

**KEN3450: Data and Image Analysis**  
**Final Exam (7/4/2017)**

Your name: \_\_\_\_\_ **KEY** \_\_\_\_\_

Your student number: \_\_\_\_\_

Expected grade (optional): \_\_\_\_ / 80

Part	Points	/max
1		/13
2		/14
3		/15
4		/6
5		/16
6		/16
B		/1
<b>Total</b>		<b>/81</b>

**General remarks**

- This exam consists of 13 pages and 6 parts (and 1 bonus question). Each part indicates a number of points. These points are indicative of how important the questions are and will be used proportionally for grading.
- If you multiply points by two you get how much time (in min. approx.) you have to spend on each question (e.g. 5 points, means 10 minutes).
- The exam contributes 80% to your final grade (data clinics is 10% and data madness is 10%).
- The answer boxes are small on purpose: You should try and give short answers that are to the point.
- If you need "experimental space" just use the back (white) pages.
- While you shouldn't spend too much time on calligraphy, please make sure I will have at least a chance deciphering your exam

## Warm up with EDA! (13 points)

(7) a. The Board of Examiners of "Super University" collects information about how much time students spend in toilet during exams. Here are the results of some statistics:

mean	7,39 minutes
st.dev.	3,05 minutes
min	3
Q1	3
median	6
Q3	11
max	16

(2) i. Notice that the 1<sup>st</sup> quartile and minimum number of minutes are the same. Explain how this can be.

(2) ii. Were there any outliers? Explain how you made your decision.

(3) iii. The rules and regulations of Super University declare that students that stay in toilet more than 15 minutes are "suspect for fraud". Would you consider 15 minutes too much? Explain.

i) The lower 25% are 3

e.g.  $\underbrace{3, 3, 3, 3}_{25\%}, 6, 6, 6, 6, 6, 6, 6, 6, 11, 11, 11, 16$   
 $\parallel$   
 Q1

(ii) I need to compute the fences of boxplot:

$$IQR = Q3 - Q1 = 11 - 3 = 8$$

$$\text{Upper Fence: } Q3 + 1.5IQR = 11 + \frac{3}{2} \times 8 = 23$$

$$\text{Lower Fence: } Q1 - 1.5IQR = 3 - \frac{3}{2} \times 8 = -9$$

Nothing is  $< -9$  or  $> 23$ , so no outliers.

(iii) Compute Z-score:

$$Z = \frac{15 - \mu}{\sigma} = \frac{15 - 7.39}{3.05} = 2.5 \stackrel{> 1}{\sim} \text{kind of rare}$$

d. For the following questions just put in a circle the correct answer. Only if you have doubts, justify your choice in the space provided, otherwise it's perfectly fine to leave it empty.

(2) i) "If  $r = -0.4$  for the relationship between the time of day and amount of coffee in an office worker's mug, which are true?"

- I.  $r^2$  (R-squared) = - 16%
- II. There is a linear relationship between time and amount of coffee.
- III. In 16% of the cases, amount of coffee in the mug is correctly predicted by time of day.

A) I (very short) explanation:

B) II

C) III

D) II and III only

E) none of these

(2) ii) "The auto insurance industry crashed some test vehicles into a cement barrier at speeds of 5 to 25 mph to investigate the amount of damage to the cars. They found a correlation of  $r = 0.60$  between speed (MPH) and damage (\$). If the speed at which a car hit the barrier is 1.5 standard deviations above the mean speed, we expect the damage to be \_\_?\_\_ the mean damage."

A) equal to (very short) explanation:

B) 0.36 SD above

C) 0.60 SD above

D) 0.90 SD above

E) 1.5 SD above

(2) iii) "The correlation between  $X$  and  $Y$  is  $r = 0.35$ . If we double each  $X$  value, decrease each  $Y$  by 0.20, and interchange the variables (put  $X$  on the  $Y$ -axis and vice versa), the new correlation"

A) is 0.35 (very short) explanation:

B) is 0.50

C) is 0.70

D) is 0.90

E) cannot be determined

## The key to happiness is text analysis (14 points)

Consider the following documents about what is making people happy, where we already applied some pre-processing and got rid of some of the words:

Doc1	phone ring person happy person
Doc2	dog pet happy run jump
Doc3	cat purr pet person happy
Doc4	life smile running happy
Doc5	life laugh walk run run happy

