## UNIVERSITY OF OSLO

# Faculty of Mathematics and Natural Sciences 

Exam in INF5830 - Natural language processing<br>Day of exam: 20 December 2017<br>Exam hours: at 2:30 PM - 4 hours<br>This examination paper consists of 5 pages including this one.<br>Appendices:<br>Statistical formulas - 1 page<br>Statistical table - 4 pages<br>Permitted materials: None

Make sure that your copy of this examination paper is complete before answering.

- You should answer all questions. The weights of the various exercises are indicated.
- You should read through the whole set to see whether anything is unclear so that you can ask your questions to the teachers when they arrive.
- If you think some assumptions are missing, make your own and explain them!


## 1 Experiments (10\%)

What does it mean to perform n-fold cross-validation, e.g. 5-fold crossvaildation experiments? What are the advantages of using cross-validation?

## 2 Evaluation (20\%)

Kim is not satisfied with the automatic language recognizer installed on her PC. It often mistakes Norwegian Bokmål as other languages. Kim therefore implements a classifier and evaluates it on a test set of 900 sentences. This is the result.

|  | Correct class |  |  |
| :--- | ---: | ---: | ---: |
|  | Bokmål | Nynorsk | Danish |
| Assigned Bokmål | 240 | 0 | 10 |
| Assigned Nynorsk | 50 | 300 | 20 |
| Assigned Danish | 10 | 0 | 270 |

(a) What is the accuracy of this classifier?
(b) Kim is mainly interested in whether the classifier recognizes Bokmål correctly or not. What is the accuracy, precision, recall and $F$-score for the Bokmål class on the test set?
(c) Returning to the three-class classifier and the accuracy found in question (a). Assume that the test set is a random sample of sentences from a large population of sentences in the three languages. Estimate a confidence interval for the accuracy at the $95 \%$ confidence level.

## 3 Information extraction (20\%)

(a) What is meant by NP-chunking? Propose an NP-chunk structure for the following sentence. Use parenthetical notation. You do not have to include the POS-tags.

```
[('American', 'NNP'),
('Petrofina', 'NNP'),
('Inc.', 'NNP'),
(',', ','),
('an', 'DT'),
('integrated', 'VBN'),
('oil', 'NN'),
('company', 'NN'),
('based', 'VBD')
('in', 'IN'),
('Dallas', 'NNP'),
(',', ','),
('yesterday', 'NN'),
('said', 'VBD'),
('net', 'JJ'),
('income', 'NN'),
('dropped', 'VBD'),
('to', 'TO'),
('$', '$'),
('15.1', 'CD'),
('million', 'CD'),
(',', ','),
('from', 'IN'),
('$', '$'),
('35.2', 'CD'),
('million', 'CD'),
    ('.', '.')]
```

(b) A popular format for representing chunk structure is the so-called BIO (or IBO) tags. Display the chunk structure from (a) using BIO-tags and a CoNLL-type format. Include the POS-tags.
(c) What is meant by named entity recognition (NER)?
(d) Propose a set of named entity types suitable for the business world with revenues, mergers and acquisitions. Annotate the example sentence with named-entities accordingly by extending the representation from (b).
(e) How is NER performed by supervised machine learning? What kind of features could be useful?

## 4 Data-driven dependency parsing (20\%)

### 4.1 Dependency trees

| ID | FORM | LEMMA | UPOSTAG | XPOSTAG | FEATS | HEAD | DEPREL |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 1 | I | I | PRON | PRP | Case=Nom\|Number=Sing | 3 | nsubj |
| 2 | never | never | ADV | RB | - | 3 | advmod |
| 3 | saw | see | VERB | VBD | Tense=Past\|VerbForm=Fin | 0 | root |
| 4 | anyone | anyone | PRON | NN | Number=Sing | 3 | obj |
| 5 | there | there | ADV | RB | PronType=Dem | 4 | advmod |
| 6 | . | $\cdot$ | PUNCT | . | - | 3 | punct |

(a) Draw the dependency graph for the English sentence 'I never saw anyone there.', provided above in the CoNLL format.
(b) Is this sentence projective? Why or why not?
(c) Which of the dependents of 'saw' are arguments, and which are adjuncts? Why?

### 4.2 Automatic dependency parsing

(a) Recall any two features commonly used in data-driven dependency parsing and use the dependency tree in 4.1 to give an example for each of the features.
(b) Briefly describe the problems with categorical discrete features for dependency parsing that artificial neural networks somewhat helped to alleviate (and how).

