

Statistics for Medicine

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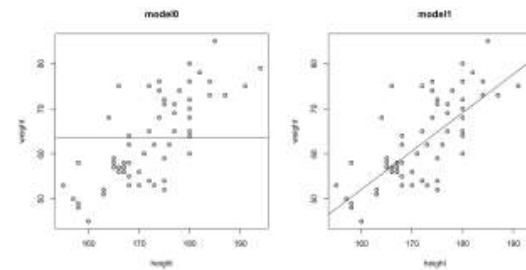
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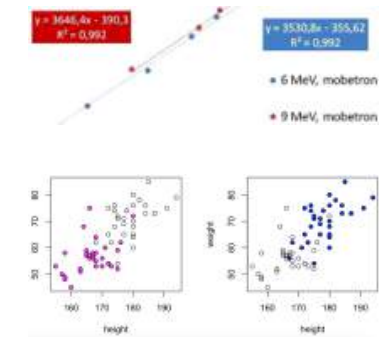
Recap: regression line

the fresher.ods dataset



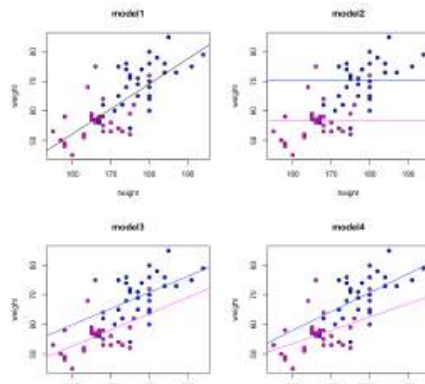
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today



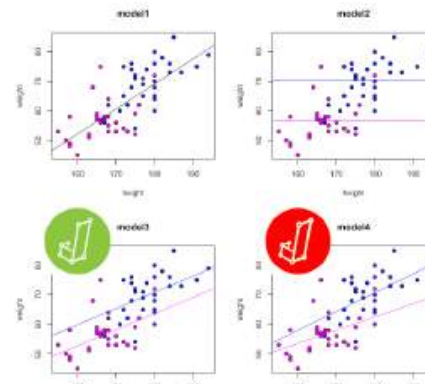
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there are many possibilities



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there are many possibilities



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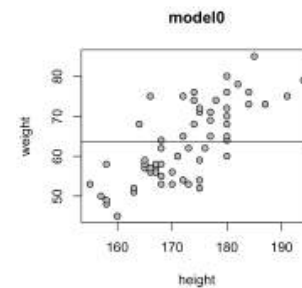
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let's move to R



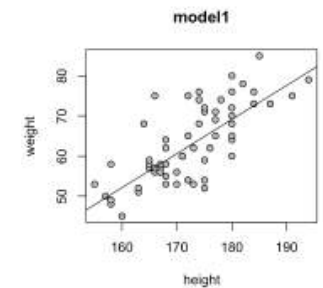
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weight ~ 1



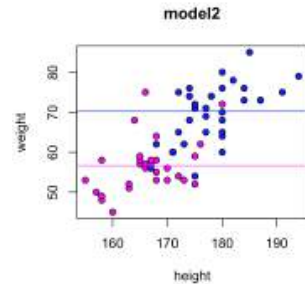
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weight ~ height



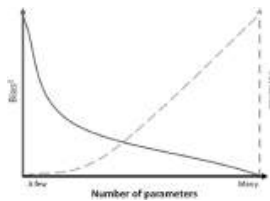
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● $\text{weight} \sim \text{gender}$



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the Akaike criterion

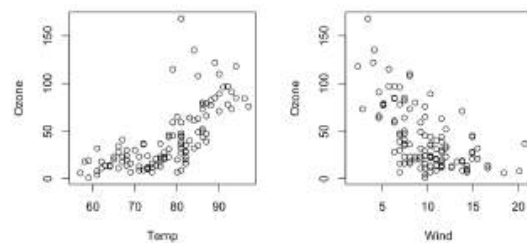


<https://www.sciencedirect.com/science/article/pii/S2468042719300508>

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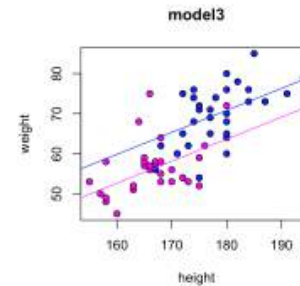
curvature in linear models
generalized linear model
repeated measures

the airquality dataset



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● $\text{weight} \sim \text{gender} + \text{height}$



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curvature in linear models
generalized linear model
repeated measures

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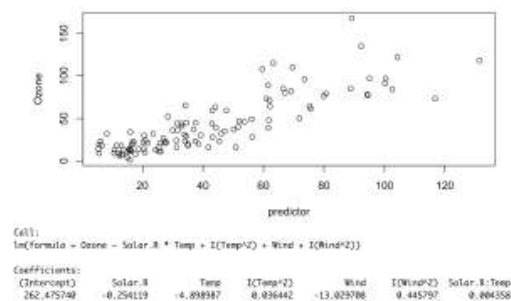
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curvature in linear models
generalized linear model
repeated measures