(5) a. Construct the document-by-term matrix (DTM) for this collection and report its dimensions and its sparsity level (%). As a weighting method use the simple counts for each word (i.e. how many times each word appears in each document)

	phone	ring	person	happy	dog	pet	run	jump	cat	purr	life	smile	running	laugh	walk
doc1	1	1	2	1											
doc2			1	1	1	1	1	1							
doc3			1	1		1			1	1					
doc4				1							1	1	1		
doc5				1			2				1			1	1

DTM:  $[5 \times 15]$  : sparsity:  $\frac{52}{75} \approx 69\%$  sparse

(3) b. Justify (by using your computed TDM) why we are also using the Inverse Document Frequency (IDF) as a weighting technique. Compute IDF for terms: **person**, **happy**, **life**, **purr**. Can you intuitively explain the result?

e.g. happy appears in all documents, no discriminative ability

$$IDF_{\text{person}} = \log_{10} \frac{5}{2} = 0.398$$

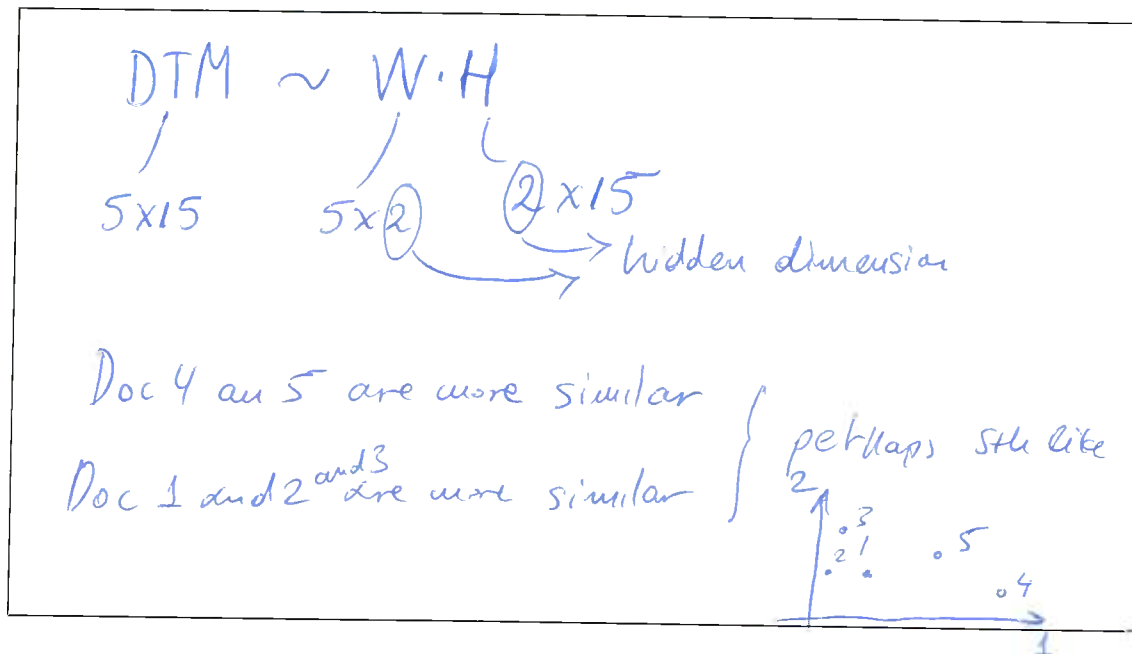
$$IDF_{\text{life}} = \log_{10} \frac{5}{2} = 0.398$$

$$IDF_{\text{happy}} = \log_{10} \frac{5}{5} = 0$$

$$IDF_{\text{purr}} = \log_{10} \frac{5}{1} = 0.699$$

Happy (appears in all docs) gets 0, purr appears in 1 gets the highest

If we were to perform a dimensionality reduction on the DTM matrix (in order to get down to 2 dimensions), show how Non-Negative Matrix Factorization would be applied (no need to compute it, but present how NMF will work, the intuition and the result we get). Can you intuitively predict how documents will be spread across the two "hidden" dimensions by looking at the DTM?



(3) d. Why do we say that the TF-IDF/Vector Space model ignores "semantic" information? Use the DTM matrix you computed to motivate your answer.

