



Black hole orbiter (Orb)

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Any of the concentric spheres in old astronomy surrounding the earth and carrying the celestial bodies in their revolutions

WHY?

- Dostali sme to ako zadanie
 - Kvalitných simulátorov
- GC je žalosťne málo
- Je to výzva!





Introduction

Aneb jak tím A* k Orbu došiel

Tím



Team



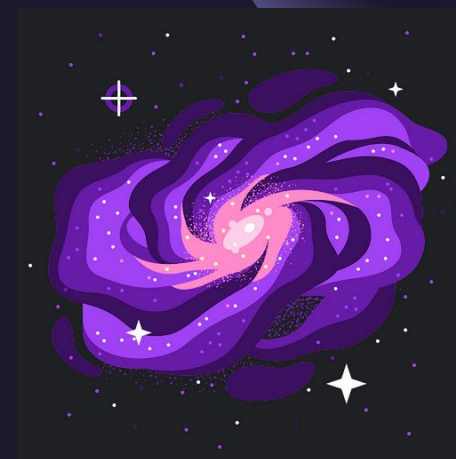
**Dominik
Pavlov**



**Samuel
Amrich**



**Filip
Stempel**



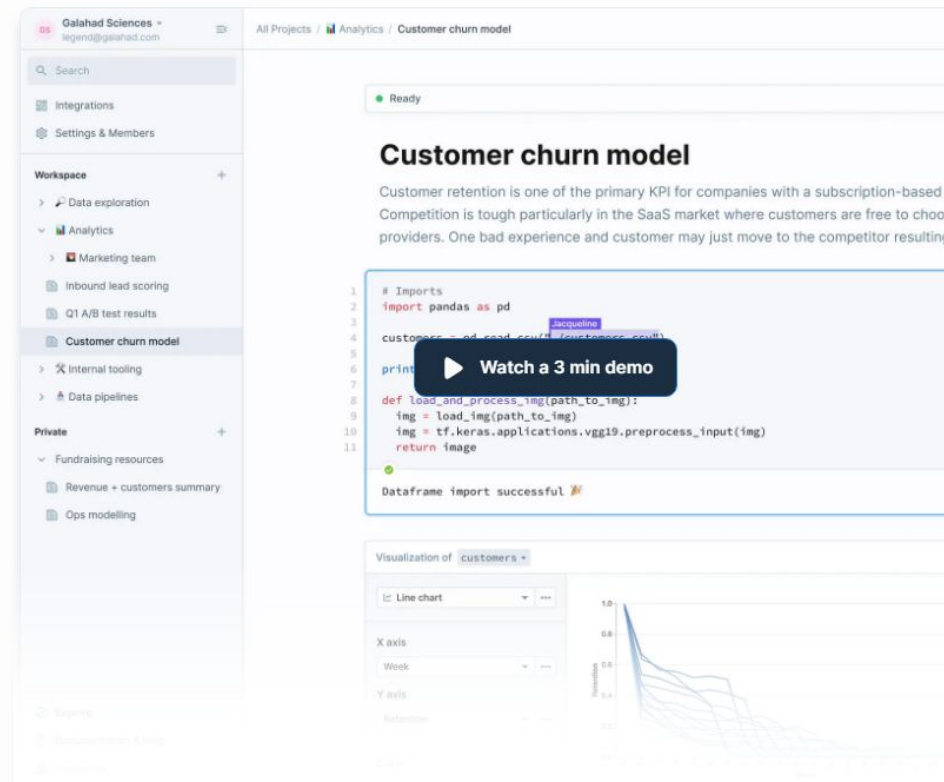
**Lukáš
Randuška**

Prostredie

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Where data teams do their best work

Deepnote is a new kind of data notebook that's built for collaboration — Jupyter compatible, works magically in the cloud, and sharing is as easy as sending a link.

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The screenshot displays the Deepnote web application interface. On the left is a sidebar with a search bar and a list of workspace items: 'Data exploration', 'Analytics' (expanded), 'Marketing team', 'Inbound lead scoring', 'Q1 A/B test results', 'Customer churn model' (selected), 'Internal tooling', 'Data pipelines', 'Private', 'Fundraising resources', 'Revenue + customers summary', and 'Ops modelling'. The main area is titled 'All Projects / Analytics / Customer churn model'. It shows a 'Ready' status and a 'Customer churn model' section with a description: 'Customer retention is one of the primary KPI for companies with a subscription-based business. Competition is tough particularly in the SaaS market where customers are free to choose providers. One bad experience and customer may just move to the competitor resulting in...'. Below this is a code editor with Python code for importing pandas and loading data. A 'Watch a 3 min demo' button is overlaid on the code. At the bottom, there's a 'Visualization of Customers' section with a line chart showing a decreasing trend over time.

```
1 # Imports
2 import pandas as pd
3
4 customers = pd.read_csv('data/customers.csv')
5
6 print(customers)
7
8 def load_and_process_img(path_to_img):
9     img = load_img(path_to_img)
10    img = tf.keras.applications.vgg19.preprocess_input(img)
11    return image
```

