```
In [27]: print("Overlap:", set(alz_ids) & set(cancer_ids))
Overlap: {'36321615', '36321363'}
```

A total of 2000 papers were extracted, of which two were duplicates. The speculation is that the reason is that the research areas overlap.

There are some missing texts, probably because the text is not in English

```
dictionary = {}
for i in alz_ids:
    title, abstract = get_info(i)
    dictionary[i] = {'ArticleTitle': title, 'ArticleAbstract': abstract, 'query': 'Alzheimer'}
for i in cancer_ids:|
    title, abstract = get_info(i)
    dictionary[i] = {'ArticleTitle': title, 'ArticleAbstract': abstract, 'query': 'Cancer'}
```

I used a dictionary format, using each PMID to correspond to an article, easy to sort.

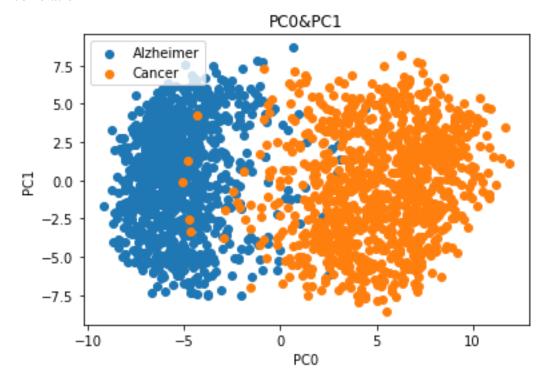
See the code for details.

EX2 Apply principal component analysis (PCA) to identify the first three principal components.

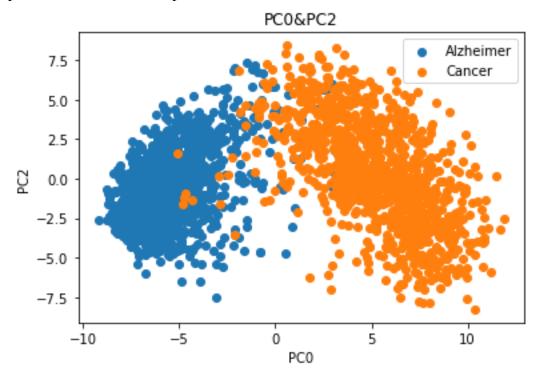
	PC0	PC1	PC2	query
0	-1.040141	-5.184434	-0.222884	Alzheimer
1	-5.402351	4.374268	0.928445	Alzheimer
2	-5.520820	4.610436	-1.998771	Alzheimer
3	-5.684639	-4.256922	-0.069922	Alzheimer
4	-5.909279	-2.143645	0.723237	Alzheimer
1993	1.974529	-4.315622	5.541237	Cancer
1994	5.472513	6.085035	3.709071	Cancer
1995	3.539508	-4.110456	2.859843	Cancer
1996	2.834004	-2.746227	4.597492	Cancer
1997	7.089354	-7.298180	2.489772	Cancer

1998 rows × 4 columns

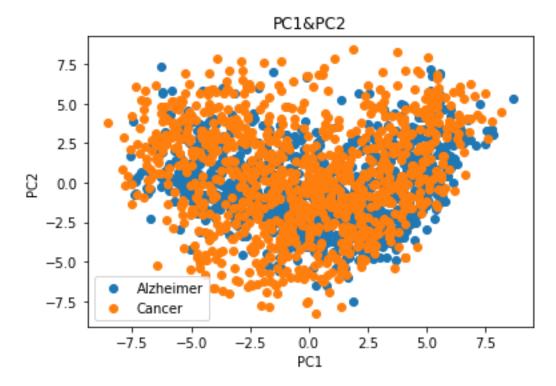
PC0 and PC1 are well separated and have close distribution directions, so there may be a co-relation



PC0 and PC2 are well separated, and the distribution direction is nearly vertical, with a possible linear relationship