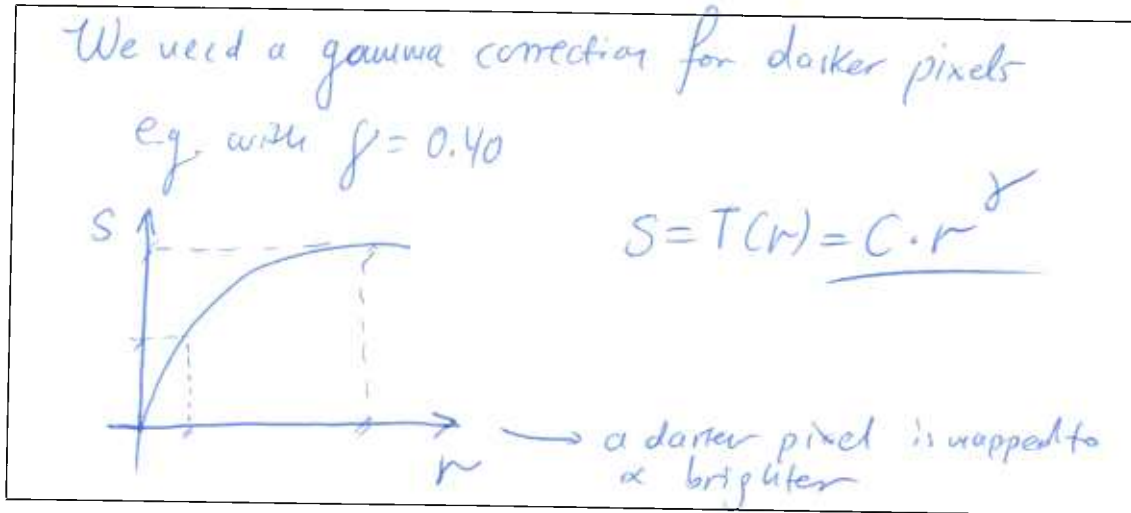
e.g. "run", "running" are different components

similarly: "smile", "laugh" are related but also different components

Mention that VSM assumes orthogonal dimensions

### 3 Understanding images (15 points)

(5) a. If you attended one or more lectures (or the Data Madness), you must have realized that the beamer displayed everything "a bit" darker. What kind of pixel transformation would you use in order to restore the brightness of the display? Sketch your solution with an explanatory figure.



(6) b. For the two filters below, describe what is their functionality (approximately). Explain your reasoning.

$$H_1 = \frac{1}{4} \begin{bmatrix} -1 & -2 & -1 \\ -2 & 16 & -2 \\ -1 & -2 & -1 \end{bmatrix}, \quad H_2 = \begin{bmatrix} -1 & -3 & -1 \\ 0 & 0 & 0 \\ 1 & 3 & 1 \end{bmatrix}$$

$H_1$  : - sum of coefficients is 1 (if it was 0, it would be "blur")

- I also see negative values which means that I compute differences (thus details)

It's a combination of details & averaging

$H_2$  : Rough edge detection towards ~~X~~ direction.

Looks like Sobel but looks for larger differences

(4) c. Associate the images (left) with their corresponding Fourier Transform (FFT). Justify your choices by explaining the role of the filter applied (if any).



