Statistics for Medicine

Massimo Borelli

Master of Advanced Studies in Medical Physics



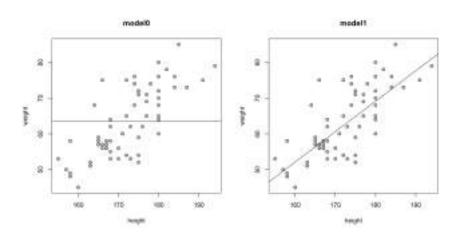


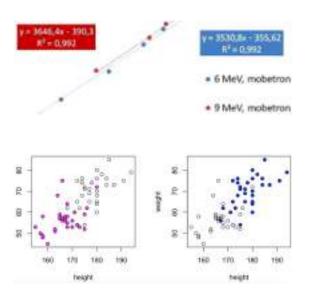




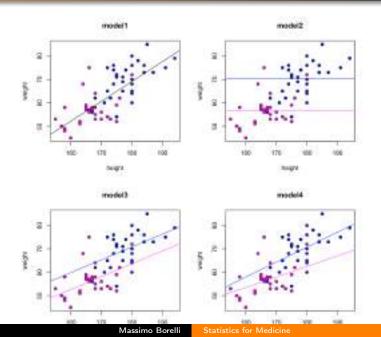
Recap: regression line

the fresher.ods dataset

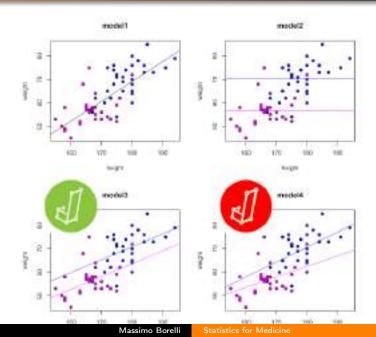


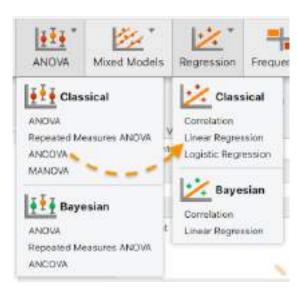


there are many possibilities



there are many possibilities



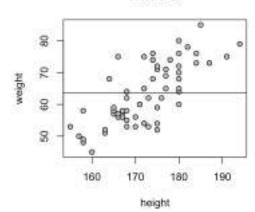


let's move to R



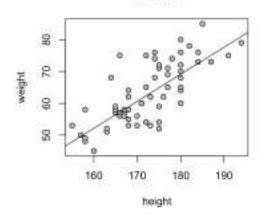


model0



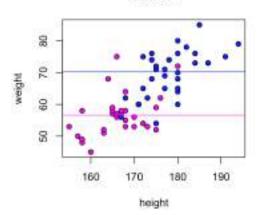
ullet weight \sim 1

model1



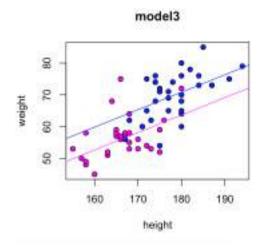
ullet weight \sim height

model2



ullet weight \sim gender

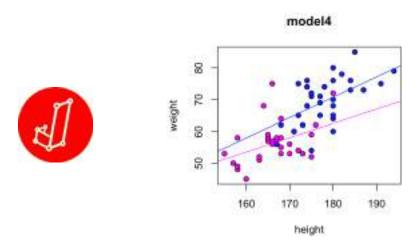
ullet weight \sim gender + height



ullet weight \sim gender * height

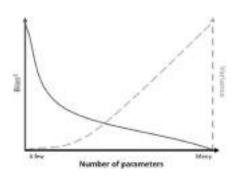
the same:

ullet weight \sim gender + height + height:gender



the Akaike criterion





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

Statistics for Medicine

Massimo Borelli

Master of Advanced Studies in Medical Physics







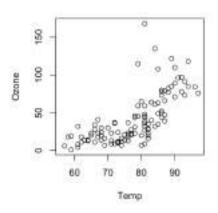
curvature in linear models generalized linear model repeated measures

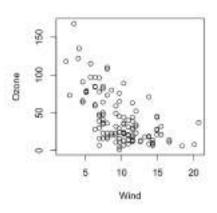
curvature in linear models

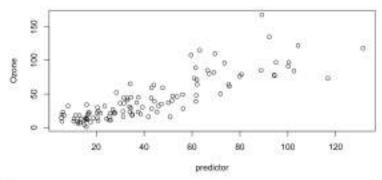
generalized linear model

repeated measures

the airquality dataset



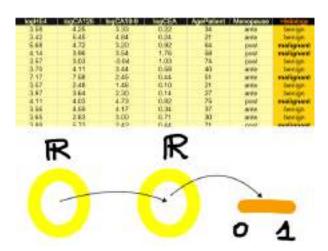




Cell: lm(formula = Oppre = Solar A * Temp + I(TempAI) + Wind + I(Wind*2))

Coefficients:

(Intercept) Solar 8 Temp I(Temp*2) Wind I(Wind*2) Solar 8: Temp 262 475749 -0.254119 -4.898987 0.036442 -13.029786 0.445797 0.804358



Logit

From Whipedia, the they excystopedia.

The article discusses the binary legit function only like discrete allows for a decursion of multiconial legit, annalist legit, neutral legit, excelled legit, and undered legit. For the basic regression truthique that uses the legit function, are explaint regression. For scandard regit, and continued by multiplication, are legit suits.

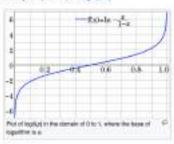
in electrics, the legit (*lookge* con-jib function is the quantile function associated with the standard logistic distribution. It has many uses in data unayels and machine learning, especially in this transformations.

intervariously, the logil is the inverse of the standard togeth: for an $\sigma(x) = 1/(1 + e^{-x})$, so the logil is defined as

$$logit(p) = \sigma^{-1}(p) = ln\left(\frac{p}{1-p}\right) \text{ for } p \in (0,1).$$

Because of this, the logit is also called the log-odds since it is equal to the logarithm of the code $\frac{p}{1-a}$ where p is a

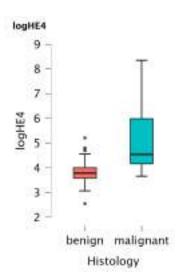
probability. Thus, the logit is a type of function that maps



The standard logistic function is the logistic function with p

$$f(x) = rac{1}{1+e^{-x}} = rac{e^x}{e^x+1} = rac{1}{2} + rac{1}{2} anh \Big(rac{x}{2}\Big).$$

	logHE4		
	benign malignar		
Minimum	2.550	3.660	
Maximum	5.200	8.350	

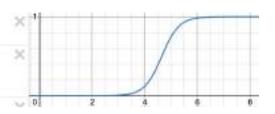




				Wald Test		
	Est.	St. Error	Z	Wald	df	р
(Intercept)	-14.28	2.38	-6.00	35.98	1	< .001
logHE4	3.07	0.57	5.38	28.94	1	< .001

$$f(x) = -14.28 + 3.07x$$

$$y = \frac{\exp(f(x))}{(1 + \exp(f(x)))}$$



repeated measures



Alice Ellen 73.60 73.80

Massimo Borelli

Statistics for Medicine

repeated measures



	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

```
Two Sample t-test
```

```
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:

73.56 73.86

repeated measures

	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