```
In [2]:
 In [4]: from IPython.display import Image
             Type Markdown and LaTeX: \alpha^2
In [11]:
             print('TASK 1 A')
             Image("C:/Users/daria\Desktop\perseption\Problem_Set_3\PS3\photos\Task_1_A_res.png", width=320, height=240)
             TASK 1 A
Out[11]:
             (base) PS C:\Users\daria\besktop\perseption\from Tobtem_Stellan
Task 1A
Status of graph: Nodes = 1, Factors = 1, Eigen Factors = 0
Printing NodePose2d: 0, state =
               nting Factor: 0, obs=
In [24]:
            print('Code 1A')
             Image("C:/Users\daria\Desktop\perseption\Problem Set 3\PS3\photos\code 1A.png", width=320, height=240)
             Code 1A
Out[24]:
```

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```
In [12]:
    print('TASK 1 B')
    Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos\Task_1_B_res.png", width=320, height=240)
```

TASK 1 B

```
In [23]: print('Code 1 B')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos\code_1B.png", width=320, height=240)
```

Code 1 B

```
Out[23]:

def predict(self, v):
    print('Task1B')
    print('estimated-state_1',self.graph.get_estimated_state())
    self.pose_for_jac = self.graph.get_estimated_state()[self.pose].T[0]
    J = state_jacobian(self.pose_for_jac, v)[1]
    new_node = self.graph.add_node_pose_2d(np.zeros(3))

W_u = J @ get_motion_noise_covariance(u, self.alphas) @ J.T

    self.graph.add_factor_2poses_2d_odom(u, targs: self.pose, new_node, inv(W_u))
    self.pose = new_node
    print('estimated-state_2', self.graph.get_estimated_state())
```

TASK 1 C

In [13]: Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/task1_c_1.png", width=320, height=240)

```
Out[13]: Node: 0, Pos: [[180.]
          [ 50.]
          [ 0.]]
         Node: 1, Pos: [[190.]
          [ 50.]
          [ 0.]]
         Node: 4, Pos: [[200.]
          [ 50.]
          [ 0.]]
         Node: 5, Pos: [[210.]
          [ 50.]
          [ 0.]]
         Node: 6, Pos: [[220.]
          [ 50.]
          [ 0.]]
         Node: 7, Pos: [[230.]
          [ 50.]
          [ 0.]]
         Node: 8, Pos: [[240.]
          [ 50.]
          [ 0.]]
         Node: 9, Pos: [[250.]
          [ 50.]
          [ 0.]]
         Node: 11, Pos: [[260.]
          [ 50.]
          [ 0.]]
         Node: 12, Pos: [[270.]
          [ 50.]
          [ 0.]]
         Node: 13, Pos: [[280.]
          [ 50.]
          [ 0.]]
         Node: 14, Pos: [[290.]
          [ 50.]
          [ 0.]]
         Node: 15, Pos: [[300.]
          [ 50.]
          [ 0.]]
```

```
In [14]: Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/task_1_c_2.png", width=320, height=240)
```

Out[14]:

In [22]: print('code 1C')

Image("C:/Users\daria\Desktop\perseption\Problem Set 3\PS3\photos/code 1C.png", width=320, height=240)

code 1C

Out[22]:

New function gives to us more information. Gauss-Newton add more ditails about estimation and poses of our robot

TASK 1 D-

In [20]: Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/Task_1_D.png", width=320, height=240)

```
Out[20]:
        TaskD
        Node: 0, Pos: [[1.80000000e+02]
          [5.0000000e+01]
          [1.18460566e-13]]
         Node: 1, Pos: [[1.90000000e+02]
          [5.00001185e+01]
         [2.36921134e-05]]
         Node: 2, Pos: [[467.20476224]
         [ 0.88213239]]
        Node: 3, Pos: [[317.49146288]
         [ 7.73966435]]
        Node: 4, Pos: [[2.00045301e+02]
          [5.00186572e+01]
          [3.62517477e-03]]
         Node: 5, Pos: [[2.10054954e+02]
          [5.00671383e+01]
          [5.97725355e-031]
         Node: 6, Pos: [[2.20033540e+02]
          [5.01331359e+01]
          [7.16992733e-03]]
         Node: 7, Pos: [[2.30048079e+02]
          [5.02048981e+01]
          [7.16144571e-03]]
         Node: 8, Pos: [[2.40049923e+02]
          [5.02713796e+01]
          [6.11955049e-03]]
         Node: 9, Pos: [[2.50064066e+02]
          [5.03258702e+01]
          [4.78200064e-03]]
         Node: 10, Pos: [[2.60165377e+02]
          [5.03802453e+01]
          [6.06344313e-03]]
         Node: 11 Dos: [[2 702082816+02]
```

```
In [21]: print('Code 1D')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code_1D.png", width=320, height=240)
```

Code 1D

Out[21]:

```
92
93
94
95
#manual
96
main() > with movie_writer.saving(fig, a... > for t in tqdm(range(data.num_st...