FFT no 2



1

*low-pass*



FFT no 4

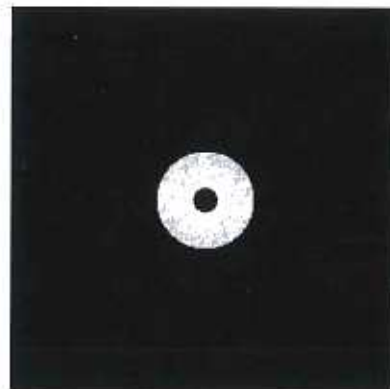


2

*no filter*



FFT no 1

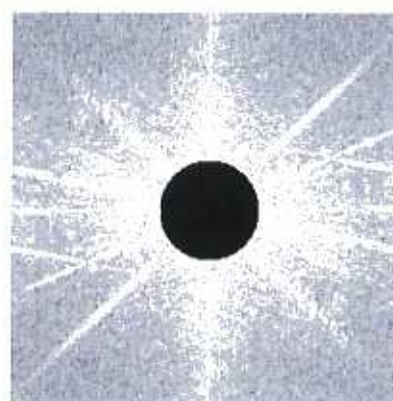


3

*band-pass*



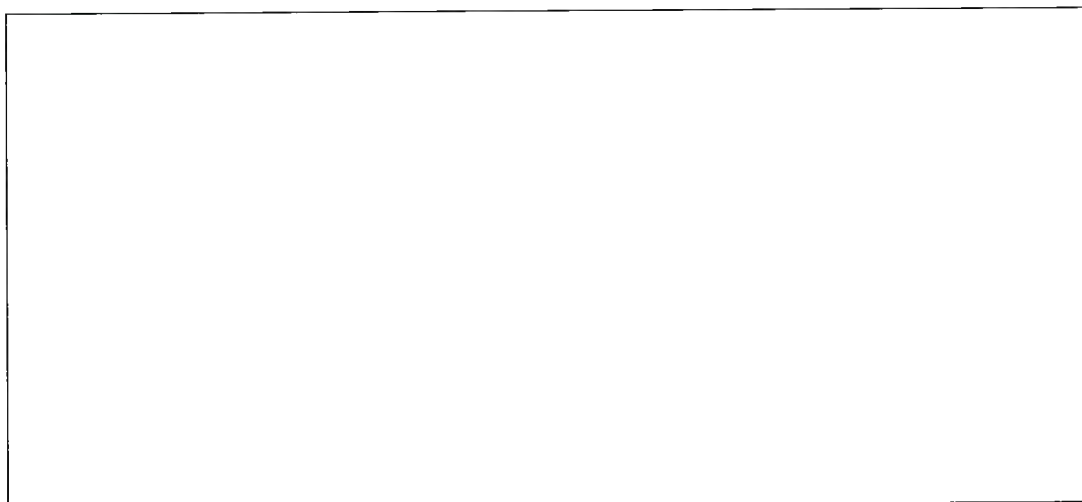
FFT no 3



4

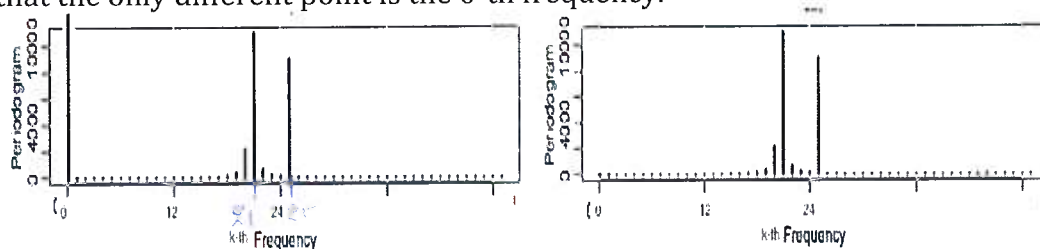
*high-pass*





#### 4 Looking at the stars (6 points)

Two data scientists (really interested in astronomy), computed the periodogram of a time series which shows the magnitude of a star taken at midnight for 600 consecutive days. They came up with the following figures where you can see that the only different point is the 0-th frequency.



- (1) a. Explain why there is this slight difference in the form of the periodogram and relate that to the form of the original timeseries (how you expect it to be).  
 (5) b. Compute the two most important critical periods in the magnitude of the star (excluding the 0-th frequency).

a) Remove the trend/mean.  
 DC (0-th) corresponds to mean. I expect original time series to have a "trend"/mean

b)  $k=21 \rightarrow \frac{21}{600} = 0.035 \text{ cycles/period} \text{ or } 29 \text{ days}$   
 $k=25 \rightarrow \frac{25}{600} \approx 0.042 \text{ cycles/period} \text{ or } 24 \text{ days}$



## 5 Modeling skiing (16 points)

I have never been skiing and I would love to try it but I am also not fond of crowded places, so I decided to build a model in order to estimate when a small ski resort near Maastricht is crowded. Since I am also very busy, I assigned this problem to some young hard-working students. They collected data from the past two ski seasons, measuring different variables and then they started fighting with each other because everyone claimed to have the best model.

Let's see the solution of a student that came to me having fit a model using the following variables:

**Skiers** the number of skiers who visit the resort on that day.  
**Snow** the number of inches of snow on the ground.  
**Temp** the high temperature for the day in degrees F.  
**Weekend** an indicator variable, if it's weekend (YES/NO).