Dataframe import successful 🎉

Visualization of Customers

Line chart

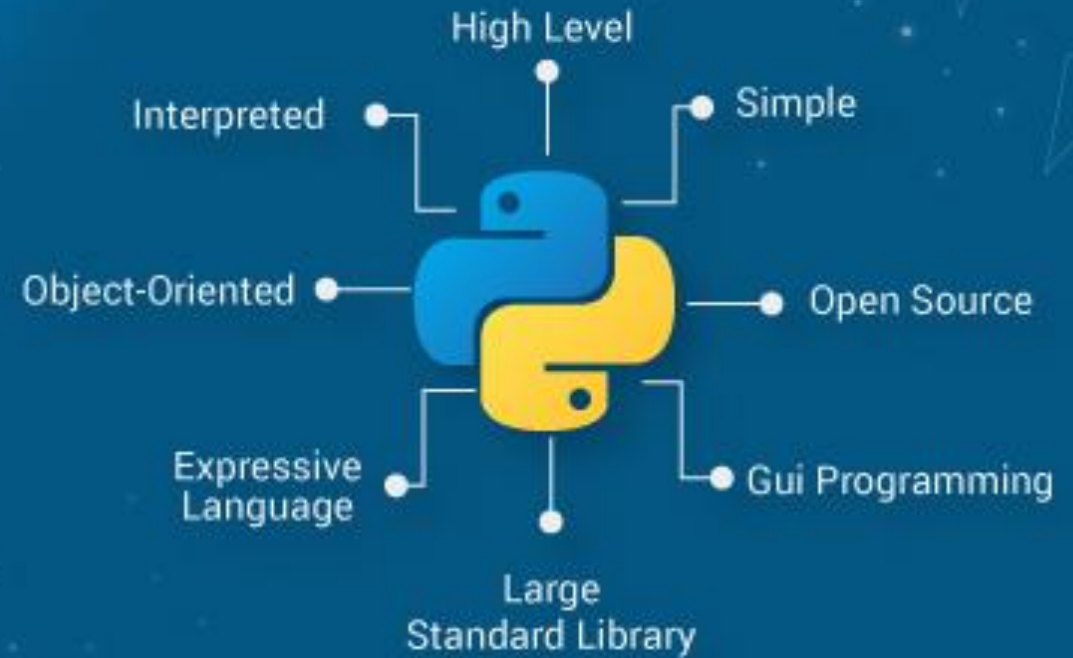
X axis: Week

Y axis: Retention

Python

Features of Python

edureka!



Nie je to C++

Problém



Globular cluster (GC)



