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06 February 2019

First and Last name: \_\_\_\_\_

ETH number: \_\_\_\_\_

Signature: \_\_\_\_\_

**General Remarks**

- At first, please check that your exam questionnaire is complete (there are 36 pages total).
- Remove all material from your desk which is not allowed by examination regulations.
- Fill in your first and last name and your ETH number and sign the exam. Place your student ID in front of you.
- You have 3 hours for the exam. There are 9 questions, where you can earn a total of 180 points. You don't have to score all points to obtain a very good grade.
- Answer each question directly on the exam sheets. There is enough space left for you to fully answer the questions. If, for whatever reason, your answer is on the back of the sheet (or on an additional piece of paper), state it clearly on the space reserved for the answer.
- Please do not use a pencil or red color pen to write your answers.
- You may provide at most one valid answer per question. Invalid solutions must be cancelled out clearly.

	Topic	Max. Points	Points Achieved	Visum
1	Filtering	21		
2	Miscellaneous	10		
3	PCA	16		
4	Optical Flow	20		
5	Fourier Transform	23		
6	Geometry and Rendering	30		
7	Transformations	30		
8	Animation	22		
9	Optimization	8		
Total		180		

Grade: .....

## Question 1: Filters and Image Features (21 pts.)

a) What is the purpose of applying a blur filter before edge or corner detection? **1 pt.**

ANSWER

*Removes noise (corresponding to very high frequencies).* **1 pt.**

ANSWER

b) Briefly explain what a separable kernel is and why it is favorable. **2 pts.**

ANSWER

*A separable kernel is  $K(m,n) = f(m)g(n)$ .  
It is computationally more efficient since we need to convolve the image with 1D kernels twice.  
For a kernel with size  $l \times l$ , the complexity in terms of the kernel size is  $O(l + l)$  rather than  $O(l^2)$ .* **1 pt.**

ANSWER

c) What could be the purpose of the kernel given below? What is the effect of convolving an image with this kernel? Explain whether it is a separable kernel or not. **4 pts.**

1	0	-1
2	0	-2
1	0	-1

ANSWER

*Sobel x-direction kernel.  
Calculates the derivative in the horizontal direction.* **1 pt.**  
*Detects vertical edges.* **1 pt.**

*We can apply separate 1D kernels in horizontal and vertical directions to achieve the same result.  
First, convolve the image by  $[1, 0, -1]$  filter in the horizontal direction. **1 pt.**  
Then, convolve the resulted image by  $[1, 2, 1]$  filter in the vertical direction. **1 pt.**  
Or the filter is separable because the kernel matrix has rank 1.*

ANSWER

d) We apply the following filtering operation on an image  $I$ :

$$I'(x, y) = \sum_{i, j \in \mathcal{N}(x, y)} f(i, j) I(x + i, y + j)$$

What is the effect if  $f$  is a Gaussian kernel?

**1 pt.**

[illegible]

*Blurring or smoothing.*

**1 pt.**

[illegible]

e) Bilateral filtering makes the following modification:

$$I'(x, y) = \frac{1}{Z} \sum_{i, j \in \mathcal{N}(x, y)} f(i, j) g(I(x, y) - I(x + i, y + j)) I(x + i, y + j)$$

where  $g : \mathbb{R} \rightarrow \mathbb{R}$  is a Gaussian on the intensity difference and  $Z$  is the normalization factor.

i) Explain what this modification implies.

**1 pt.**

[illegible]

*Edge-preserving smoothing. The assumption of slow spatial variations fails at edges, which are consequently blurred by Gaussian filter. Bilateral filtering introduces a content-based filter by considering intensity difference between the neighboring pixels.* **1 pt.**

**1 pt.**

[illegible]

ii) Write down the normalization factor  $Z$  such that the weights in the weighted combination of pixels sum to 1. **1 pt.**

**1 pt.**

[illegible]

$$Z = \sum_{i,j \in \mathcal{N}(x,y)} f(i,j)g(I(x,y) - I(x+i,y+j))$$

**1 pt.**

[illegible]

iii) Explain if bilateral filtering is linear and shift-invariant.

**2 pts.**

[illegible]

Due to the non-linear intensity difference function  $g$  and the normalization factor  $Z$ , the weights depend on the spatial location. In other words, the weights are different for each output pixel. Hence, it is not linear and not shift-invariant. **2 pts.**

**2 pts.**

[illegible]

- f) Explain how the Hough Transform detects straight lines in an image using the polar parameterization  $(\theta, \rho)$ . **3 pts.**

[illegible]

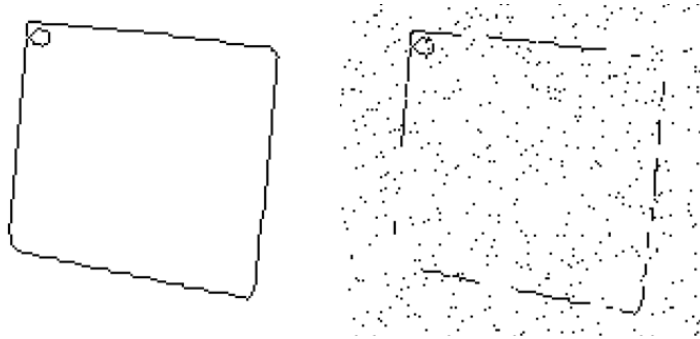
Each point  $(x_i, y_i)$  in the  $xy$ -plane gives a sinusoid in the  $\theta\rho$ -plane. **1 pt.**