Variable	Coefficient	SE (Coeff.)	t-ratio	p-value
Intercept	559.869	76.78	7.29	<0.0001
Snow	1.424	2.70	0.53	0.6019
Temp	-1.604	2.77	-0.58	0.5677
Weekend: YES	147.349	51.86	2.84	0.0086

(2) a. What is the predicted number of skiers for a Saturday with a temperature of 40 F and a snow cover of 25 inches? What would change if it was a Friday?

model is:

$$\hat{y} = 559.869 + 1.424 \times \text{Snow} - 1.604 \times \text{Temp} + 147.349 \times \text{weekend}$$

$$\hat{y}_1 = 559.869 + 1.424 \times 25 - 1.604 \times 40 + 147.349 \approx 679 \text{ people}$$

$$\hat{y}_2 \approx 531 \text{ people (only the 147.349 changes)}$$

(2) b. Which of the explanatory variables appear to be associated with the number of skiers and which do not? Explain how you reached your conclusion.

only the weekend seems to have a significant effect

(2) c. Compute a 95% confidence interval (assume for simplicity that  $t=2$ , i.e. simply compute the intervals as for simple regression) for the slope of the variable **weekend** and explain the meaning of this interval in the context of this specific problem.

$$147.349 \pm 2 \times 51.86 \rightarrow [44, 251]$$

Given fixed values for snow/temperature  
people on weekends will vary from 44 to 251

(3) d. One of the students is claiming that, given the above results, if we run a single regression where **skiers** remains the dependent variable and **snow** is the only independent variable, we will get a similar result in terms of significance of the coefficient. Do you agree? Why yes/no?

It seems that snow is not significant in the presence  
of weekend (perhaps it "sucks" the effect of snow)  
But if we take the effect of snow by itself  
the result might be different

(3) e. If you think that the temperature might affect ski attendance differently on weekends than on weekdays, how would you change the regression to test this? Sketch how the regression equation would look like.

Introduce an interaction term  
Weekend  $\times$  temp

$$\hat{y} = d_0 + d_1 \text{temp} + d_2 \text{snow} + d_3 \text{weekend} + d_4 \text{weekend} \times \text{temp}$$

(4) f. Following my advice to fit simple models, two students selected only one variable (**snow**), however they had different opinions on how they should fit the model. Here are the two models they presented:

A:  $y = w_1^2 x + w_2 x$

B:  $y = wx$

Note that model A now uses two parameters (though both multiply with the same input value  $x$ ). Which of the following is correct?

- I) A will perform better than B -most of the times-
- II) B will perform better than A -most of the times-
- III) They would perform equally well on all cases

*Hint: Think of how regression coefficients are estimated.*

Least squares estimation:

$$\text{minimize } \sum_i (\hat{y}_i - y_i)^2 \text{ or } \sum_i (w_1^2 x_i + w_2 x_i - y_i)^2$$

$$(w_1^2 + w_2) \cdot \sum_i (x_i - y_i)$$

In the end  $w_1^2 + w_2 = w$

so it's the same model

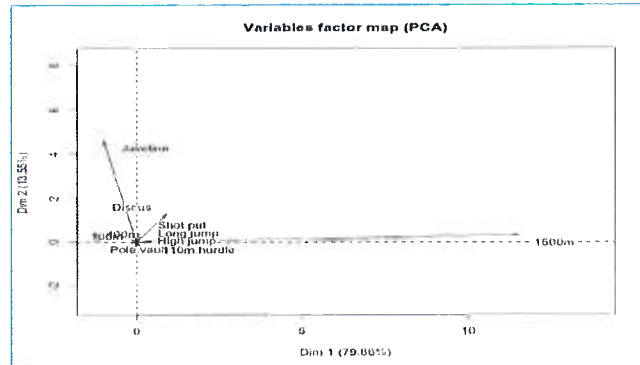
## 6 Is Decathlon really 10 events? (15 points)

(9) a. Decathlon (from the greek word "δέκα" which means "ten") consists of 10 events (namely: 100m, Long Jump, Shot Put, High Jump, 400m, 110m Hurdles, Discus Throw, Pole Vault, Javeline Throw, 1500m). A sports bet company is looking into some data from some athletes in order to reveal which events/sports are most significant.