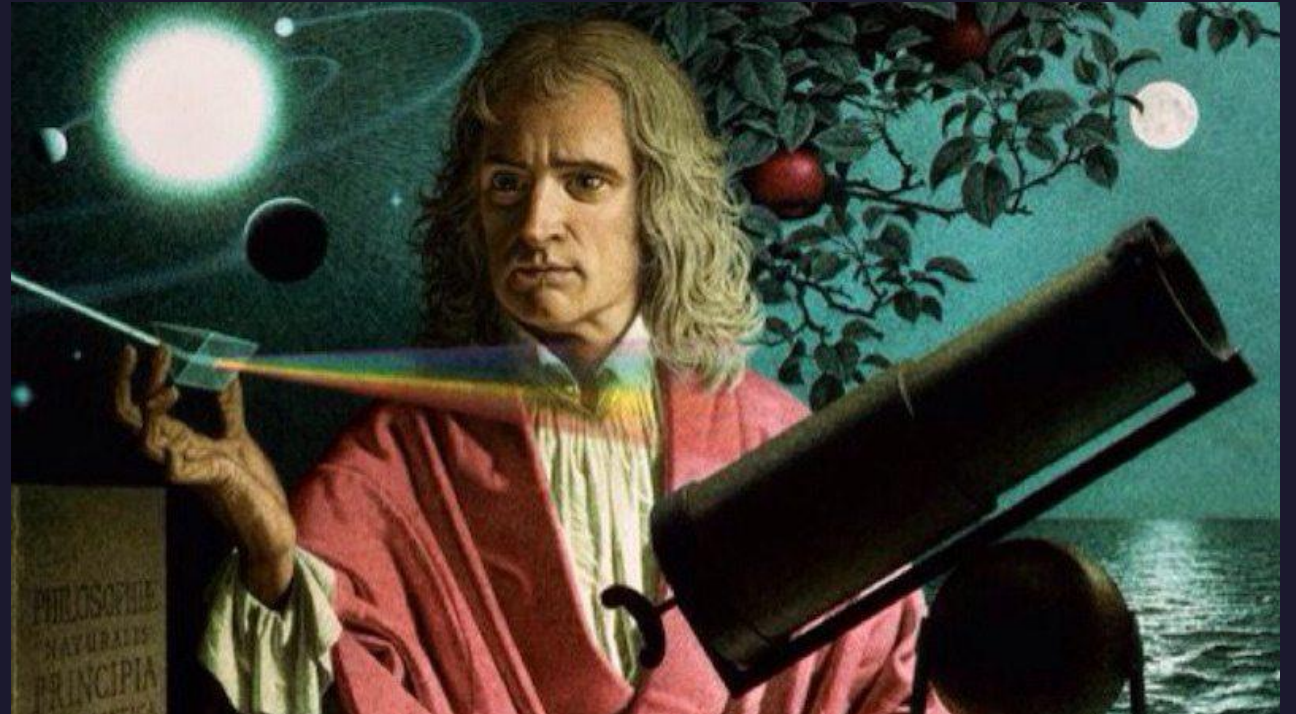
Newtonovská gravitácia

$$\vec{F} = -G \frac{M_1 \cdot M_2}{r^2} \vec{e}_r$$

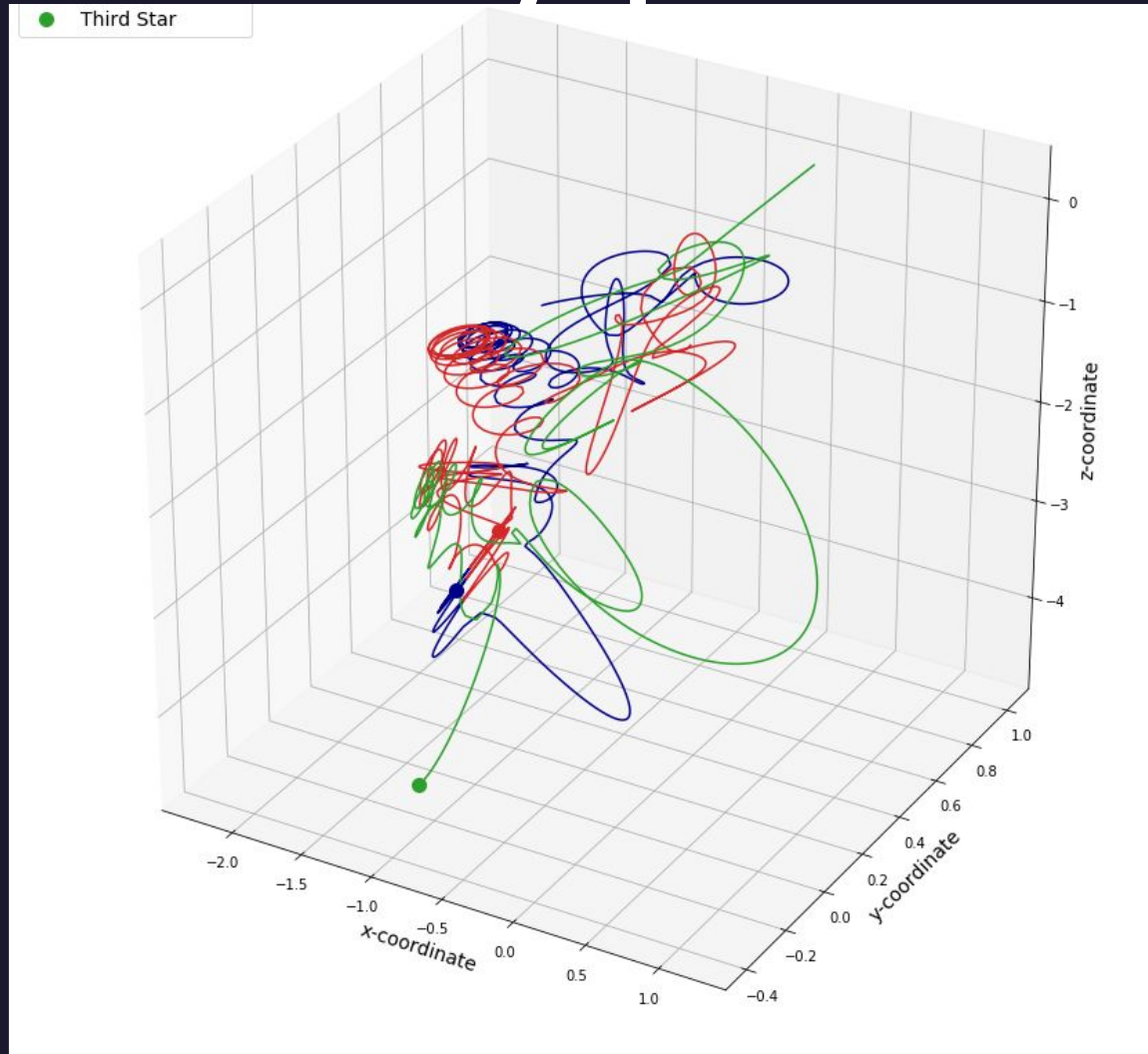
Okamžitý prenos informácie

Univerzálna sila

Jediná uvažovaná sila



N-body problem



Analyticky neriešiteľné

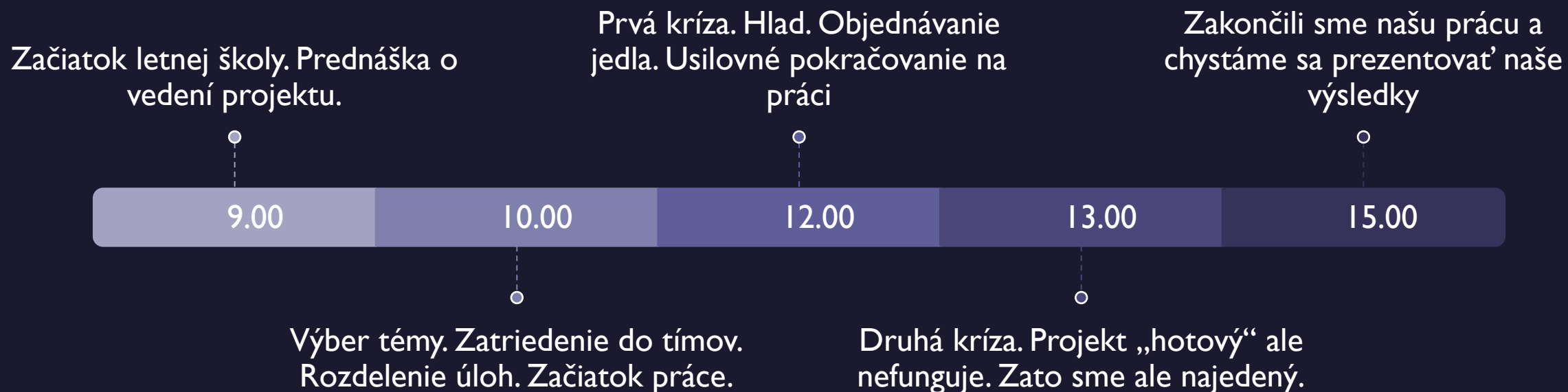
Výpočetne náročné

Vedecky podstatné

Riešenie



Timeline



Počiatkové podmienky

- Hmotnosť $M = 2 \cdot 10^5 M_{\odot}$
- Polomer $R = 50 \text{ ly}$
- Počet hviezd $N = 330\,000$
- Zgranulovanie $gran = 1000$
- Doba simulácie $T = 1\,000\,000$
- Časový krok $dt = 1000$



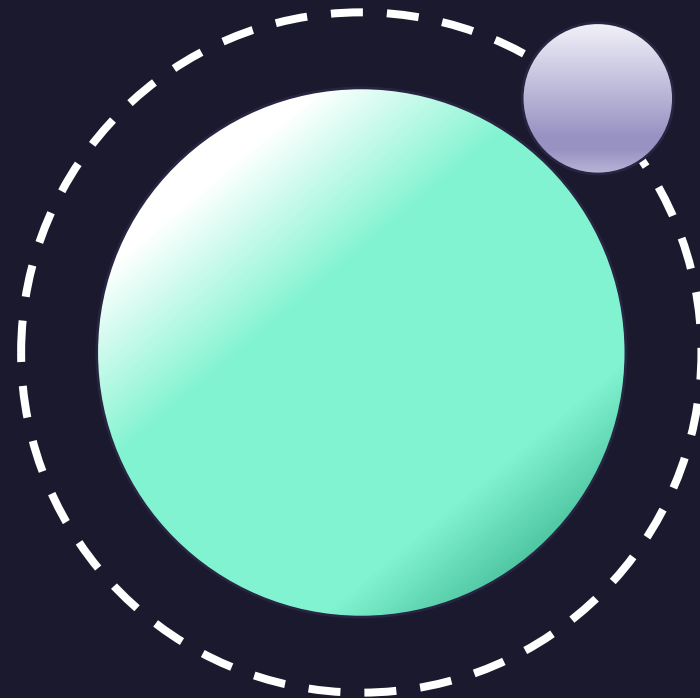
Rozdelenie hmotnosti a polôh