Colinear points lying on the line give curves intersecting at the same point in the polar parameter plane. **1 pt.**

*Local maxima give significant lines.* **1 pt.**

[illegible]

- g) We apply Hough Transform to detect 4 straight lines on the left image below. How does degrading the original image as shown in the right image affect the result? **2 pts.**

[illegible]

*The local maximums in the polar plane do not change.* **1 pt.**

Hence, the algorithm will find the same major lines. **1 pt.**

*Note: still **2 pts.** if the student's reasoning follows high amount of noise and not being able to detect the same lines.*

[illegible]

- h) Explain if Harris corner detector is invariant to intensity shift  $I(x, y) \rightarrow I(x, y) + a$  for some constant  $a$ . **1 pt.**

[illegible]

Since only the derivatives are used, it is invariant to intensity shift. **1 pt.**

[illegible]

- i) Explain if Harris corner detector is invariant to **intensity scale**  $I(x, y) \rightarrow bI(x, y)$  for some constant  $b$ . **1 pt.**

ANSWER

*It depends on the corner response threshold. The gradient is scaled by  $b$  and hence the eigenvalues of the matrix  $M$  by  $b^2$ . Depending on the value  $b$  and the original corneriness threshold, new corner pixels may appear.* **1 pt.**

*Note: if the reasoning is provided it doesn't matter if the conclusion is invariant or not invariant.*

ANSWER

- j) When may temporal redundancy reduction for video compression be ineffective?. **2 pts.**

ANSWER

*When there are many scene changes* **1 pt.**  
*Or/and high motion* **1 pt.**

ANSWER

### Question 2: Miscellaneous (10 pts)

a) Typically, why are the colors green and blue chosen for chromakeying?

**1 pts.** ANSWER

*Because typically these systems are used to acquire humans and the color of their skin does not have large green and blue components. (Humans: 1)*

[illegible]

b) List three potential problems that may arise when doing background subtraction using a chromakeying approach with a solid colored background (e.g. blue or green screen). Justify each answer. **3 pts.** ANSWERANSWERANSWERANSWERANSWERANSWERANSWERANSWERANSWERANSWER

**3 pts.** ANSWERANSWERANSWERANSWERANSWERANSWERANSWERANSWER

People wearing cloth with the same or similar color to the background one (they cannot be recognized as foreground). Motion blur (the recorded colors result as a mixture between background and foreground). Shadows on the background or incorrect lighting of the scene (changes the color of the background leading to an increase of the color space reserved for the background). Blue or green halo around the foreground (due to aliasing artifact of the sensor or the lens). Blue or green halo around the hair of a foreground character. Transparency. (Problem: 0.5) (Explanation: 0.5) (Problem and explanation (if not obvious): 1)

[illegible]

c) Suppose you want to capture a video of an actor in front of a static background and you need to perform background subtraction to further process your footage. Assume that the camera is not moving and you cannot cover the surroundings with colored sheets. 1) What else can you do? 2) What if the filming location is inside the Zurich Hauptbahnhof where you cannot stop people from passing by the field of view of your camera? **2 pts.** ANSWER

**2 pts.** ANSWER

[illegible][illegible][illegible]

### Question 3: PCA (16 pts.)

- a) Show that the variance of the projected data on a principal component is equivalent to the corresponding eigenvalue of the original covariance matrix. **4 pts.**

[illegible]

*First part: writing the covariance matrix of the projected data.*

*Data mean  $\bar{x}$*

*Data covariance*  $\Sigma = \frac{1}{N-1} \sum_{n=1}^N (x^{(n)} - \bar{x})(x^{(n)} - \bar{x})^T$ , where  $x^{(n)}$  is a data sample.

$u_1$  defines the direction of the first principal component where  $u_1^T u_1 = 1$

Projected sample  $x_p^{(n)} = u_1^T x^{(n)}$

Mean of the projected data  $\bar{x}_p$

### Variance of the projected data

$$\Sigma_p = \frac{1}{N-1} \sum_{n=1}^N \{(x_p^{(n)} - \bar{x}_p)\}^2$$

$$\Sigma_p = \frac{1}{N-1} \sum_{n=1}^N u_1^T (x^{(n)} - \bar{x})(x^{(n)} - \bar{x})^T u_1$$

$$\Sigma_p = u_1^T \Sigma u_1$$

*Note: Still 2 pts. if the answer is only  $\Sigma_p = u_1^T \Sigma u_1$  or  $\Sigma_p = U^T \Sigma U$  where  $U$  is the transformation matrix.*

Second part: showing the equivalence by using either eigenmatrix properties or Lagrangian formulation **2 pts.**

$\Sigma u_1 = u_1 \lambda_1$  where  $u_1$  and  $\lambda_1$  are the eigenvector and eigenvalue of  $\Sigma$ .

$$\Sigma_p = u_1^T \Sigma u_1$$

$$\Sigma_p = u_1^T u_1 \lambda_1$$

$$\Sigma_p = \lambda_1$$

Alternatively, writing the Lagrangian of the problem:  $\max_{u_1} u_1^T \Sigma u_1$  subject to  $u_1^T u_1 = 1$  is a valid answer.

