

# UNIVERSITY OF OSLO

## Faculty of Mathematics and Natural Sciences

Examination in: STK4900 — Statistical methods and applications.

Day of examination: Tuesday 12 June 2012.

Examination hours: 14.30–18.30.

This problem set consists of 6 pages.

Appendices: Tables for the standard normal distribution, the chi-square distributions, the  $t$  distributions, and the  $F$  distributions.

Permitted aids: All printed and hand-written resources. Approved calculator.

Please make sure that your copy of the problem set is complete before you attempt to answer anything.

### Problem 1

In this problem we want to study how the yield of wheat depends on moisture in the ground and the variety (or type) of wheat. Each of ten varieties of wheat was planted once on each of six plots, giving a total of 60 observations.

The data set contains the variables:

`yield` yield of wheat in bushels per acre  
`moist` moisture in the top 36 inches of soil before planting  
`variety` ten different types of wheat (labeled 1 to 10)  
`plot` six randomly chosen plots, each of size 1 acre (labeled 1 to 6)

Below is given an ANOVA-table with `yield` as the response variable and `variety` as a categorical covariate. Two numbers in the table are replaced by question marks.

#### Output 1:

```
anova(lm(yield~factor(variety),data=wheat))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
factor(variety)	?	4089.1	454.34	?	0.00013
Residuals	50	4756.3	95.13		

(edited output)

- Describe the model that the ANOVA-table is based on. Are the varieties significantly different? Give the two numbers in the table that are replaced by question marks and explain how they are determined.

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In a second model we have also included the covariate `moist`. An edited version of this regression analysis is presented below.

### Output 2:

```
summary(lm(yield~moist+factor(variety),data=wheat))
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	31.995733	0.496150	64.488	< 2e-16
moist	0.670836	0.008473	79.173	< 2e-16
factor(variety)2	-2.884687	0.515474	-5.596	9.74e-07
factor(variety)3	3.183343	0.502109	6.340	6.99e-08
factor(variety)4	1.284064	0.516489	2.486	0.016372
....				
factor(variety)10	3.074284	0.506524	6.069	1.83e-07

```
Residual standard error: 0.8677 on 49 degrees of freedom
Multiple R-squared: 0.9958, Adjusted R-squared: 0.995
F-statistic: 1170 on 10 and 49 DF, p-value: < 2.2e-16
```

(edited output)

- b) State the model for this analysis, and give an interpretation of the estimated coefficients.
- c) Discuss the concept of  $R^2$  and explain how this measure has been computed for the model in question b). Also calculate  $R^2$  for the model in question a). Compare the two models with respect to their ability to predict the yield of wheat.

## Problem 2

In the period from 1999 to 2001 a number of cod along the coast of Finmark in North Norway were examined for infection by the blood parasite *Trypanosoma murmanensis*. In this problem we will study how the risk of infection depends on some covariates.

The response variable is `parasite`, which is coded as 0 if a fish is not infected by the blood parasite and as 1 if a fish is infected. We will relate this response to the following covariates:

`year` the year when the fish is caught  
(1: year 1999; 2: year 2000; 3: year 2001)

`weight` weight of the fish (in kg)

`age` age of the fish (in years)

Throughout the problem `year` is treated as a categorical covariate (with three levels).

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Further we center the numeric covariates `weight` and `age` by subtracting their means (the mean weight is 1.75 kg and the mean age is 4.4 years).

- a) Explain why logistic regression is an appropriate model for analysing the data. Give an explicit formulation of the logistic regression model when year is the only covariate. (Remember that we treat year as a categorical covariate.)

When we fit the logistic regression model with `year` as the only covariate, we get the result:

### Output 3:

Call:

```
glm(formula = parasite ~ factor(year), family = binomial)
```

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-0.99119	0.19516	-5.079	3.80e-07
factor(year)2	1.28666	0.30429	4.228	2.35e-05
factor(year)3	0.07936	0.26313	0.302	0.763

```
Null deviance: 467.82 on 364 degrees of freedom
Residual deviance: 445.80 on 362 degrees of freedom
```

(edited output)

- b) Describe how one may ascertain that there is a significant effect of year. Discuss how the probability that a fish is infected depends on the year it is caught.

Next we fit a model with the covariates `year` and `weight`:

### Output 4:

Call:

```
glm(formula = parasite ~ factor(year) + I(weight-1.75), family = binomial)
```

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-1.06287	0.19920	-5.336	9.52e-08
factor(year)2	1.40229	0.31186	4.497	6.91e-06
factor(year)3	0.15834	0.26622	0.595	0.552
I(weight-1.75)	-0.22404	0.09995	-2.241	0.025

(edited output)

- c) Define the odds ratio corresponding to 1 kg increase in the weight of a fish. Estimate the odds ratio and find a 95% confidence interval for it. Describe what the estimated odds ratio and the confidence interval tell you.