HUSTOTNÉ ROZDELENIE

- $\rho(r) = \frac{3M}{4\pi a^3 \left(1 + \frac{r^2}{a^2}\right)^{\frac{5}{2}}}$

PRIESTOROVÉ ROZDELENIE

- Prechod od spojitého rozdelenia hustoty k diskretným hmotnostiam



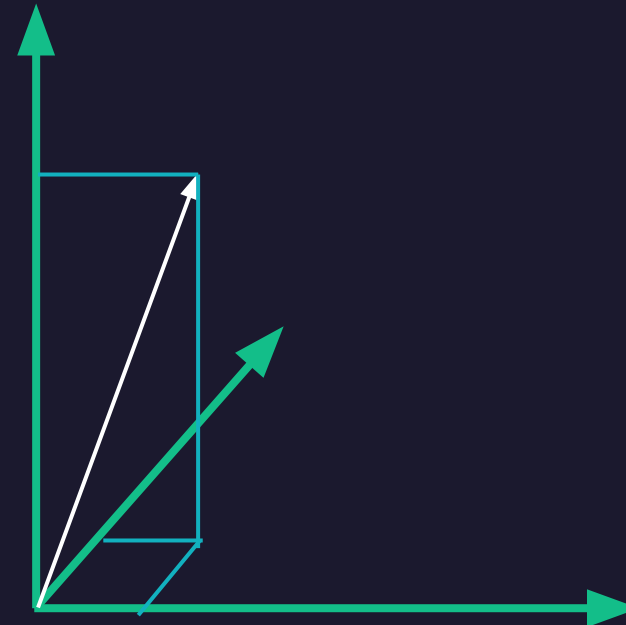
Náhodné rozdelenie rýchlosti



ROZDELENIE VEĽKOSTI RÝCHLOSTI

- $v(r) = \left(1 - \frac{r}{R+0.5}\right) \cdot 15 \text{ km/s}$

NÁHODNÝ PRIEMET



Kód

```
import random as rnd
import numpy as np
import tensorflow as tf
```

Format dat

[[[x1, y1, z1] , [x2, y2, z2] , [x3, y3, z3] , ..., [x1000, y1000, z1000]], v case 0

[[x1, y1, z1] , [x2, y2, z2] , [x3, y3, z3] , ..., [x1000, y1000, z1000]], v case 1