[illegible]



- b) We apply PCA on two datasets  $A, B \in \mathbb{R}^2$ . The normalized eigenvalues of dataset A and B are  $[0.95, 0.05]$  and  $[0.55, 0.45]$ , respectively. Illustrate the distribution of dataset A and B in separate plots and explain the relationship between the shape of dataset distribution and eigenvalues. **4 pts.**

[illegible]

*Dataset A has an ellipsoid shape where the majority of samples are dispersed in a particular direction*

*Dataset B has a round/spherical shape* **1 pt.**

PCA looks for the directions that the variance along this axis is maximized. The eigenvalues corresponding to this eigenvectors explain the amount of total variation. Hence, we observe larger variance/dispersion on the axis of larger eigenvalue **2 pts.**

[illegible]

- c) The PCA algorithm is trained on dataset F of faces and is then used to compress photographs from the dataset C of chairs. Why does not it work well despite "packing the most energy in its first J coefficients"? **2 pts.**

[illegible]

PCA is a data-dependent compression **1 pt.**

*"Packing the most energy" is in context of faces dataset  $F$ . The transformation matrix, i.e., truncated eigenmatrix of  $F$  captures features relevant to faces, not chairs* **1 pt.**

[illegible]

- d) You are given a dataset of 1000 images with size  $50 \times 50$ . Calculate the maximum number of principal components that allows us to save the dataset by using at least 1052500 unit space. You should consider all information required to decompress an image or compress a new one. We assume that each array entry or number costs 1 unit space. For example, the original dataset takes  $1000 \times 50 \times 50 = 2500000$  unit space. **6 pts.**

[illegible]

$K$  is the dimensionality of compressed images.

*Compression and decompression formula:*

$$\begin{aligned} I_K &= (I - \bar{I})\Phi \\ \hat{I} &= I_K\Phi^T + \bar{I} \end{aligned}$$

where  $I, \bar{I} \in \mathbb{R}^{2500}$  are the vectorized image and the mean image.  $I_K \in \mathbb{R}^K$  is the  $K$ -dimensional compressed image.  $\Phi \in \mathbb{R}^{2500 \times K}$  is the truncated eigenmatrix of covariance.

*We need to store*

1. *dataset mean*  $2500 = 50 \times 50$  **2 pt.**
2. *truncated eigenmatrix*  $2500 \times K$  **2 pt.**
3. *compressed images*  $1000 \times K$  **2 pt.**

$$\begin{aligned} 2500 + 2500 \times K + 1000 \times K &\leq 1052500 \\ K &\leq 300 \end{aligned}$$

**-1 pt.** if the left side of the inequality is wrong.

[illegible]

### Question 4: Optical flow (20 pts.)

- a) The Lucas-Kanade algorithm can be used to estimate the optical flow of an image sequence. Nevertheless, it doesn't work for all cases. State the 3 assumptions that have to be fulfilled so that this method works **3 pts.** ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER

1. *Brightness constancy: the intensity of the objects in the scene do not change in time*
2. *Small motion: objects move very slowly from frame to frame, which means that corresponding points from 2 consecutive images are not far apart*
3. *Spatial coherence: all points in a neighborhood have the same motion*

- b) The Lucas-Kanade method provides the following formulation

$$\begin{pmatrix} \sum I_x^2 & \sum I_x I_y \\ \sum I_x I_y & \sum I_y^2 \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} = - \begin{pmatrix} \sum I_x I_t \\ \sum I_y I_t \end{pmatrix}$$

Where  $(u, v)$  is the displacement of a pixel within an image patch.

Mathematically derive the above result using the 3 basic assumptions of this method. Indicate these 3 assumptions during derivation and also point out what is the aperture problem. **8 pts.**

First use assumption 1 to conclude  $I(x, y, t) = I(x + u, y + v, t + 1)$  (2 pt)  
 Then perform a Taylor expansion about  $x, y$ , using the second assumption. This gives:  
 $I(x + u, y + v, t + 1) = I(x, y, t) + I_x u + I_y v + I_t$  (2 pt)  
 Since  $I(x, y, t) = I(x + u, y + v, t + 1)$ , then  $I_x u + I_y v = -I_t$  This leads to a single equation  
 with two unknowns, which is the aperture problem (1 pt). To solve it, we use the third  
 assumption to consider all points within a patch  $\omega$  to have a similar motion: (2 pts)



