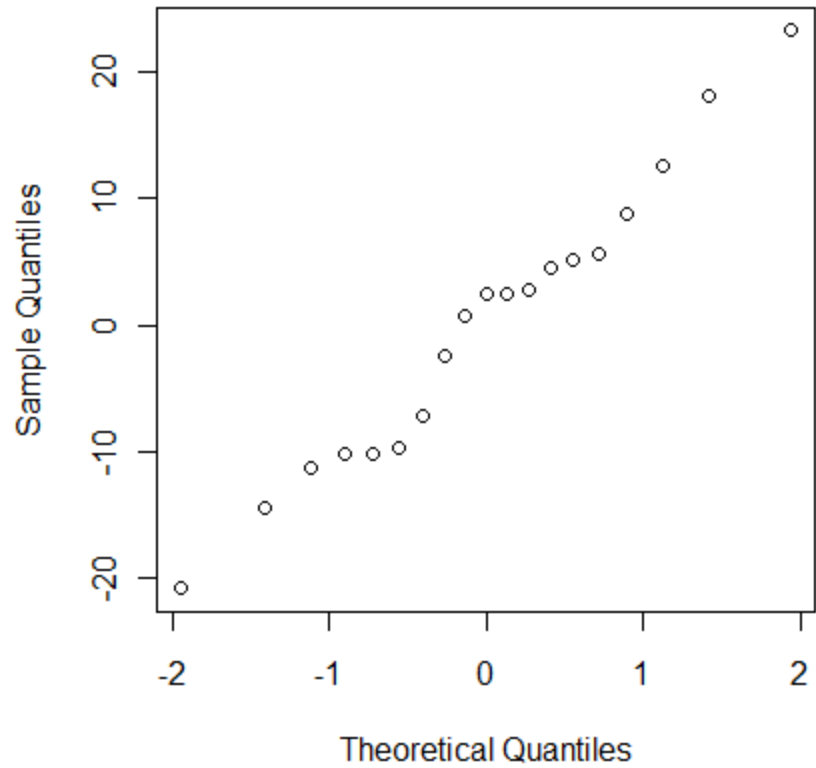


log(y)

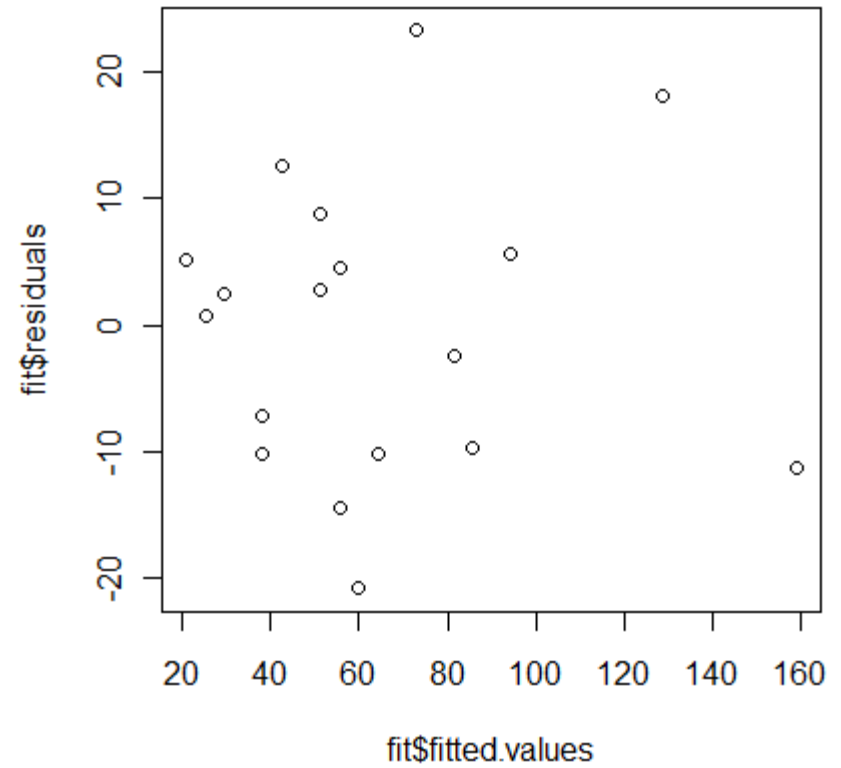


# Residuenanalyse: Supermarkt

Normal Q-Q Plot OK

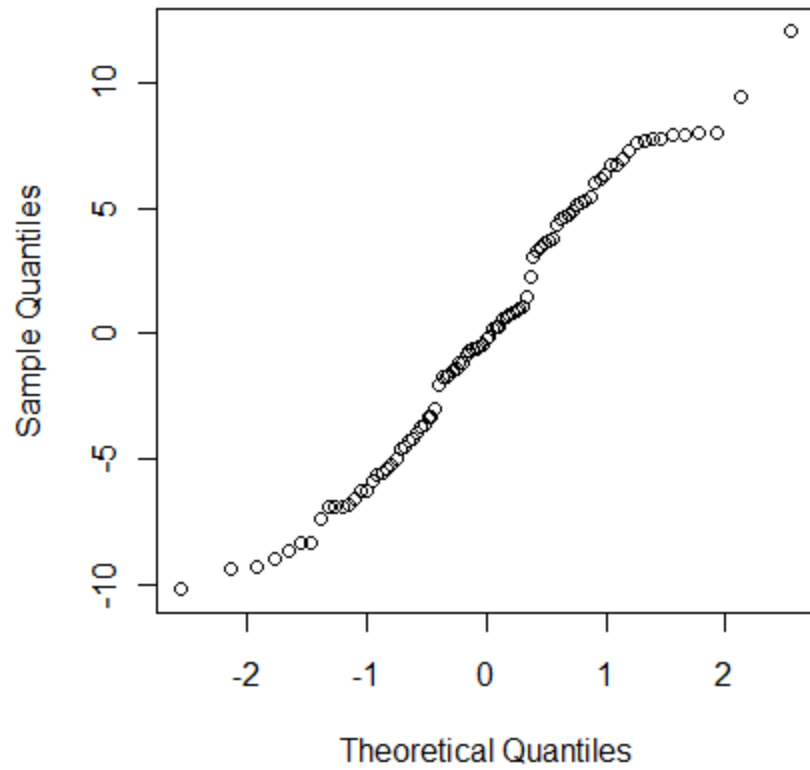


TA-Plot OK



# Residuenanalyse: Beep-Test

Normal Q-Q Plot OK



TA-Plot OK