...,

[[x1, y1, z1] , [x2, y2, z2] , [x3, y3, z3] , ..., [x1000, y1000, z1000]] v case 100000

```
# Vygeneruje nahodne projekcie pre jednu zadanu rychlost
def vel_projection(vel):
    velx = rnd.randrange(np.floor(vel))
    vely = rnd.randrange(np.floor(np.sqrt(vel*vel - velx*velx)))
    velz = np.sqrt(vel*vel - velx*velx - vely*vely)
    return np.array([velx, vely, velz])
```

```
# Vygeneruje nahodne projekcie pre numpy array rychlosti
def vel_projections(vel):
    velx = np.random.random()*np.floor(vel)
    vely = np.random.random()*np.floor(np.sqrt(vel*vel - velx*velx))
    velz = np.sqrt(vel*vel - velx*velx - vely*vely)
    return velx, vely, velz
```

Pociatocne podmienky M92

```
M = 4*(10**35) # Hmotnost gulovej hviezdokopy v kg
R = 5*(10**16) # Polomer hviezdokopy v m
N = 330000 # Pocet hviezd v hviezdokope
gran = 1000 # Kolko hviezd reprezentuje jeden bod (granula)
n = N//gran # Prvy odhad Pocet bodov do simulacie
m = M/gran # Hmotnost jednej granule v kg
G = 6.6743*10**(-11) # Gravitacna konstanta
a = 0.6*R # Plumerova sfera
dt = 1000*365*24*60*60 # Casovy krok
T = 1000000*365*24*60*60 # Doba simulacie
Q = T//dt # Pocet krokov simulacie
```

Generovanie pociatocnych dat

```
pos, n = gen_pos(M=M, R=R, n=n, m=m, a=a)
vel = gen_vel(R=R, n=n, pos=pos)
vel = vel_vari(vel, delta=0.05)
```

get acceleration based on gravitational pull and update sub-clusters (granule) postions

```
def tf_delete(tensor, index, row=True):

    if row:
        sub = list(range(tensor.shape[0]))
    else:
        sub = list(range(tensor.shape[1]))
        sub.pop(index)

    if row:
        return tf.gather(tensor, sub)
    return tf.transpose(tf.gather(tf.transpose(tensor), sub))

def acc_by_G_pull(pos, m, G, dt, vel):
```

get acceleration based on gravitational pull and update sub-clusters (granule) postions

```
def tf_delete(tensor, index, row=True):

    if row:
        sub = list(range(tensor.shape[0]))
    else:
        sub = list(range(tensor.shape[1]))
        sub.pop(index)

    if row:
        return tf.gather(tensor, sub)
    return tf.transpose(tf.gather(tf.transpose(tensor), sub))
```

```
def acc_by_G_pull(pos, m, G, dt, vel):
```

```
pos_tf = tf.convert_to_tensor(pos, dtype=tf.float64)
trans_pos = tf.transpose(pos_tf)
```

```
m_temp = tf.convert_to_tensor(m, dtype=tf.float64)
m_tf = tf.fill(pos_tf.shape, m_temp)
```

```
G_tf = tf.convert_to_tensor(G, dtype=tf.float64)
vel_tf = tf.convert_to_tensor(vel, dtype=tf.float64)
```

```
x = pos_tf[:, 0:1]
y = pos_tf[:, 1:2]
z = pos_tf[:, 2:3]
```

```
dx = tf.transpose(x) - x
dy = tf.transpose(y) - y
dz = tf.transpose(z) - z
```

```
inv_r3 = (dx ** 2 + dy ** 2 + dz ** 2 + 0.1 ** 2) ** (-1.5)
```

```
ax = G_tf * (dx * inv_r3) @ m_tf
ay = G_tf * (dy * inv_r3) @ m_tf
az = G_tf * (dz * inv_r3) @ m_tf
```

```
ax = tf_delete(ax, 1, False)
ax = tf_delete(ax, 1, False)
ay = tf_delete(ay, 1, False)
ay = tf_delete(ay, 1, False)
az = tf_delete(az, 1, False)
az = tf_delete(az, 1, False)
```