Image pyramids in a coarse-to-fine approach(1 pt)  
Explain how the approach works (2 pts)

d) i) Explain the general idea of how optical flow information can be used for video compression. **1 pts.**

*Encode the motion of the pixels instead of image frames.*

ii) What image artifacts can appear in video compression if you compute frame  $n + 1$  by applying forward propagation to frame  $n$  using the optical flow computed from image  $n$  to  $n + 1$ . **2 pts.** ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER

*OF of 2 different pixels can point to the same coordinates, holes appear in the video*

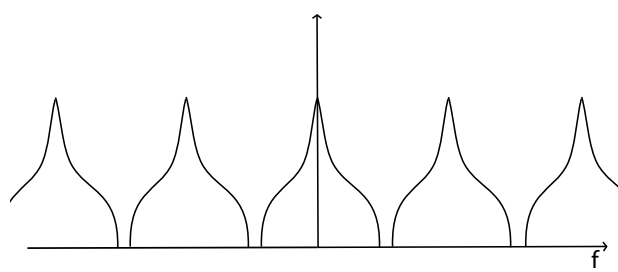
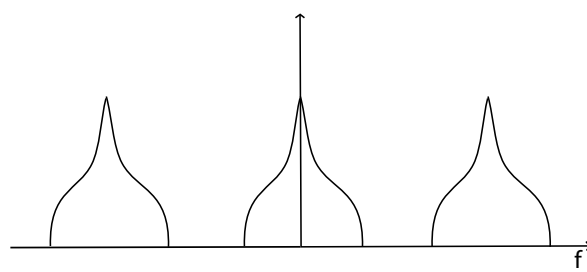
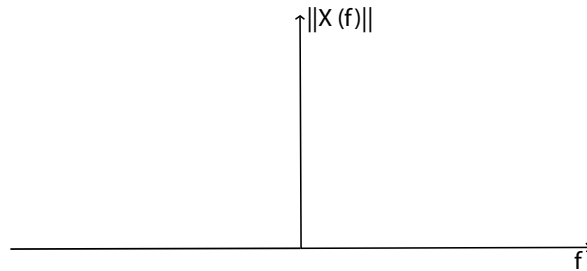
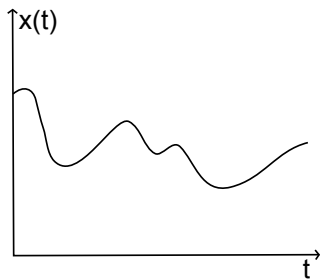
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### Question 5: Fourier transform (23 pts.)

a) Briefly state the convolution theorem in one or two sentences.

**5 pts.**

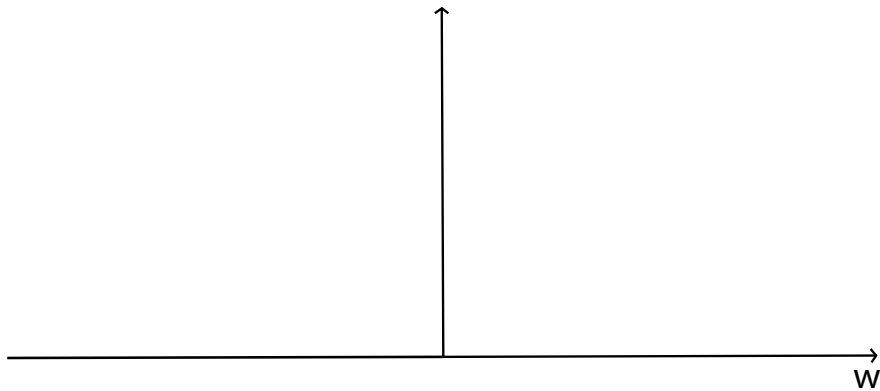
b) Consider the continuous-time signal and the Fourier transforms of two signals that were sampled from the original below. Roughly fill in the missing diagrams. **9 pts.**



- c) Assume that the signal in the previous part was sampled at a frequency that is below the Nyquist rate. Roughly draw the sampled signal and its Fourier transform below. **5 pts.**



- d) Briefly explain what the `fftshift` function we use to visualize the Fourier domain representations in MATLAB does using the 1-D diagram for the frequency domain below. **4 pts.**









where  $\mathbf{n}_a$ ,  $\mathbf{n}_b$  and  $\mathbf{n}_c$  are the normals at points a, b and c.

Compute the scalars  $\alpha$ ,  $\beta$  and  $\gamma$ .

Use the notation  $A_{uvw}$  for the area of a triangle formed by points  $u$ ,  $v$  and  $w$ . **3 pts.**

[illegible]

$$- \alpha = A_{xbc}/A_{abc} - \beta = A_{xac}/A_{abc} - \gamma = A_{xab}/A_{abc}$$

[illegible]

iii) Name one method to avoid aliasing using textures.

**2 pts.**

ANSWER      ANSWER

*undersampling texture. or Mipmapping.*

[illegible]

iv) In the graphics pipeline, what is the depth-buffer used for?