## 5 Word sense disambiguation ( $10 \%$ )

(a) Outline the difference between word sense disambiguation and word sense induction.
(b) What machine learning paradigms are usually associated with the former and the latter?

## 6 Semantic role labeling (20\%)

### 6.1 Semantic datasets

| 1 | Apparently | apparently | RB | 4 | adv | - | AM-DIS |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 2 | the | the | DT | 3 | nmod | - | - |
| 3 | commission | commission | NN | 4 | sbj | - | A 0 |
| 4 | did | do | VBD | 0 | root | - | $\bar{A}$ |
| 5 | not | not | RB | 4 | adv | - | $\overline{\text { AM-NEG }}$ |
| 6 | really | really | RB | 4 | adv | - | AM-DIS |
| 7 | believe | believe | VB | 4 | vc | believe.01 | - |
| 8 | in | in | IN | 7 | adv | - | A1 |
| 9 | this | this | DT | 10 | nmod | - | - |
| 10 | ideal | ideal | NN | 8 | pmod | - | - |

(a) The sentence above has been annotated with semantic roles in the CoNLL08 format. Identify the predicate of this sentence and its semantic arguments along with their roles.
(b) Provide a short description for each of the roles of the core arguments in this sentence.
(c) Invent two English sentences describing one event with different dependency trees but identical semantic role structures. Briefly explain the discrepancies between syntax and semantics in them.

### 6.2 Automatic SRL

(a) Outline the general workflow of a machine learning based SRL system taking raw text as an input.
(b) Choose any non-core argument from the sentence in 6.1. Give examples of features which can be helpful in correctly identifying its role.

## INF5830, 2015, some statistical formulas

## Z-score

Given a normal distribution with mean $\mu$ and standard deviation $\sigma$. The $Z$-score of a data point $x$

$$
Z=\frac{x-\mu}{\sigma}
$$

expresses the distance of $x$ from $\mu$ in terms of standard deviations.

## t-test

The t-statistics

$$
t=\frac{\bar{x}-\mu}{\sqrt{\frac{s^{2}}{n}}}
$$

where

- $\bar{x}$ is the mean of a simple random sample
- $n$ is the size of the sample
- $s$ is the sample standard deviation


## Two sample t-test

$$
t=\frac{\bar{x}_{1}-\bar{x}_{2}}{\sqrt{\frac{s_{1}^{2}}{n_{1}}+\frac{s_{2}^{2}}{n_{2}}}}
$$

## Standard deviation of proportion

When $p$ is a proportion $\frac{k}{n}$ ( $k$ successes out of $n$ ), the variance is

$$
p(1-p)
$$

# STATISTICAL TABLES 

Cumulative normal distribution<br>Critical values of the $t$ distribution<br>Critical values of the $F$ distribution<br>Critical values of the chi-squared distribution

Table A. 1

## Cumulative Standardized Normal Distribution


$A(z)$ is the integral of the standardized normal distribution from $-\infty$ to $z$ (in other words, the area under the curve to the left of $z$ ). It gives the probability of a normal random variable not being more than $z$ standard deviations above its mean. Values of $z$ of particular importance:

| $z$ | $A(z)$ |  |
| :---: | :---: | :--- |
| 1.645 | 0.9500 | Lower limit of right $5 \%$ tail |
| 1.960 | 0.9750 | Lower limit of right $2.5 \%$ tail |
| 2.326 | 0.9900 | Lower limit of right $1 \%$ tail |
| 2.576 | 0.9950 | Lower limit of right $0.5 \%$ tail |
| 3.090 | 0.9990 | Lower limit of right $0.1 \%$ tail |
| 3.291 | 0.9995 | Lower limit of right $0.05 \%$ tail |


| $z$ | 0.00 | 0.01 | 0.02 | 0.03 | 0.04 | 0.05 | 0.06 | 0.07 | 0.08 | 0.09 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0.0 | 0.5000 | 0.5040 | 0.5080 | 0.5120 | 0.5160 | 0.5199 | 0.5239 | 0.5279 | 0.5319 | 0.5359 |
| 0.1 | 0.5398 | 0.5438 | 0.5478 | 0.5517 | 0.5557 | 0.5596 | 0.5636 | 0.5675 | 0.5714 | 0.5753 |
| 0.2 | 0.5793 | 0.5832 | 0.5871 | 0.5910 | 0.5948 | 0.5987 | 0.6026 | 0.6064 | 0.6103 | 0.6141 |
| 0.3 | 0.6179 | 0.6217 | 0.6255 | 0.6293 | 0.6331 | 0.6368 | 0.6406 | 0.6443 | 0.6480 | 0.6517 |
| 0.4 | 0.6554 | 0.6591 | 0.6628 | 0.6664 | 0.6700 | 0.6736 | 0.6772 | 0.6808 | 0.6844 | 0.6879 |
| 0.5 | 0.6915 | 0.6950 | 0.6985 | 0.7019 | 0.7054 | 0.7088 | 0.7123 | 0.7157 | 0.7190 | 0.7224 |
| 0.6 | 0.7257 | 0.7291 | 0.7324 | 0.7357 | 0.7389 | 0.7422 | 0.7454 | 0.7486 | 0.7517 | 0.7549 |
| 0.7 | 0.7580 | 0.7611 | 0.7642 | 0.7673 | 0.7704 | 0.7734 | 0.7764 | 0.7794 | 0.7823 | 0.7852 |
| 0.8 | 0.7881 | 0.7910 | 0.7939 | 0.7967 | 0.7995 | 0.8023 | 0.8051 | 0.8078 | 0.8106 | 0.8133 |
| 0.9 | 0.8159 | 0.8186 | 0.8212 | 0.8238 | 0.8264 | 0.8289 | 0.8315 | 0.8340 | 0.8365 | 0.8389 |
| 1.0 | 0.8413 | 0.8438 | 0.8461 | 0.8485 | 0.8508 | 0.8531 | 0.8554 | 0.8577 | 0.8599 | 0.8621 |
| 1.1 | 0.8643 | 0.8665 | 0.8686 | 0.8708 | 0.8729 | 0.8749 | 0.8770 | 0.8790 | 0.8810 | 0.8830 |
| 1.2 | 0.8849 | 0.8869 | 0.8888 | 0.8907 | 0.8925 | 0.8944 | 0.8962 | 0.8980 | 0.8997 | 0.9015 |
| 1.3 | 0.9032 | 0.9049 | 0.9066 | 0.9082 | 0.9099 | 0.9115 | 0.9131 | 0.9147 | 0.9162 | 0.9177 |
| 1.4 | 0.9192 | 0.9207 | 0.9222 | 0.9236 | 0.9251 | 0.9265 | 0.9279 | 0.9292 | 0.9306 | 0.9319 |
| 1.5 | 0.9332 | 0.9345 | 0.9357 | 0.9370 | 0.9382 | 0.9394 | 0.9406 | 0.9418 | 0.9429 | 0.9441 |
| 1.6 | 0.9452 | 0.9463 | 0.9474 | 0.9484 | 0.9495 | 0.9505 | 0.9515 | 0.9525 | 0.9535 | 0.9545 |
| 1.7 | 0.9554 | 0.9564 | 0.9573 | 0.9582 | 0.9591 | 0.9599 | 0.9608 | 0.9616 | 0.9625 | 0.9633 |
| 1.8 | 0.9641 | 0.9649 | 0.9656 | 0.9664 | 0.9671 | 0.9678 | 0.9686 | 0.9693 | 0.9699 | 0.9706 |
| 1.9 | 0.9713 | 0.9719 | 0.9726 | 0.9732 | 0.9738 | 0.9744 | 0.9750 | 0.9756 | 0.9761 | 0.9767 |
| 2.0 | 0.9772 | 0.9778 | 0.9783 | 0.9788 | 0.9793 | 0.9798 | 0.9803 | 0.9808 | 0.9812 | 0.9817 |
| 2.1 | 0.9821 | 0.9826 | 0.9830 | 0.9834 | 0.9838 | 0.9842 | 0.9846 | 0.9850 | 0.9854 | 0.9857 |
| 2.2 | 0.9861 | 0.9864 | 0.9868 | 0.9871 | 0.9875 | 0.9878 | 0.9881 | 0.9884 | 0.9887 | 0.9890 |
| 2.3 | 0.9893 | 0.9896 | 0.9898 | 0.9901 | 0.9904 | 0.9906 | 0.9909 | 0.9911 | 0.9913 | 0.9916 |
| 2.4 | 0.9918 | 0.9920 | 0.9922 | 0.9925 | 0.9927 | 0.9929 | 0.9931 | 0.9932 | 0.9934 | 0.9936 |
| 2.5 | 0.9938 | 0.9940 | 0.9941 | 0.9943 | 0.9945 | 0.9946 | 0.9948 | 0.9949 | 0.9951 | 0.9952 |
| 2.6 | 0.9953 | 0.9955 | 0.9956 | 0.9957 | 0.9959 | 0.9960 | 0.9961 | 0.9962 | 0.9963 | 0.9964 |
| 2.7 | 0.9965 | 0.9966 | 0.9967 | 0.9968 | 0.9969 | 0.9970 | 0.9971 | 0.9972 | 0.9973 | 0.9974 |
| 2.8 | 0.9974 | 0.9975 | 0.9976 | 0.9977 | 0.9977 | 0.9978 | 0.9979 | 0.9979 | 0.9980 | 0.9981 |
| 2.9 | 0.9981 | 0.9982 | 0.9982 | 0.9983 | 0.9984 | 0.9984 | 0.9985 | 0.9985 | 0.9986 | 0.9986 |
| 3.0 | 0.9987 | 0.9987 | 0.9987 | 0.9988 | 0.9988 | 0.9989 | 0.9989 | 0.9989 | 0.9990 | 0.9990 |
| 3.1 | 0.9990 | 0.9991 | 0.9991 | 0.9991 | 0.9992 | 0.9992 | 0.9992 | 0.9992 | 0.9993 | 0.9993 |
| 3.2 | 0.9993 | 0.9993 | 0.9994 | 0.9994 | 0.9994 | 0.9994 | 0.9994 | 0.9995 | 0.9995 | 0.9995 |
| 3.3 | 0.9995 | 0.9995 | 0.9995 | 0.9996 | 0.9996 | 0.9996 | 0.9996 | 0.9996 | 0.9996 | 0.9997 |
| 3.4 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9997 | 0.9998 |
| 3.5 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 | 0.9998 |
| 3.6 | 0.9998 | 0.9998 | 0.9999 |  |  |  |  |  |  |  |