The chief data scientist of the company (after some googling) decided to run a Principal Component Analysis (PCA), however has no idea on how to interpret the results. The dataset consists of 41 athletes and the performance in 10 sports (see a part of it below).

	100m	Long.jump	Shot.put	High.jump	400m	110m.hurdle	Discus	Pole.vault	Javeline	1500m
SEBRLE	11.04	7.58	14.83	2.07	49.81	14.69	43.75	5.02	63.19	291.7
CLAY	10.76	7.40	14.26	1.86	49.37	14.05	50.72	4.92	60.15	301.5
KARPOV	11.02	7.30	14.77	2.04	48.37	14.09	48.95	4.92	50.31	300.2
BERNARD	11.02	7.23	14.25	1.92	48.93	14.99	40.87	5.32	62.77	280.1
YURKOV	11.34	7.09	15.19	2.10	50.42	15.31	46.26	4.72	63.44	276.4
WARNERS	11.11	7.60	14.31	1.98	48.68	14.23	41.10	4.92	51.77	278.1

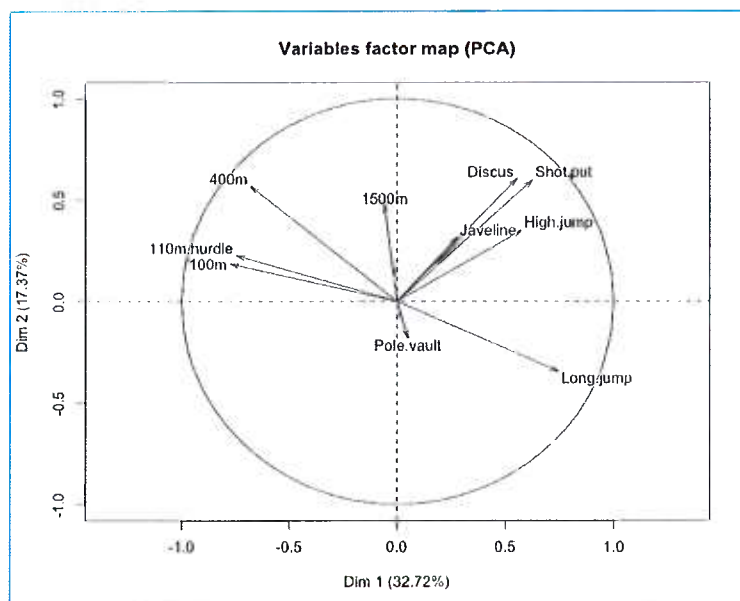
(3) i) The data scientist ran a PCA and by plotting the first 2 PCs got the result on the right. Can you explain the mistake here, why the figure has this form and how you would correct it? Use the data sample above to make assumptions for the data.



- No scaling of the variables
- The 1st dimension is dominated by 1500m which has the largest variance scale
- Rescaling/normalizing variables

(7) ii) A more reasonable PCA plot using the two first components appears on the right. Answer the following questions.

- Is in this case, is the use of two PCs enough to explain the data? Why yes/no?
- Given the loadings of the figure, what is the 1<sup>st</sup> and 2<sup>nd</sup> PC are capturing/ explaining? Can we somehow differentiate between characteristics of different sports? Provide an intuitive description of the above PCA plot.



- First 2 PCs explain around 50%, so it's not enough
- 1st PC  $\sim$  jump
- 2nd PC  $\sim$  running (100m/150m) - } filed & track events

(6) b, We discussed in class that regularization can be applied to any "learning" algorithm in order to avoid overfitting and improve generalization over unseen data. Thinking about Non-Negative Matrix Factorization (NMF), how would you formulate the optimization problem in order to account for regularization? Provide the adjusted optimization equation and the interpretation of parameter shrinking in this case. If we prefer a non-sparse decomposition, would you prefer Ridge (L2) or Lasso (L1) regularization in this case?

normal NMF : minimize  $\|A - WH\|$

parameters are the actual coefficients

so : minimize  $\|A - WH\| + \lambda \cdot \|u\|$  Lasso

or  $\lambda \cdot \|u\|^2$  Ridge

where  $\|u\|$  is a vector with all parameters of  $WH$  matrices

Lasso yields many 0s, so go with Ridge

### Bonus question (1 point)

How old is Jerry?

Answer: \_\_\_\_\_