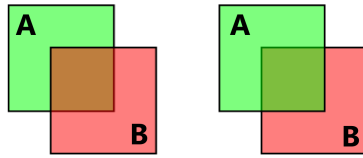
**2 pts.**

ANSWER ANSWER

*to test for occlusion*

[illegible]

- v) In the figure below, two transparent rectangles with transparencies  $\alpha_A$  and  $\alpha_B$  are overlaid. Below is the *incomplete* equation to compute the composite color  $C$ . Complete the equation, such that it computes the composite color of the **right** image below. **3 pts.**



$$C = \alpha_B B + \alpha_A A$$

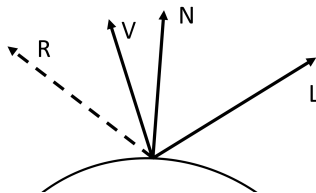
[illegible]

$$C = (1 - \alpha_A)\alpha_B B + \alpha_A A$$

[illegible]

- vi) In many reflection models, e.g. the Phong reflection model, the reflection vector  $\mathbf{R}$  is used, which is the direction that a perfectly reflected ray of light would take from a point on the surface.

Given the normal vector  $\mathbf{N}$ , the vector pointing towards the light source  $\mathbf{L}$  and the vector pointing towards the camera  $\mathbf{V}$ , derive the formula to compute the reflection vector  $\mathbf{R}$ . **4 pts.** ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER ANSWER



$$R = (L \cdot N)N + (L \cdot N)N - L = 2(L \cdot N)N - L$$

[illegible]

- vii) The Phong reflection model describes the way surface reflects light by the combination of three reflection terms. What are these three terms called? And which of these terms uses the reflection vector  $\mathbf{R}$ ? **3 pts.** ANSWER ANSWER ANSWER ANSWER ANSWER

*ambient, diffusive and specular. reflection vector used onl in specular term*

ANSWER ANSWER

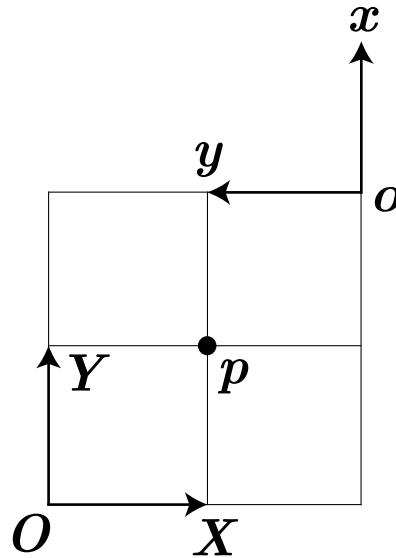
### Question 7: Coordinate Systems and Transformations (30 pts.)

## a) Coordinate System Basics

This is a 2D problem in homogeneous coordinates.

The 2D Cartesian point  $\begin{pmatrix} x \\ y \end{pmatrix}$  is written as a homogeneous point  $\begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$ .

The 2D Cartesian vector  $\begin{pmatrix} x \\ y \end{pmatrix}$  is written as a homogeneous vector  $\begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$ .



The world coordinates are defined by axes  $\mathbf{X} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ ,  $\mathbf{Y} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ , and origin  $\mathbf{O} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ .

The object coordinates are defined by axes  $x, y$  and origin  $o$ .

The grid cells are unit length.

- i) Write  $p_{\text{world}}$  ( $p$  in world coordinates) as a homogeneous point.

**1 pt.**

[illegible]

$$\mathbf{p}_{world} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

2

[illegible]





viii) (Challenge) Say  $\mathbf{F} = \mathbf{RT}$ , where  $\mathbf{R}$  is a rotation matrix, and  $\mathbf{T}$  is a translation matrix. Write down valid choices for  $\mathbf{R}$  and  $\mathbf{T}$  (using numbers). **4 pts.**  
HINT:  $\mathbf{F}$  maps  $\mathbf{X} \rightarrow \mathbf{x}$ ,  $\mathbf{Y} \rightarrow \mathbf{y}$ , and  $\mathbf{O} \rightarrow \mathbf{o}$ .

[illegible]

*This question asks for  $RT$ , i.e. a translation followed by a rotation.*

Look at the picture. Note that we can map the world coordinate system to the object coordinate system by translating  $(2, -2)$  and then rotating (counter-clockwise about the origin)  $90^\circ$ .

$$\Rightarrow \mathbf{R} = \begin{pmatrix} \cos(90^\circ) & -\sin(90^\circ) & 0 \\ \sin(90^\circ) & \cos(90^\circ) & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \mathbf{T} = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{pmatrix}$$

*Alternate Solution: Write down the general forms of  $\mathbf{R}$  and  $\mathbf{T}$ .*

$$\mathbf{R} = \begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \mathbf{T} = \begin{pmatrix} 1 & 0 & u \\ 0 & 1 & v \\ 0 & 0 & 1 \end{pmatrix}$$

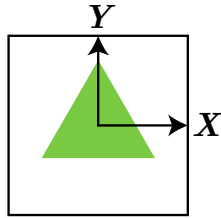
$$\mathbf{RT} = \begin{pmatrix} \cos \theta & -\sin \theta & u \cos \theta - v \sin \theta \\ \sin \theta & \cos \theta & u \sin \theta + v \cos \theta \\ 0 & 0 & 1 \end{pmatrix} \text{ require equal to } \begin{pmatrix} 0 & -1 & 2 \\ 1 & 0 & 2 \\ 0 & 0 & 1 \end{pmatrix} = \mathbf{F}$$

Choose  $\theta = 90^\circ$  so the top-left  $2 \times 2$  submatrix matches. Substituting into the matrix equation requires  $-v = 2$  and  $u = 2$ , and so the choice  $(u, v)^T = (2, -2)^T$  satisfies the entire matrix equation.