LF UTF-8 4 s

93
94
95
solve(self, type):
96
print('TaskD')
97
self.graph.solve(type)
98
for i in range(len(self.graph.get_estimated_state())):
99
print(f'Node: {i},Pos: {self.graph.get_estimated_state()[i]}')
```

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```
In [16]: Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/Task_1_EF.png", width=320, height=240)
```

Out[16]: manual informat

manual information matrix 8.883613599764046e-16 diff between manual and auto 4.498945286664313e-14

We can see that both results of calculation really very closed. So we can say that using automatic functions or manually it is on you

Type *Markdown* and LaTeX: α^2

```
In [18]: print('Code task 1E-F')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code_1_EF.png", width=320, height=240)
```

Code task 1E-F

Out[18]:

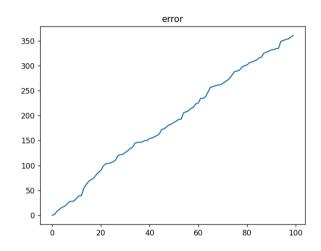
```
sam.update(z, Q)
state_1 = []
state_2 = []
st = sam.graph.get_estimated_state()
    flat = list(s.flatten())
    state_1 = state_1 + flat
state_1 = np.array(state_1)
sam.solve(mrob.GN)
st = sam.graph.get_estimated_state()
for s in st:
    state_2 = state_2 + flat
state_2 = np.array(state_2)
A = sam.graph.get_adjacency_matrix()
W = sam.graph.get_W_matrix()
I = sam.graph.get_information_matrix()
inf_matrix = A.todense().T @ W.todense() @ A.todense
norm_inf_m = norm(I - inf_matrix)
b = sam.graph.get_vector_b()
d_x = inv(inf_matrix) @ b
norm_d_x = norm(d_x - (state_1 - state_2))
print('manual information matrix', norm_inf_m_)
print('diff between manual and auto', norm_d_x)
```

TASK2

In [26]: print('TASK 2A')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/error.png", width=320, height=240)

TASK 2A

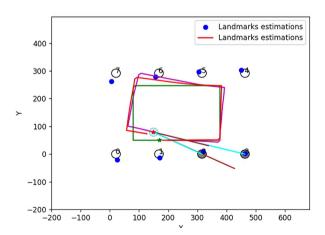
Out[26]:



In [28]: print('TASK 2B')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/task2b.png", width=320, height=240)

TASK 2B

Out[28]:



In [27]: print('Code 2B')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code2b.png", width=320, height=240)

Code 2B

Out[27]:

```
# TODO plot SLAM solution

pos_rob = []
pos_mark = []
new_pose = sam.graph.get_estimated_state()
for in in sam.landmarks.values():
    pos_rob.append(sam.graph.get_estimated_state()[im].T[0])
    new_pose[im] = '0'
    pos_rob = np.array(pos_rob)

for i in range(len(sam.graph.get_estimated_state())):
    if new_pose[i] != '0':
        pos_mark.append(new_pose[i])
    pos_mark = np.array(pos_mark)

plt.scatter(pos_rob[:t, 0], pos_rob[:t, 1], c='b', label='Landmarks estimations')

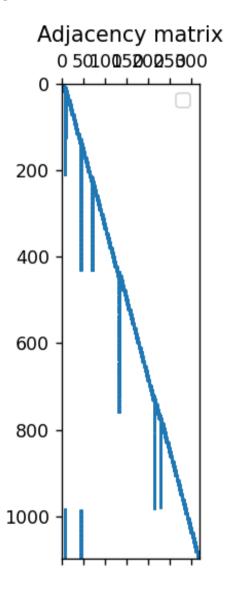
plt.plot(l'ags.pos_mark[:t, 0], pos_mark[:t, 1], c = 'r', label='Landmarks estimations')

if should_show_plots:
    Draw all the plots_and pause to create an animation effect_
    plt.plot(l'ags.pos_mark_x, pos_rob_x, 'r', label='Robot Estimated States')
    plt.plot(l'ags.pos_mark_x) pos_mark_y, 'bo', label='Robot Estimated States')
    plt.plot(l'ags.pos_mark_x) pos_mark_y, 'bo', label='Landmarks estimations')
    plt.legend(loc='upper right')
    plt.pause(args.plot_pause_len)
```

```
In [30]: print('TASK 2C')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/matrix_1.png", width=320, height=240)
```

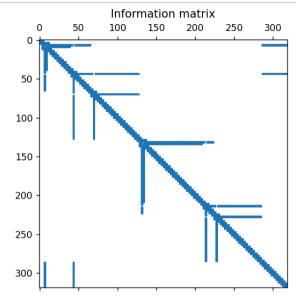
TASK 2C

Out[30]:



In [31]: Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/matrix_2.png", width=320, height=240)

Out[31]:



```
In [32]: print('Code 2C')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code2c.png", width=320, height=240)
```

Code 2C

Out[32]:

```
plt.title("Adjacency matrix")
plt.spy(sam.graph.get_adjacency_matrix())
plt.show(True)

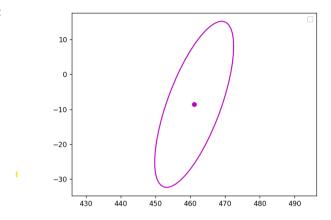
plt.title("Information matrix")
plt.spy(sam.graph.get_information_matrix())
plt.show(True)
```

We can see relationships between states and factors. We can see that information matrix has more relations

```
In [33]: print('TASK 2D')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/iso.png", width=320, height=240)
```

TASK 2D

Out[33]:



In [35]: print('Code 2D')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code2d.png", width=320, height=240)

Code 2D

Out[35]:
| State | Company | Compan

In [37]: print('TASK 2E')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/task2e.png", width=320, height=240)

TASK 2E

Out[37]:
Graph Solution: GN: i:3, chi2: 19.651626371764653

```
In [36]: print('Code 2E')
Image("C:/Users\daria\Desktop\perseption\Problem_Set_3\PS3\photos/code2e.png", width=320, height=240)
```

Code 2E

Out[36]:

```
for i in range(10):
    n = i + 1
    sam.solve(mrob.LM)
    print(f"Graph Solution: GN: i:{n}, chi2: {sam.graph.chi2()}")
    sam.solve(mrob.LM)
    __name__ == '__main__':
main()
```

In []: