

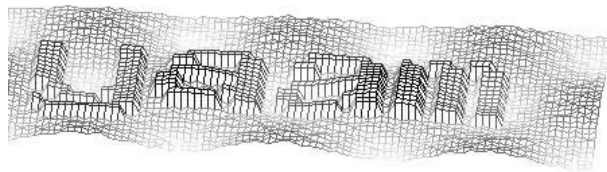
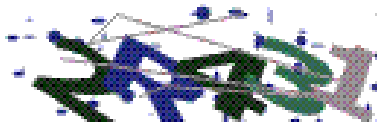
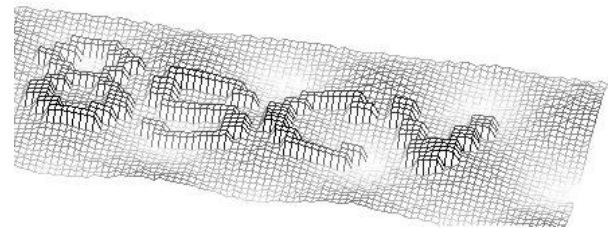
# Recognizing some of the modern CAPTCHAs

The background of the slide is black. In the center, there is a faint, dark grey world map. At the bottom, there are several horizontal, wavy lines in shades of orange and yellow, creating a gradient effect. Scattered across the slide are several circular icons with orange borders. One icon on the left contains a white arrow pointing up and to the right. Three other icons on the right side contain white arrows pointing straight up.

Dmitry Nikulin

LCME, Saint-Petersburg, 2011

# Examples





# Stands for

- Completely
- Automated
- Public
- Turing test to tell
- Computers and
- Humans
- Apart



# Turing test

- Introduced by the mathematician Alan Turing in 1950
- Aimed to distinguish between a machine and a human
- The classic version is carried out by a human
- Loebner Prize has not been won yet

# Reverse Turing Test

- Carried out by a computer
- A widespread example is CAPTCHA
  - Checks for human presence
  - Protects against spam and automated registrations
  - Uses human ability to recognize distorted text (Google reCAPTCHA)

# Requirements for a CAPTCHA

- Simple for a human
- Difficult for a machine
- Does not require large computational resources

Let us call a CAPTCHA efficient if a machine can successfully bypass it in no more than 1% of attempts.

# Objectives

- Study the efficiency of the widespread CAPTCHAs
- CAPTCHAs from the largest Russian mobile network operators web sites were chosen



## Reasons of choice

- Operators have enough money to hire a programmer of any qualification
- Operators need to minimize the amount of spam in order to safeguard their reputation



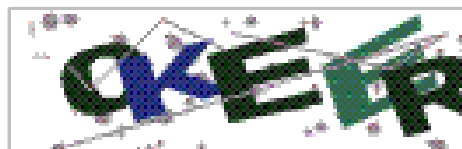
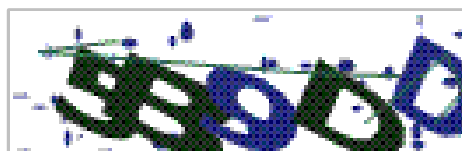
# Recognition method overview

- Preprocessing
- Segmentation
- Recognition

In the following slides details on these stages will be given.

# Preprocessing

- Clearing the noise
- Removing distortions

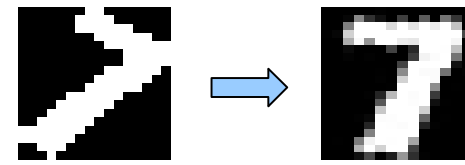
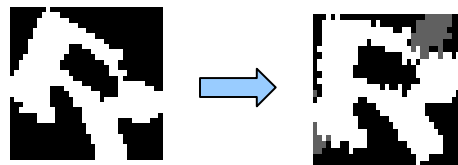
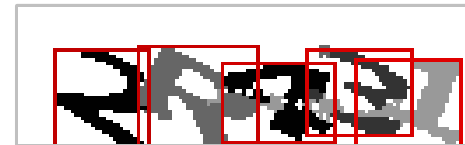


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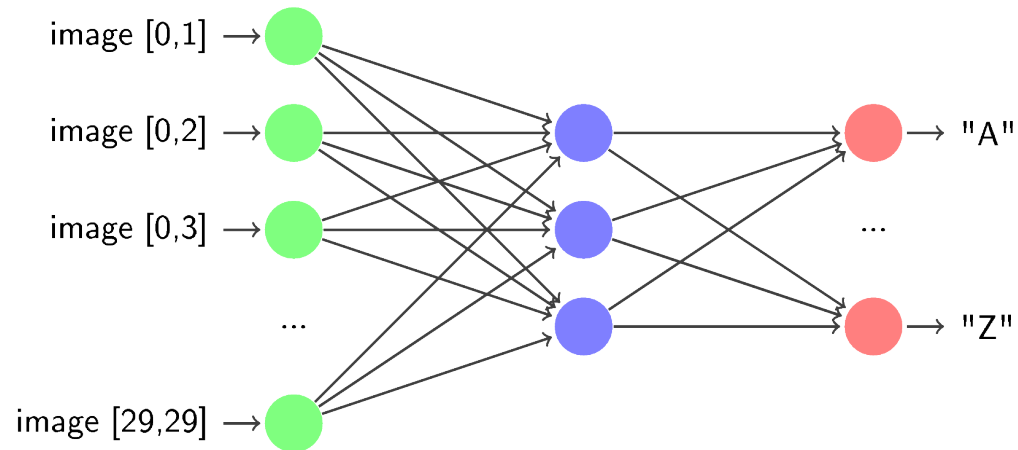
# Segmentation

- Extracting characters
- Post-processing characters



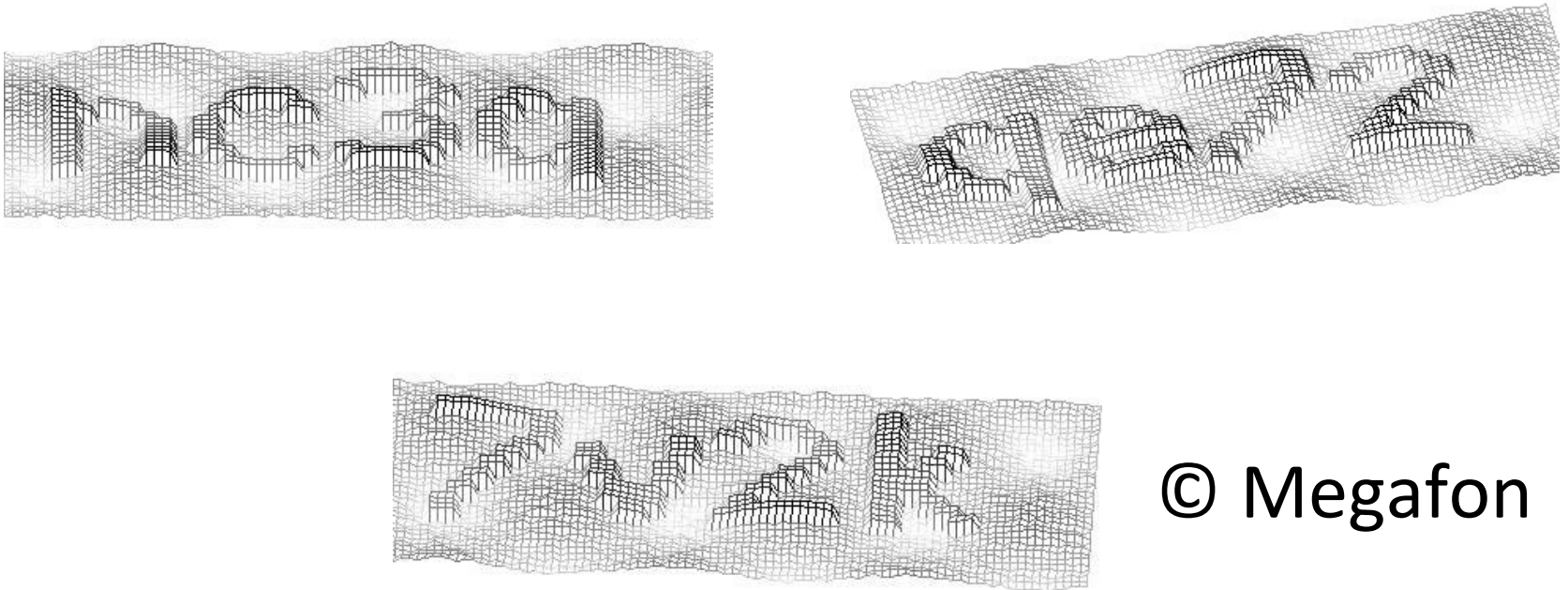
# Recognition

- Classification of characters with a pre-trained neural network



# Example

Let us consider the following type of CAPTCHA:



© Megafon

# Analyzing the problem

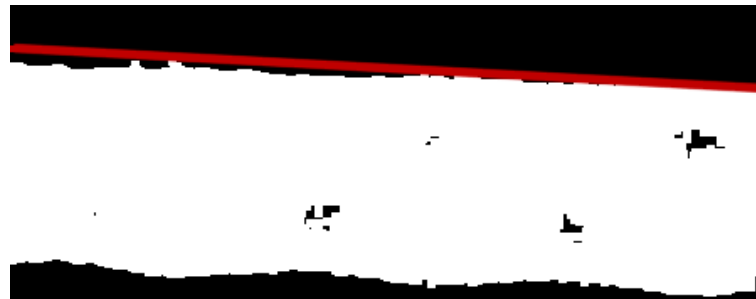
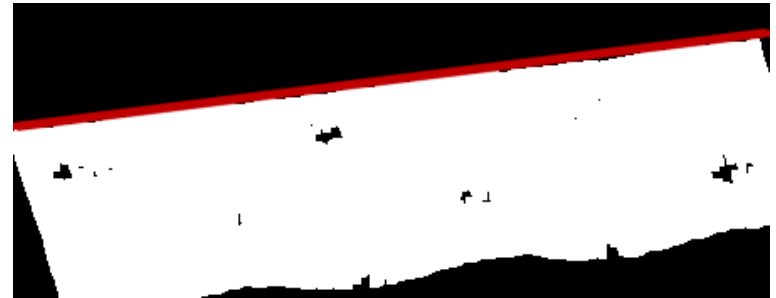
- Characters lie on a 3D wireframe
- The wireframe is rotated and moved
- The brightness is inconsistent
- Seems to be quite bad :(



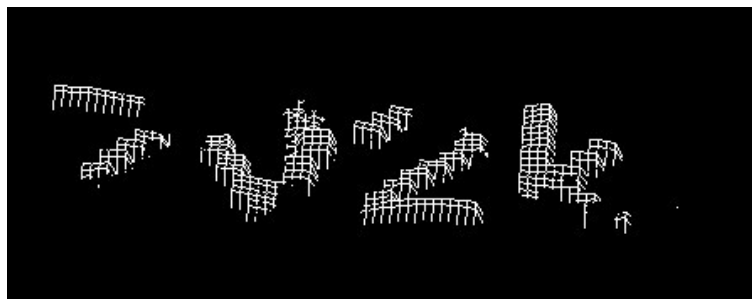
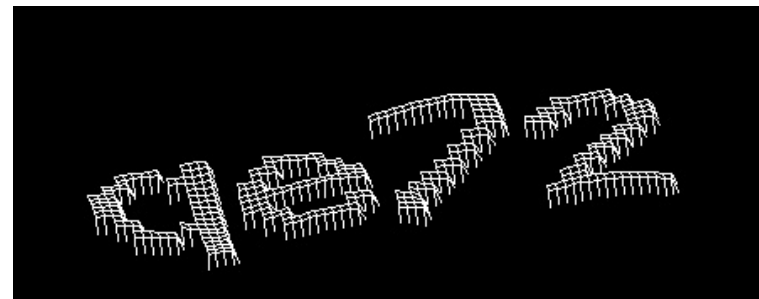
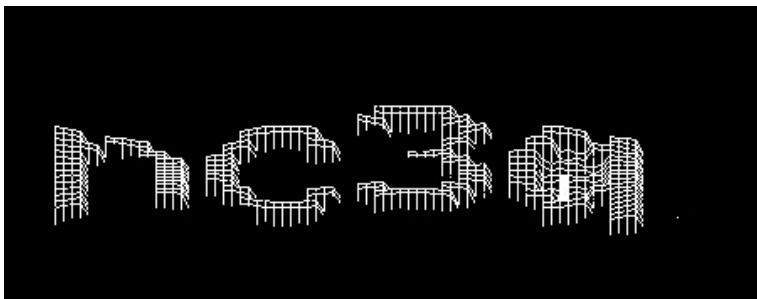
# Ideas of the solution

- Ignore the three-dimensionality and use classic methods
- The characters are generally darker than the background and can be separated by brightness
- The upper side of the wireframe is clearly seen – this can be used for the reverse rotation

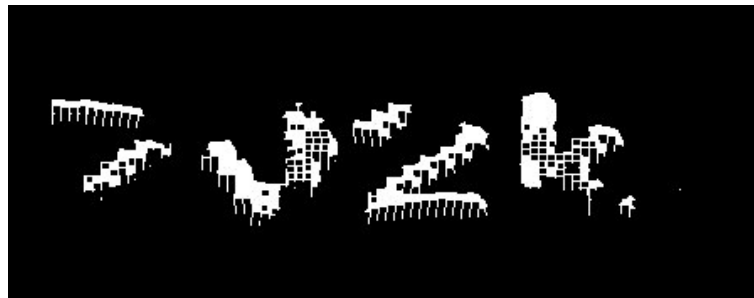
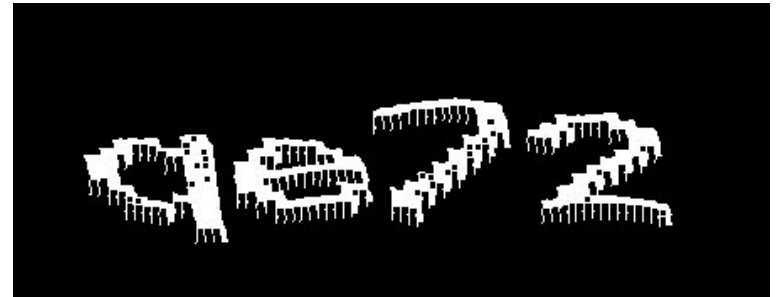
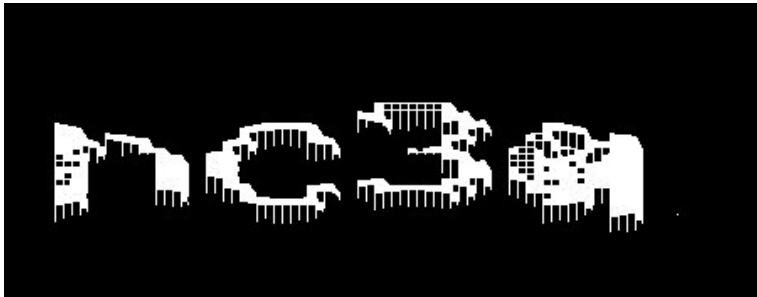
# Estimating the rotation angle



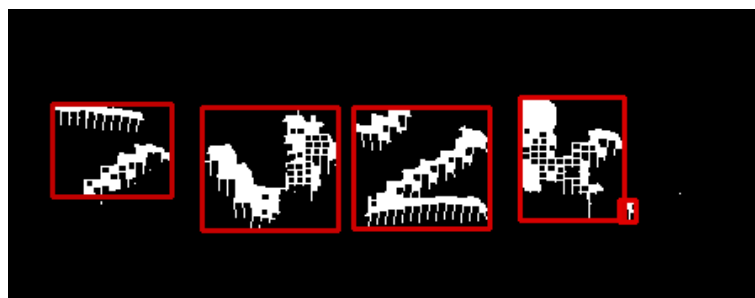
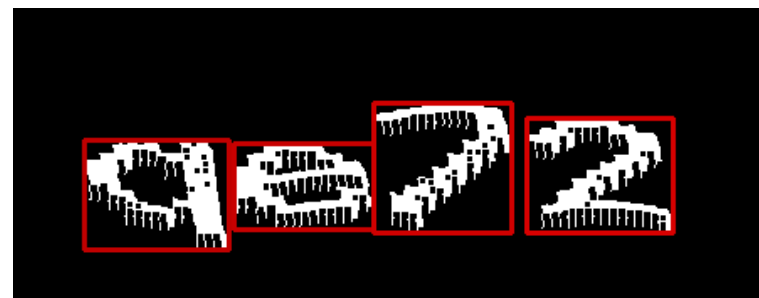
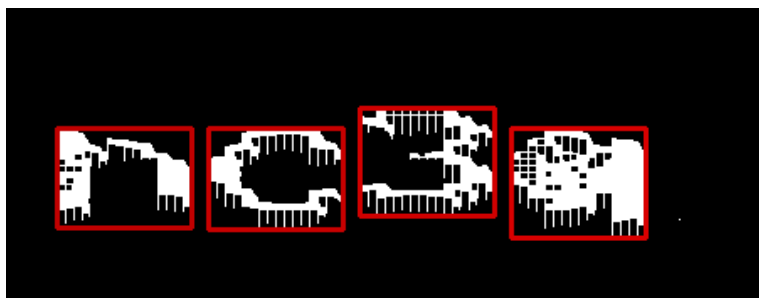
# Removing the background



# Removing tiny holes



# Segmentation



# Statistics

- Total number of images – 100
- Recognized successfully – 69
- Recognition error – 31
- Average error – 0.3 characters

# Other types of CAPTCHAs

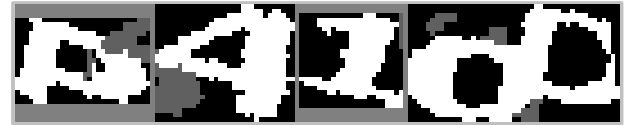
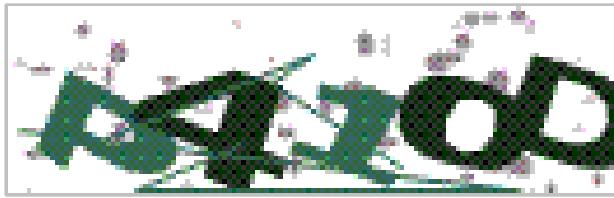
- Preprocessing varies greatly
- Segmentation is quite similar
- Almost identical recognition

Conclusion — the more transformations are applied to the original image, the more general methods can be used.

# Neural network segmentation

- In Beeline's CAPTCHA, the classic method did not show satisfactory results
- A new method which combines the segmentation and recognition was developed

# Example



© Beeline

# Conclusion

- Only preprocessing varies significantly
- All considered types of CAPTCHAs proved to be inefficient reverse Turing tests

Questions?