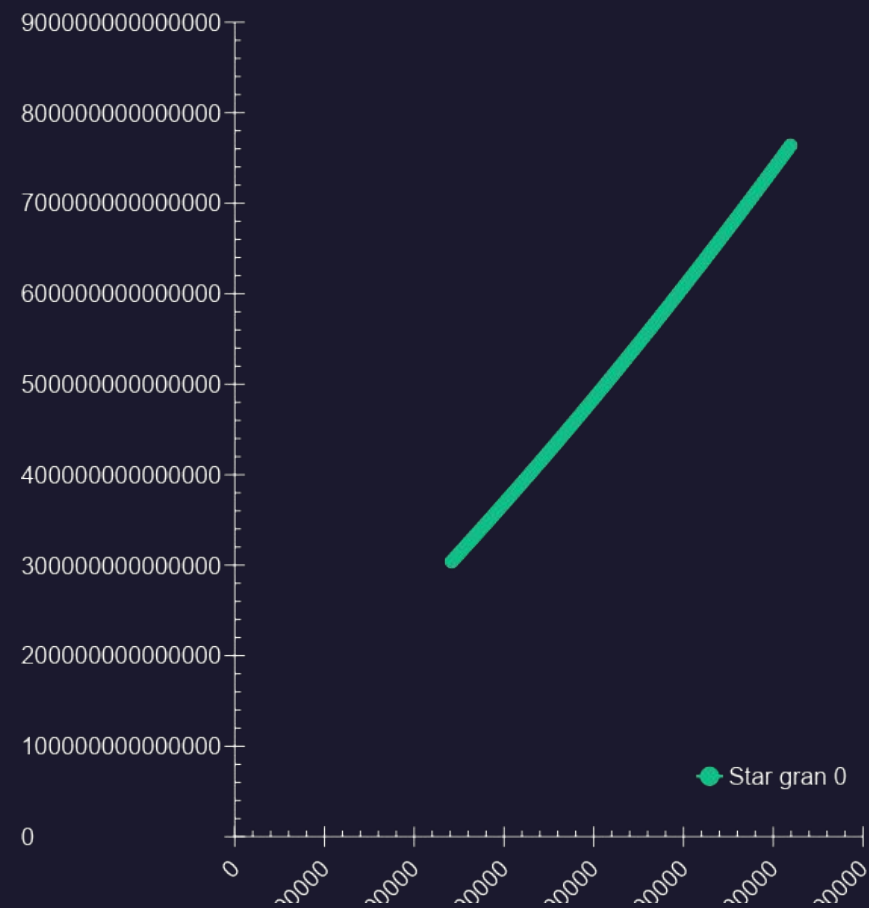
```
a = tf.concat([ax, ay, az], axis=1)
```

```
vel += a.numpy() * dt / 2.0
pos += vel * dt
```

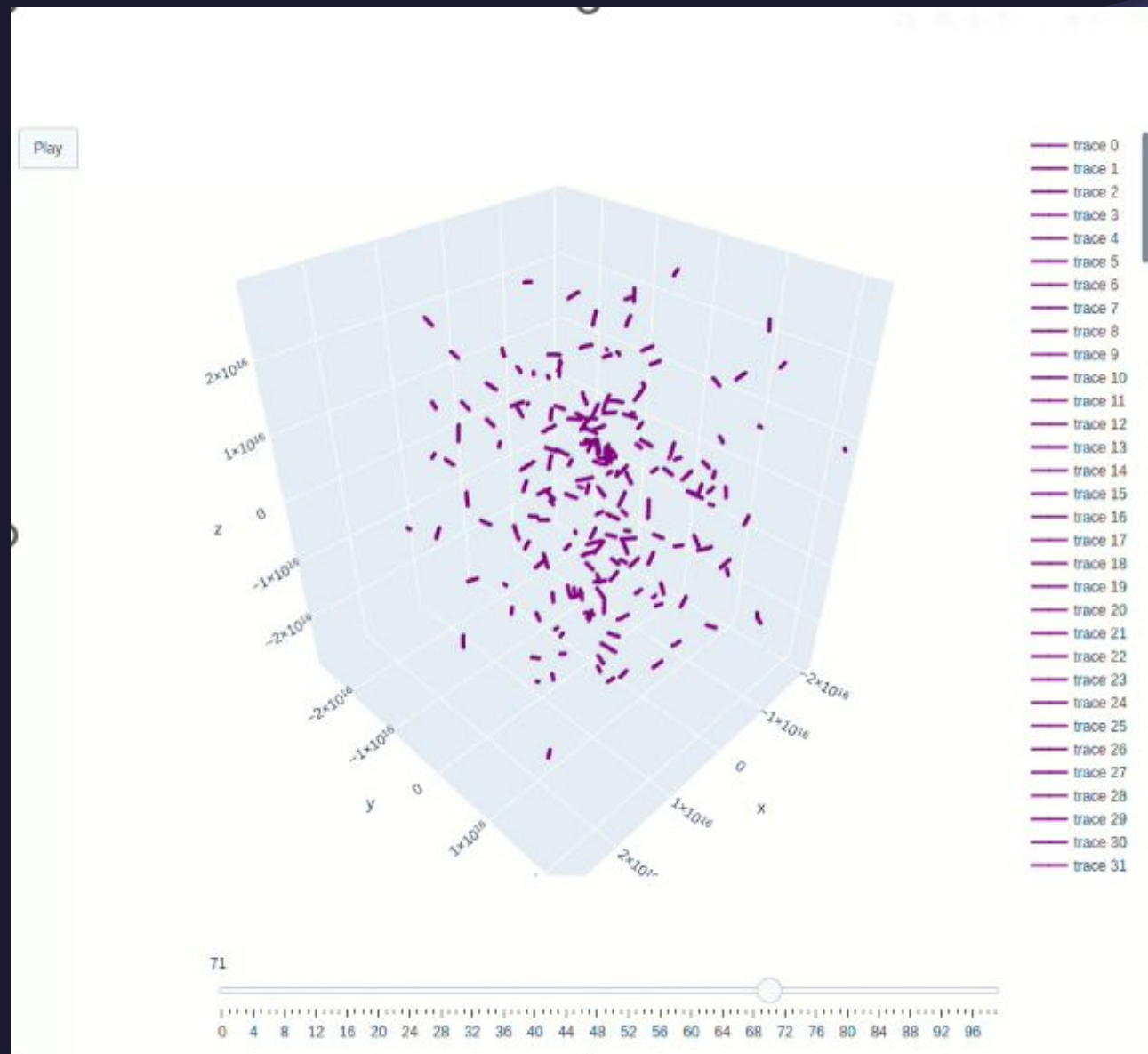
```
return vel, pos
```

```
acc_by_G_pull(pos, m, G, dt, vel)
```