Table A. 2
$t$ Distribution: Critical Values of $t$

|  |  | Significance level |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Degrees of freedom | Two-tailed test: One-tailed test: | $\begin{aligned} & 10 \% \\ & 5 \% \end{aligned}$ | $\begin{aligned} & 5 \% \\ & 2.5 \% \end{aligned}$ | $\begin{aligned} & 2 \% \\ & 1 \% \end{aligned}$ | $\begin{aligned} & 1 \% \\ & 0.5 \% \end{aligned}$ | $\begin{aligned} & 0.2 \% \\ & 0.1 \% \end{aligned}$ | $\begin{aligned} & 0.1 \% \\ & 0.05 \% \end{aligned}$ |
| 1 |  | 6.314 | 12.706 | 31.821 | 63.657 | 318.309 | 636.619 |
| 2 |  | 2.920 | 4.303 | 6.965 | 9.925 | 22.327 | 31.599 |
| 3 |  | 2.353 | 3.182 | 4.541 | 5.841 | 10.215 | 12.924 |
| 4 |  | 2.132 | 2.776 | 3.747 | 4.604 | 7.173 | 8.610 |
| 5 |  | 2.015 | 2.571 | 3.365 | 4.032 | 5.893 | 6.869 |
| 6 |  | 1.943 | 2.447 | 3.143 | 3.707 | 5.208 | 5.959 |
| 7 |  | 1.894 | 2.365 | 2.998 | 3.499 | 4.785 | 5.408 |
| 8 |  | 1.860 | 2.306 | 2.896 | 3.355 | 4.501 | 5.041 |
| 9 |  | 1.833 | 2.262 | 2.821 | 3.250 | 4.297 | 4.781 |
| 10 |  | 1.812 | 2.228 | 2.764 | 3.169 | 4.144 | 4.587 |
| 11 |  | 1.796 | 2.201 | 2.718 | 3.106 | 4.025 | 4.437 |
| 12 |  | 1.782 | 2.179 | 2.681 | 3.055 | 3.930 | 4.318 |
| 13 |  | 1.771 | 2.160 | 2.650 | 3.012 | 3.852 | 4.221 |
| 14 |  | 1.761 | 2.145 | 2.624 | 2.977 | 3.787 | 4.140 |
| 15 |  | 1.753 | 2.131 | 2.602 | 2.947 | 3.733 | 4.073 |
| 16 |  | 1.746 | 2.120 | 2.583 | 2.921 | 3.686 | 4.015 |
| 17 |  | 1.740 | 2.110 | 2.567 | 2.898 | 3.646 | 3.965 |
| 18 |  | 1.734 | 2.101 | 2.552 | 2.878 | 3.610 | 3.922 |
| 19 |  | 1.729 | 2.093 | 2.539 | 2.861 | 3.579 | 3.883 |
| 20 |  | 1.725 | 2.086 | 2.528 | 2.845 | 3.552 | 3.850 |
| 21 |  | 1.721 | 2.080 | 2.518 | 2.831 | 3.527 | 3.819 |
| 22 |  | 1.717 | 2.074 | 2.508 | 2.819 | 3.505 | 3.792 |
| 23 |  | 1.714 | 2.069 | 2.500 | 2.807 | 3.485 | 3.768 |
| 24 |  | 1.711 | 2.064 | 2.492 | 2.797 | 3.467 | 3.745 |
| 25 |  | 1.708 | 2.060 | 2.485 | 2.787 | 3.450 | 3.725 |
| 26 |  | 1.706 | 2.056 | 2.479 | 2.779 | 3.435 | 3.707 |
| 27 |  | 1.703 | 2.052 | 2.473 | 2.771 | 3.421 | 3.690 |
| 28 |  | 1.701 | 2.048 | 2.467 | 2.763 | 3.408 | 3.674 |