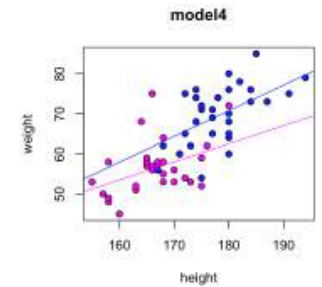


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● $\text{weight} \sim \text{gender} * \text{height}$

the same:

● $\text{weight} \sim \text{gender} + \text{height} + \text{height:gender}$



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curvature in linear models
generalized linear model
repeated measures

● curvature in linear models

● generalized linear model

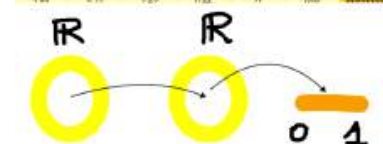
● repeated measures

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curvature in linear models
generalized linear model
repeated measures

the roma dataset

Year	Month	Day	Time	Temp	Humidity	Wind	Cloud	Pressure	Visibility	Station
1994	1	1	12:00	12.0	65	1.0	100	1013.2	10	Roma
1994	1	2	12:00	11.5	68	1.0	100	1013.1	10	Roma
1994	1	3	12:00	11.0	70	1.0	100	1013.0	10	Roma
1994	1	4	12:00	10.5	72	1.0	100	1012.9	10	Roma
1994	1	5	12:00	10.0	75	1.0	100	1012.8	10	Roma
1994	1	6	12:00	9.5	78	1.0	100	1012.7	10	Roma
1994	1	7	12:00	9.0	80	1.0	100	1012.6	10	Roma
1994	1	8	12:00	8.5	82	1.0	100	1012.5	10	Roma
1994	1	9	12:00	8.0	85	1.0	100	1012.4	10	Roma
1994	1	10	12:00	7.5	88	1.0	100	1012.3	10	Roma



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Logit

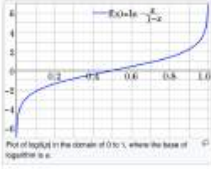
From Wikipedia, the free encyclopedia

This article discusses the binary logit function only. See choice choice for a discussion of multinomial logit, conditional logit, nested logit, mixed logit, exploded logit, and ordered logit. For the basic regression technique that uses the logit function, see logistic regression. For standard regression combined by multiplication, see logit link.

In statistics, the **logit** (1/total) (O-H) function is the quantile function associated with the standard logistic distribution. It has many uses in data analysis and machine learning, especially in data transformation.

Mathematically, the logit is the inverse of the standard logistic function $\sigma(x) = 1/(1 + e^{-x})$, so the logit is defined as $\text{logit}(p) = \sigma^{-1}(p) = \ln\left(\frac{p}{1-p}\right)$ for $p \in (0, 1)$.

Because of this, the logit is also called the **log-odds** since it is equal to the logarithm of the odds, $\frac{p}{1-p}$, where p is a probability. Thus, the logit is a type of function that maps



	Alice	Ellen
1	73.60	73.80
2	73.40	73.50
3	74.10	74.60
4	73.50	73.80
5	73.20	73.60

The standard logistic function is the logistic function with

$$f(x) = \frac{1}{1 + e^{-x}} = \frac{e^x}{e^x + 1} = \frac{1}{2} + \frac{1}{2} \tanh\left(\frac{x}{2}\right).$$

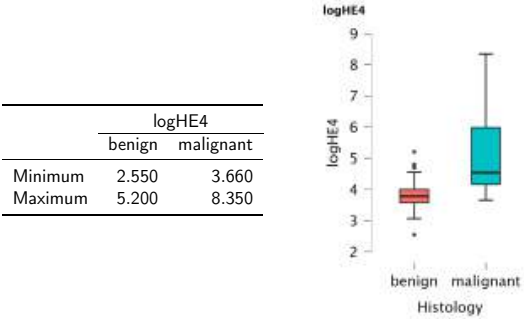
	Est.	St. Error	z	Wald Test		
				Wald	df	p
(Intercept)	-14.28	2.38	-6.00	35.98	1	< .001
logHE4	3.07	0.57	5.38	28.94	1	< .001



data: alice and ellen

t = -1.2227, df = 8, p-value = 0.2562
alternative hypothesis: true difference in means
is not equal to 0

95 percent confidence interval:
-0.865794 0.265794
sample estimates:
mean of x mean of y
73.56 73.86



Alice	Ellen
73.60	73.80

	Alice	Ellen		Alice	Ellen
1	73.60	73.80	12	74.10	74.60
2	73.40	73.50	13	73.60	73.80
3	74.10	74.60	14	73.40	73.60
4	73.50	73.80	15	74.10	74.40
5	73.20	73.60	16	73.50	73.70
6	74.00	74.40	17	73.20	73.50
7	73.60	73.80	18	74.00	74.40
8	73.30	73.50	19	73.60	73.90
9	74.20	74.30	20	73.30	73.60
10	73.60	73.90	21	74.20	74.50
11	73.40	73.60	-	-	-

Two Sample t-test

data: peso by gemella

t = -2.4594, df = 40, p-value = 0.01834
alternative hypothesis: true difference in means
is not equal to 0

95 percent confidence interval:
-0.51183215 -0.05007261

sample estimates:

mean in group alice	mean in group ellen
73.66190	73.94286