[illegible]



b) Transformation Families



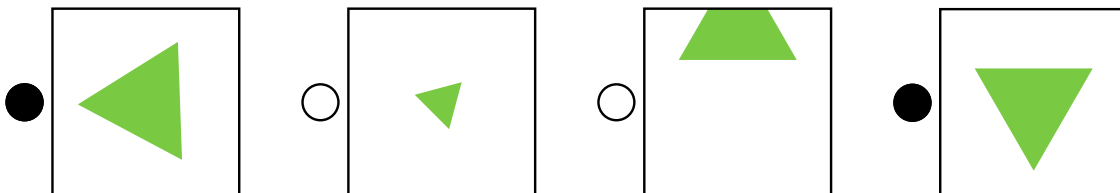
This is the original triangle. The world origin is at the center of the triangle, and the world axes are as drawn as unit vectors. The box around the triangle is **your screen**.

For each question in this section I will give you a parameterized transformation  $F_\theta$  and four pictures of triangles **clipped to the screen**. Consider each of the four pictures individually. If a picture is the result of applying  $F_\theta$  to the original triangle **for some**  $\theta$  then **fill in the circle next to it**.

NOTE: To be clear, if the vertices of the original triangle are  $\{p_1, p_2, p_3\}$ , then choose a new triangle if and only if its vertices are  $\{F_\theta p_1, F_\theta p_2, F_\theta p_3\}$  for some  $\theta$ .

NOTE: I'm not trying to trick you with the pictures. If a picture looks like the original triangle moved to the right, then it is the original triangle moved to the right.

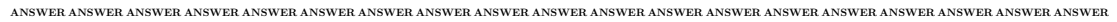
**EXAMPLE:**  $F_\theta = \begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$



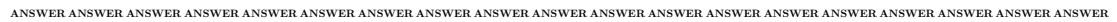
Explanation: This is a rotation by  $\theta$ . I fill in the circles next to the two pictures that could be made by rotating the original triangle about the origin by some angle  $\theta$ .

The other two pictures cannot possibly be made by  $F_\theta$  so I leave their circles blank.

**4 pts.**

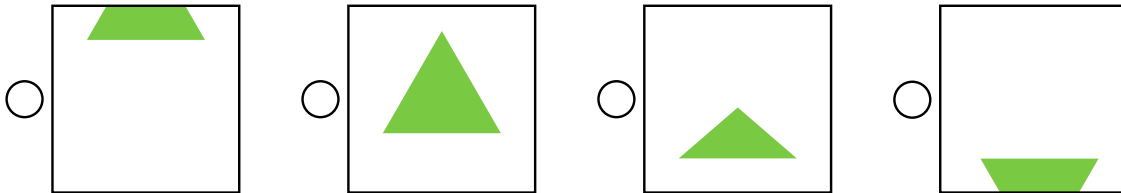
[illegible]

**4 pts.**

[illegible]

iii) (Challenge)  $\mathbf{F}_\theta = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \theta & \theta \\ 0 & 0 & 1 \end{pmatrix}$

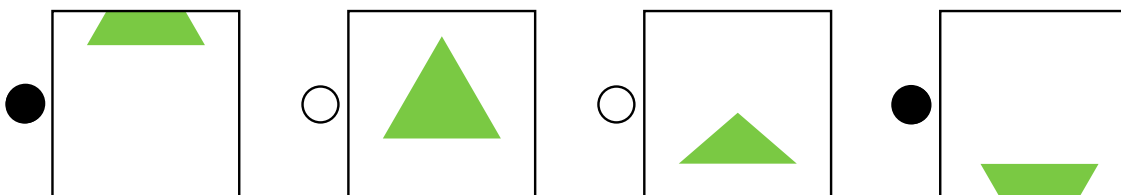
**4 pts.**

[illegible]

(Explanation Not Required) –

$$\mathbf{F}_{\theta}\mathbf{p} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \theta & \theta \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} x \\ \theta y + \theta \\ 1 \end{pmatrix}$$

*This is a scaling by  $\theta$  in  $\mathbf{y}$  followed by a translation by  $\theta$  in  $\mathbf{y}$ .*

[illegible]

### c) Random Questions

i) Do rotations commute in 2D?

**1 pt.**

[illegible]

Yes.

[illegible]

ii) Do translations commute in 7D?