| 29 |  | 1.699 | 2.045 | 2.462 | 2.756 | 3.396 | 3.659 |
| 30 |  | 1.697 | 2.042 | 2.457 | 2.750 | 3.385 | 3.646 |
| 32 |  | 1.694 | 2.037 | 2.449 | 2.738 | 3.365 | 3.622 |
| 34 |  | 1.691 | 2.032 | 2.441 | 2.728 | 3.348 | 3.601 |
| 36 |  | 1.688 | 2.028 | 2.434 | 2.719 | 3.333 | 3.582 |
| 38 |  | 1.686 | 2.024 | 2.429 | 2.712 | 3.319 | 3.566 |
| 40 |  | 1.684 | 2.021 | 2.423 | 2.704 | 3.307 | 3.551 |
| 42 |  | 1.682 | 2.018 | 2.418 | 2.698 | 3.296 | 3.538 |
| 44 |  | 1.680 | 2.015 | 2.414 | 2.692 | 3.286 | 3.526 |
| 46 |  | 1.679 | 2.013 | 2.410 | 2.687 | 3.277 | 3.515 |
| 48 |  | 1.677 | 2.011 | 2.407 | 2.682 | 3.269 | 3.505 |
| 50 |  | 1.676 | 2.009 | 2.403 | 2.678 | 3.261 | 3.496 |
| 60 |  | 1.671 | 2.000 | 2.390 | 2.660 | 3.232 | 3.460 |
| 70 |  | 1.667 | 1.994 | 2.381 | 2.648 | 3.211 | 3.435 |
| 80 |  | 1.664 | 1.990 | 2.374 | 2.639 | 3.195 | 3.416 |
| 90 |  | 1.662 | 1.987 | 2.368 | 2.632 | 3.183 | 3.402 |
| 100 |  | 1.660 | 1.984 | 2.364 | 2.626 | 3.174 | 3.390 |
| 120 |  | 1.658 | 1.980 | 2.358 | 2.617 | 3.160 | 3.373 |
| 150 |  | 1.655 | 1.976 | 2.351 | 2.609 | 3.145 | 3.357 |
| 200 |  | 1.653 | 1.972 | 2.345 | 2.601 | 3.131 | 3.340 |
| 300 |  | 1.650 | 1.968 | 2.339 | 2.592 | 3.118 | 3.323 |
| 400 |  | 1.649 | 1.966 | 2.336 | 2.588 | 3.111 | 3.315 |
| 500 |  | 1.648 | 1.965 | 2.334 | 2.586 | 3.107 | 3.310 |
| 600 |  | 1.647 | 1.964 | 2.333 | 2.584 | 3.104 | 3.307 |
| $\infty$ |  | 1.645 | 1.960 | 2.326 | 2.576 | 3.090 | 3.291 |

## Table A. 4

$\chi^{2}$ (Chi-Squared) Distribution: Critical Values of $\chi^{2}$
Significance level

| Degrees of <br> freedom | $5 \%$ | $1 \%$ | $0.1 \%$ |
| :---: | ---: | ---: | ---: |
| $\mathbf{1}$ | 3.841 | 6.635 | 10.828 |
| $\mathbf{2}$ | 5.991 | 9.210 | 13.816 |
| $\mathbf{3}$ | 7.815 | 11.345 | 16.266 |
| $\mathbf{4}$ | 9.488 | 13.277 | 18.467 |
| $\mathbf{5}$ | 11.070 | 15.086 | 20.515 |
| $\mathbf{6}$ | 12.592 | 16.812 | 22.458 |
| $\mathbf{7}$ | 14.067 | 18.475 | 24.322 |
| $\mathbf{8}$ | 15.507 | 20.090 | 26.124 |
| $\mathbf{9}$ | 16.919 | 21.666 | 27.877 |
| $\mathbf{1 0}$ | 18.307 | 23.209 | 29.588 |