Then we fit a model with all the three covariates :

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**Output 5:**

Call:

```
glm(formula=parasite~factor(year)+I(weight-1.75)+I(age-4.4), family=binomial)
```

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-1.1409	0.2057	-5.546	2.92e-08
factor(year)2	1.3709	0.3190	4.297	1.73e-05
factor(year)3	0.2198	0.2707	0.812	0.416923
I(weight-1.75)	-0.7821	0.1984	-3.943	8.06e-05
I(age-4.4)	0.4800	0.1332	3.604	0.000313

```
Null deviance: 467.82 on 364 degrees of freedom
Residual deviance: 426.46 on 360 degrees of freedom
```

(edited output)

- d) The estimated effect of weight differs for the models in output 4 and output 5. Explain the reason for this, and interpret the effect of weight for the model in output 5. Also give an interpretation of the intercept for the model in output 5. (Remember that the two numeric covariates are centered.)

**Problem 3**

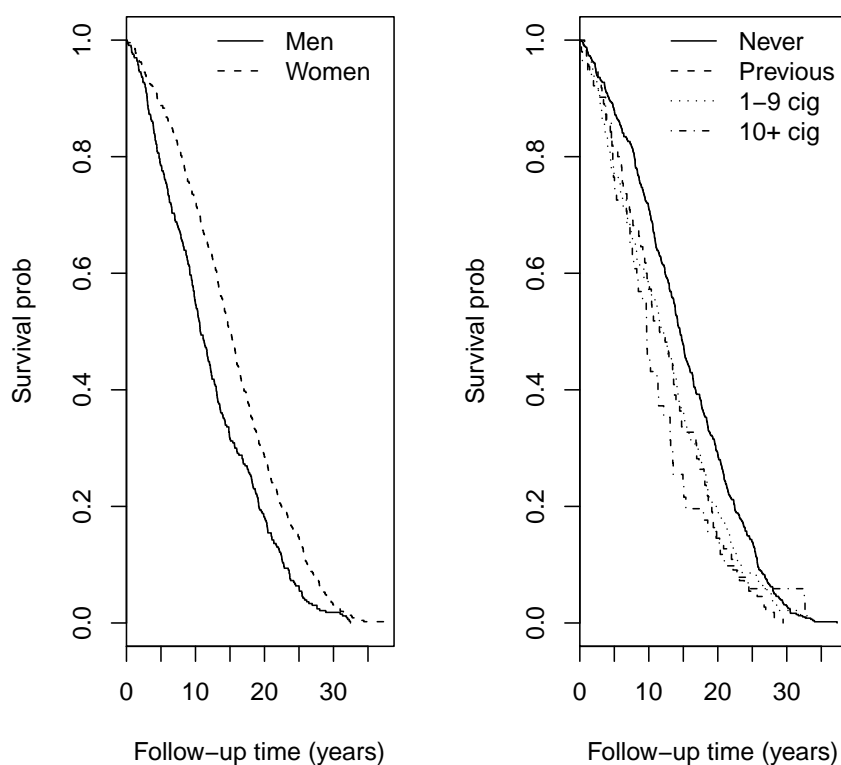
784 individuals aged 65-75 years participated in a blood pressure study in Bergen in the period 1966-1971. The individuals were followed until they died, emigrated or to the end of the study.

In our analyses we will use the covariates smoking and sex, in addition to the censored survival times and the indicators of death or censoring:

<b>time</b>	follow-up time from start of study to death or censoring
<b>death</b>	indicator of death/censoring (0: censoring; 1: death)
<b>smoking</b>	categorical smoking variable (1: never smoker; 2: previous smoker; 3: 1-9 cigarettes per day; 4: 10 or more cigarettes per day)
<b>sex</b>	sex (1: males; 2: females)

- a) Discuss the concept of right censored survival data and explain why methods like linear or logistic regression are inappropriate for analyzing such data.
- b) The plots on the next page show Kaplan-Meier estimates of the survival function according to sex and smoking categories. Discuss the observed differences in mortality based on these plots.

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Since we are dealing with two categorical covariates it is useful to model the mortality by regression methods, and the most common regression method for censored survival data is Cox's regression model.

- c) Give a description of Cox's regression model. In particular describe how we may interpret the regression coefficients as logarithms of hazard ratios.

Interpret the hazard ratios from the Cox-regression with sex and smoking category from the edited R-output below.

**Output 6:**

	coef	exp(coef)	se(coef)	z	p
sex	-0.301	0.74	0.0865	-3.48	0.0005
factor(smoking)2	0.162	1.18	0.1201	1.35	0.1800
factor(smoking)3	0.127	1.14	0.1043	1.22	0.2200
factor(smoking)4	0.232	1.26	0.1536	1.51	0.1300

- d) Below is given edited R-output of a Cox-regression with smoking as the only covariate. Compare the estimated effects of smoking for this model with the estimates for the model in question c). Discuss why the estimated effects of smoking

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differ for the two models.

**Output 7:**

	coef	exp(coef)	se(coef)	z	p
factor(smoking)2	0.358	1.43	0.1063	3.36	0.00077
factor(smoking)3	0.264	1.30	0.0962	2.74	0.00610
factor(smoking)4	0.366	1.44	0.1486	2.46	0.01400

**END**