**1 pt.**

[illegible]

Yes.

[illegible]

iii)

$$\mathbf{R} = \begin{pmatrix} \cos 1^\circ & -\sin 1^\circ & 0 \\ \sin 1^\circ & \cos 1^\circ & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\prod_{i=1}^N R = \underbrace{R * R * \cdots * R}_{N \text{ copies}} = I$$

NOTE:  $\mathbf{I}$  is the identity matrix.

What is a possible value for  $N$ , where  $N > 0$ ?

**1 pt.**

[illegible]

360. (Explanation Not Required) – This is 360 consecutive rotations of  $1^\circ$  each, i.e. a  $360^\circ$  rotation, i.e. the identity.

[illegible]

### Question 8: Animation (22 pts.)

### a) Keyframing

- i) A widely used technique to create animations is called keyframing. Describe the idea behind this technique (maximum two sentences). **1 pt.**

[illegible]

*The idea of keyframing is to specify scene properties only at specific point in time (0.5 pt) and fill in the rest via computation (0.5 pt).*

[illegible]

- ii) A property that can be keyframed is for example the position of an object. Name two other properties of a scene that can be keyframed. **1 pt.**

[illegible]

- *Orientation*
- *Size*
- *Color*
- *Light Intensity*
- ...

[illegible]

### b) Interpolation

- i) Splines are a great tool to interpolate data. The idea is to create a piece-wise polynomial function to fit data at specific locations. The most commonly used polynomial is of 3rd degree. Explain why this is preferred to linear polynomials and to higher degree polynomials. **2 pts.**

**5 pts.**



$$\mathbf{p}(t) = (2(\frac{1}{2})^3 - 3(\frac{1}{2})^2 + 1)[\mathbf{1.0}, \mathbf{0.0}]^T + \quad (2)$$

$$((\frac{1}{2})^3 - 2(\frac{1}{2})^2 + (\frac{1}{2}))[\mathbf{0.0}, \mathbf{1.0}]^T + \quad (3)$$

$$(-2(\frac{1}{2})^3 + 3(\frac{1}{2})^2)[\mathbf{2.0}, \mathbf{1.0}]^T + \quad (4)$$

$$((\frac{1}{2})^3 - (\frac{1}{2})^2)[\mathbf{0.0}, \mathbf{1.0}]^T \quad (5)$$

$$= [1.5, 0.5]^T (4pts) \quad (6)$$

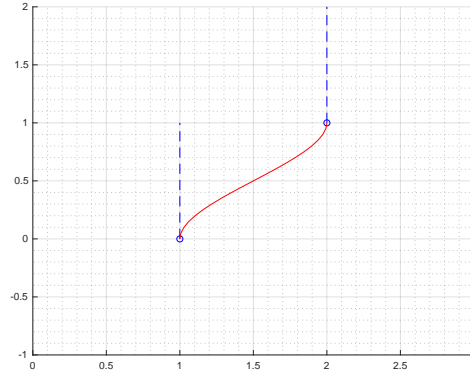


Figure 2: Data points, their tangents, and the curve (1pt).

ANSWER ANSWER

- iii) While creating an animation for a 1D parameter, you decide to use a quadratic polynomial for interpolation. You recall that a parabola has the form:

$$y = at^2 + bt + c. \quad (7)$$

You are also aware that to fully define a parabola you need three points. Therefore, you decide to use three control points:  $y_0$ ,  $y_{0.5}$ , and  $y_1$  which are defined at  $t = 0$ ,

$t = 0.5$ , and  $t = 1$ , respectively. After solving a linear system of equations, you figure out that the values of  $a$ ,  $b$ , and  $c$ :

$$a = 2y_0 - 4y_{0.5} + 2y_1 \quad (8)$$

$$b = -3y_0 + 4y_{0.5} - y_1 \quad (9)$$

$$c = y_0. \quad (10)$$

First, write down the three equations that helped you find  $a$ ,  $b$ , and  $c$ . Then, bring your quadratic spline in the following form (i.e. replace the dots between parenthesis):

$$y(t) = (\dots)y_0 + (\dots)y_{0.5} + (\dots)y_1 \quad (11)$$

**6 pts.**

[illegible]

$$y_0 = a(0)^2 + b(0) + c \quad (12)$$

$$y_{0.5} = a(0.5)^2 + b(0.5) + c \quad (13)$$

$$y_1 = a(1)^2 + b(1) + c \quad (14)$$

$$y(t) = (2t^2 - 3t + 1)y_0 + (-4t^2 + 4t)y_{0.5} + (2t^2 - t)y_1 \quad (15)$$

[illegible]

### c) Numerical Integration



- i) In order to advance a physics based animation, we need to perform numerical integration. We have seen: Forward Euler, Backward Euler, and Symplectic Euler. Write down one pro and one con for each of these methods. **3 pts.**

[illegible]

- *Forward Euler: + Easy implementation - not very stable.*
- *Backward Euler: + unconditionally stable - numerical damping, slow iterations.*
- *Symplectic Euler: + Energy is conserved - Only for second order ODEs.*

[illegible]

- ii) Figure 3 shows the image that is used as texture for a cart's wheel (note: the image is completely symmetric). The wheel is represented by a 3D (rigid) disc and its flat surface is textured with figure 3. The disc center as well as its boundaries are aligned with the image. The z-axis is perpendicular to the disc flat surface. In red you can see some measurements: the radius of the wheel is  $1m$  and the angle between the spokes of the wheel is  $45^\circ$ .