Semi výsledky

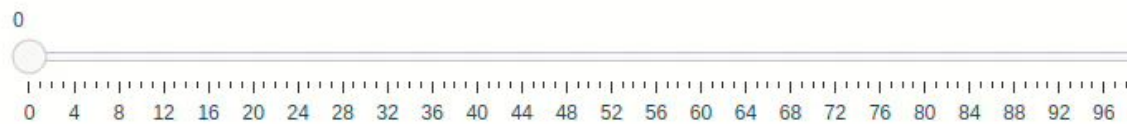


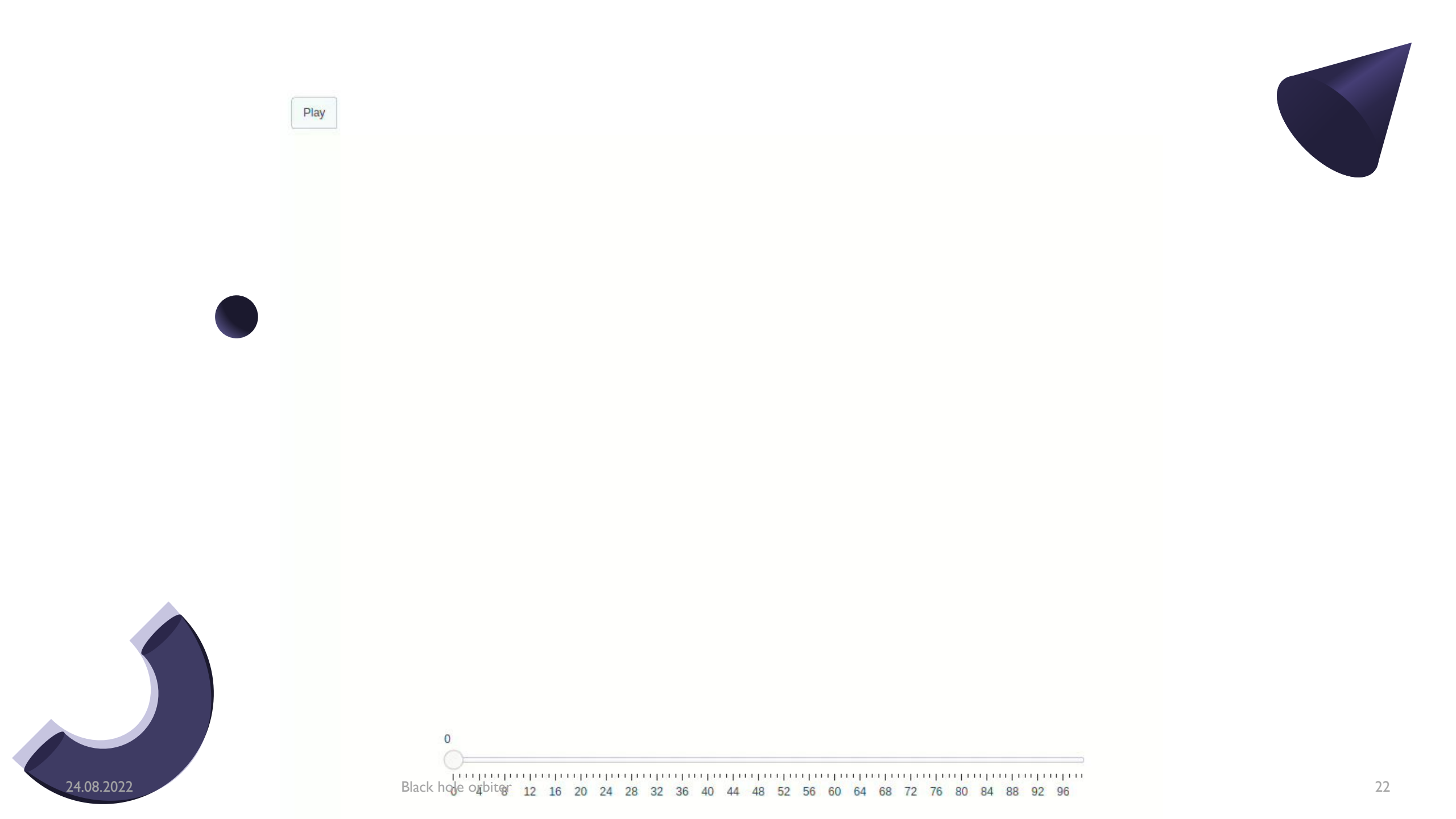
Lepšie výsledky





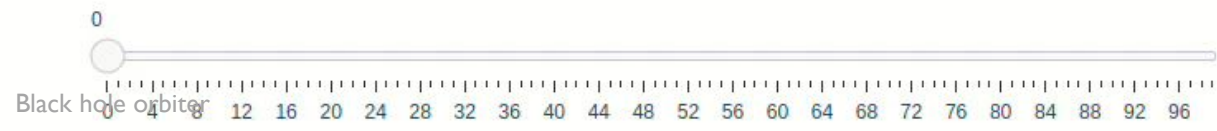
Play

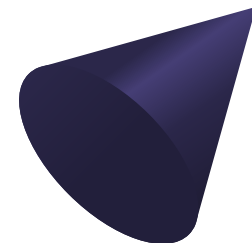




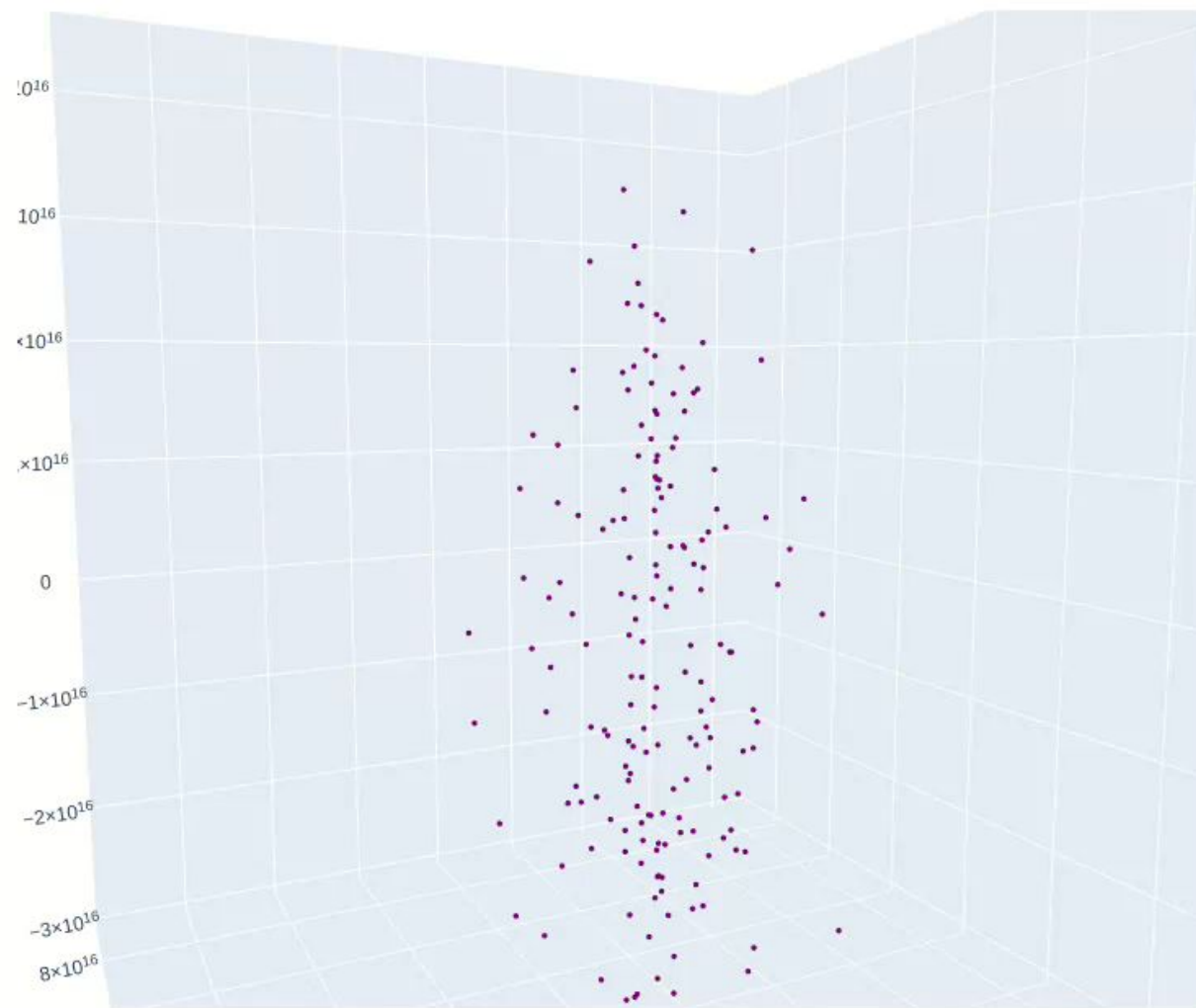
Play

24.08.2022

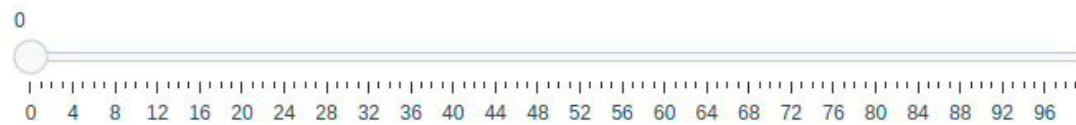




Play



- trace 0
- trace 1
- trace 2
- trace 3
- trace 4
- trace 5
- trace 6
- trace 7
- trace 8
- trace 9
- trace 10
- trace 11
- trace 12
- trace 13
- trace 14
- trace 15
- trace 16
- trace 17
- trace 18
- trace 19
- trace 20
- trace 21
- trace 22
- trace 23
- trace 24
- trace 25
- trace 26
- trace 27
- trace 28
- trace 29
- trace 30
- trace 31



For the future

ABY SIMULÁCIA FUNGOVALA



DASH VIZUALIZÁCIE



DODATOČNÉ VYLEPŠENIA





Záver

Ďakujem Vám za pozornosť

