## UNIVERSITY OF OSLO

# Faculty of Mathematics and Natural Sciences 

Exam in INF5830 - Natural language processing<br>Day of exam: 15 June 2015<br>Exam hours: at 14:30-4 hours<br>This examination paper consists of 3 pages including this.<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 may answer in English, Norwegian, Danish or Swedish.
- You should answer all questions. The weight of the various questions 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 Dependency syntax and parsing (25 \%)

(a) Describe the formal conditions on well-formed dependency graphs.
(b) Consider the dependency graph for the English sentence He heard a song today that he disliked:


Is the graph above projective? Why or why not?
(c) What are the main differences between graph-based and transition-based approaches to data-driven dependency parsing? What type of parser is Maltparser?

## 2 Semantic role labeling (25 \%)

(a) Provide three roles from the traditional set of semantic roles along with a short description. Apply these to the sentence from exercise 1.
(b) Consider the following frame taken from FrameNet:

| Definition: |  |
| :--- | :--- |
| An Agent cuts an Item into Pieces using an Instrument: |  |
| Frame Elements (core): |  |
| Agent | person cutting the Item |
| Item | item being cut |
| Pieces | parts of the original Item |
| Frame Elements (non-core): |  |
| Instrument | instrument with which the Item is cut |
| Manner | manner in which Item is cut |
| Place | where cutting takes place <br> Purpose |
| pexical Units: | carve, chor cutting |
|  | fillet, mince, pare, cut, slice |

(i) Briefly describe the notion of a frame and the different frame components above.
(ii) In what way is Fillmore's frame semantics a response to criticism of the traditional set of semantic roles?
(c) In Semantic Role Labeling (SRL), parsing is often used as a pre-processing step for feature construction. Provide two examples of features commonly used in SRL that require syntactic parsing and use the dependency graph above to give one example for each of the features you describe.

## 3 Estimation (15\%)

Kim has constructed a classifier and is testing it on a test set with 400 items. It achieves an accuracy of 0.64 . Estimate a confidence interval for the accuracy of the classifier at level 0.99.
(A little help since you do not have a computer at hand: $\sqrt{0.64 \times 0.36}=0.48$ )

## 4 Collocations (20\%)

Explain what a collocation is. There have been proposed several different measures for the association between the two words in a collocation. Describe two or three of them.

## 5 Machine learning (15\%)

Multinomial logistic regression is a method for classification which is also called Maximum entropy classification. It is similar to Naive Bayes classification and can in some respects be considered a refinement of Naive Bayes classification. When trained on the same training material with the same set of features, the Logistic regression classifier should in principle perform at least as well as the Naive Bayes classifier when measured on the training material, and sometimes it will perform better. Without going deeply into the mathematical details, explain why this is so.

Also, explain why adding more features may deteriorate the results of the Naive Bayes classifier, but not of the Logistic regression classifier.

In spite of this, Jurafsky and Martin claim "Furthermore, naive Bayes works extremely well (even better than logistic regression or SVMs) on small datasets or short documents." Why do you think that is the case?

END

## 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 |

