#### P5.2 Statistics for Medicine

#### Massimo Borelli

Master of Advanced Studies in Medical Physics







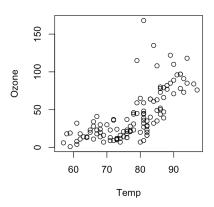
curvature in linear models generalized linear model repeated measures

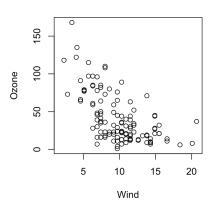
curvature in linear models

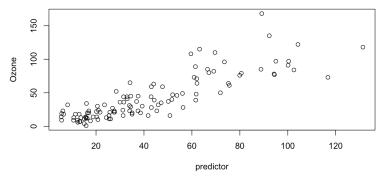
2 generalized linear model

3 repeated measures

## the airquality dataset





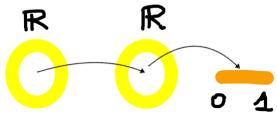


Call:  $lm(formula = Ozone \sim Solar.R * Temp + I(Temp^2) + Wind + I(Wind^2))$ 

Coefficients:

(Intercept) Solar.R Temp I(Temp^2) Wind I(Wind^2) Solar.R:Temp 262.475740 -0.254119 -4.898987 0.036442 -13.029708 0.445797 0.004358

logHE4	logCA125	logCA19-9	logCEA	AgePatient	Menopause	Histology
3.58	4.25	3.33	0.22	34	ante	benign
3.42	5.45	4.84	0.24	21	ante	benign
5.68	4.72	3.20	0.92	64	post	malignant
4.14	3.96	3.54	1.76	58	post	malignant
3.57	3.03	-0.04	1.03	74	post	benign
3.70	4.11	3.44	0.58	40	ante	benign
7.17	7.58	2.45	0.44	51	ante	malignant
3.57	2.48	1.46	0.10	21	ante	benign
3.97	3.64	2.30	0.14	27	ante	benign
4.11	4.03	4.73	0.82	75	post	malignant
3.56	4.59	4.17	0.34	37	ante	benign
3.65	2.83	3.00	0.71	30	ante	benign
3.89	5.72	2 42	0.44	71	nost	malignant



#### Logit

From Wikipedia, the free encyclopedia

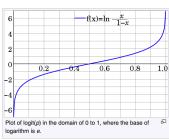
This article discusses the binary logit function only. See discrete choice for a discussion of multinomial logit, conditional logit, nested logit, mixed logit, exploded logit, and ordered logit. For the basic regression technique that uses the logit function, see logistic regression. For standard magnitudes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

In statistics, the **logit** (/lood<u>stt/</u> LOH-jit) function is the quantile function associated with the standard logistic distribution. It has many uses in data analysis and machine learning, especially in data transformations.

Mathematically, the logit is the inverse of the standard logistic function  $\sigma(x)=1/(1+e^{-x})$ , so the logit is defined as

$$\operatorname{logit}(p) = \sigma^{-1}(p) = \ln\!\left(rac{p}{1-p}
ight) \quad ext{for} \quad p \in (0,1).$$

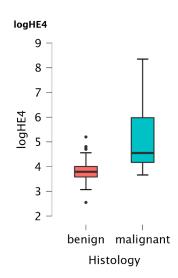
Because of this, the logit is also called the **log-odds** since it is equal to the logarithm of the odds  $\frac{p}{1-p}$  where p is a probability. Thus, the logit is a type function that maps



The standard logistic function is the logistic function with I

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

	logHE4		
	benign	malignant	
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

# 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
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```
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
```

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