The current disc's linear velocity is  $v = [0, 0, 0]^T$ , whereas its angular velocity is  $\omega = [0, 0, \frac{5}{2}\pi]^T$ . By using the Forward Euler integration scheme, what is the smallest (positive) time step  $\Delta t$  for which the wheel would appear to be still, even if you are evolving the simulation? **4 pts.**

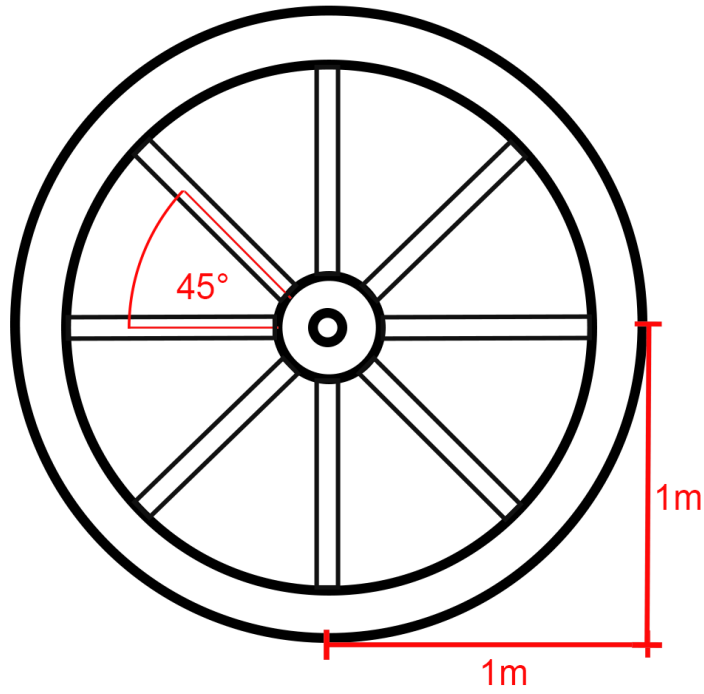


Figure 3: Orthographic view of the wheel. In black: wheel image. In red: measurements (not part of the actual image).

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$$45 \text{ deg} = \Delta angle \quad (16)$$

$$\frac{1}{8} 2\pi = \Delta t \frac{5}{2} \pi \quad (17)$$

$$\Delta t = \frac{1}{10} \quad (18)$$

ANSWER ANSWER

## Question 9: Optimization (8 pts.)

### a) Local Minima

- i) We want to perform an unconstrained continuous optimization on a function  $f(\mathbf{x}) : \mathbb{R}^n \rightarrow \mathbb{R}$ . To achieve such a task, we are interested in finding a local minima. List the two optimality conditions which make  $\mathbf{x}^*$  the solution to local minima. **2 pts.**

. ANSWER

*The gradient of  $f(\mathbf{x})$  with respect to  $\mathbf{x}$  must be zero (1pt) and its Hessian must be positive definite (1pt).*

ANSWER ANSWER

- ii) We are now interested in maximizing  $f(\mathbf{x})$  by minimizing an auxiliary function  $g(\mathbf{x})$ . Write a function  $g(\mathbf{x})$ , for which the following equivalence holds:

$$\max_{\mathbf{x}} f(\mathbf{x}) = \min_{\mathbf{x}} g(\mathbf{x}) \quad (19)$$

**1 pt.**

. ANSWER

$$g(\mathbf{x}) = -f(\mathbf{x})$$

ANSWER ANSWER

### b) Gradient Descent

- i) We want to perform the first two steps of the *Gradient Descent* algorithm to minimize the function  $f(\mathbf{x}) = f(x, y) = \frac{1}{3}x^3 + \frac{1}{2}y^2$ . First, write down the general update rule to compute the approximate solution  $\mathbf{x}_{k+1}$  given the current approximation  $\mathbf{x}_k$ . Then, starting at  $\mathbf{x}_0 = [1.0, 1.0]^T$ , perform the first two steps of Gradient Descent to find out the value of  $\mathbf{x}_2$  (for the "step control" variable, use  $\tau = \frac{1}{2}$ ). **5 pts.**

$$\mathbf{x}_1 = [1, 1]^T - \frac{1}{2}[1, 1]^T = [\frac{1}{2}, \frac{1}{2}]^T \quad (20)$$

$$\mathbf{x}_2 = [\frac{1}{2}, \frac{1}{2}]^T - \frac{1}{2}[\frac{1}{4}, \frac{1}{2}]^T = [\frac{3}{8}, \frac{1}{4}]^T \quad (21)$$

(4pts).