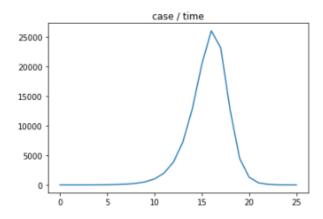
PC0 and PC2 are poorly separated and have nearly identical distributions and can be considered as co-occurring variables



```
In [86]: N = 134000
s, i, r, t = solve(N - 1, 1, 0, 2, 1, 30)
infected_drop = 0
while (i[infected_drop] >= 1):
    infected_drop += 1
print("The number of infected individuals drop below 1 at day:",infected_drop)
plt.plot(t[:infected_drop], i[:infected_drop])
plt.title("case / time")
```

The number of infected individuals drop below 1 at day: 26

```
Out[86]: Text(0.5, 1.0, 'case / time')
```



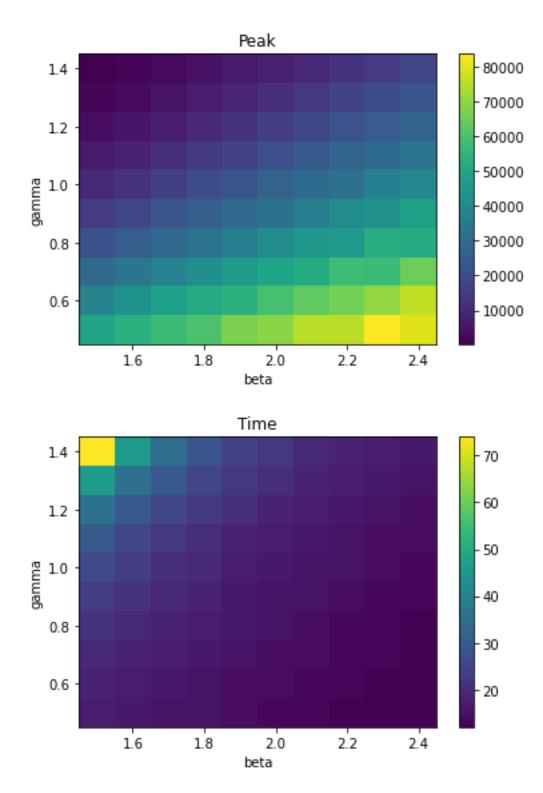
The number of infected individuals drop below 1 at day: 26

```
In [88]: N = 134000
s, i, r, t = solve(N - 1, 1, 0, 2, 1, 30)
day = 1
while (1):
    if (i[day] > i[day - 1]) and (i[day] > i[day + 1]):
        break
    day += 1
print ("Peak is day:",day,": Infected cases is:", i[day])
```

Peak is day: 16; Infected cases is: 26033.391521237274

Peak is day: 16

Infected cases are: 26033



The heatmaps if beta or gamma are not confirmed

https://www.kaggle.com/datasets/nareshbhat/health-care-data-set-on-heart-attack-possibility

I found a dataset on Kaggle about heart disease and causative factors. This dataset is a statistic on the likelihood of developing heart disease. There are 14 key variables, including age, sex, blood pressure, maximum heart rate, blood glucose, resting ECG results, and serum cholesterol. Key variables were not explicitly provided, but estimates could be derived from regression analysis. There were no derived variables, and each variable was indispensable. By simple observation only, there should be no variables that can be predicted by other variables.

The data format is CSV. there are 303 samples in the dataset. Including the header row there are 304 rows.

This dataset is open for learning purposes and does not require any additional permissions to gain full access.

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	target
0	63	1	3	145	233	1	Ō	150	Ō	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1
5	57	1	0	140	192	0	1	148	0	0.4	1	0	1	1
6	56	0	1	140	294	0	0	153	0	1.3	1	0	2	1
7	44	1	1	120	263	0	1	173	0	0.0	2	0	3	1
8	52	1	2	172	199	1	1	162	0	0.5	2	0	3	1
9	57	1	2	150	168	0	1	174	0	1.6	2	0	2	1
10	54	1	0	140	239	0	1	160	0	1.2	2	0	2	1
11	48	0	2	130	275	0	1	139	0	0.2	2	0	2	1
12	49	1	1	130	266	0	1	171	0	0.6	2	0	2	1
13	64	1	3	110	211	0	0	144	1	1.8	1	0	2	1
14	58	0	3	150	283	1	0	162	0	1.0	2	0	2	1
15	50	0	2	120	219	0	1	158	0	1.6	1	0	2	1
16	58	0	2	120	340	0	1	172	0	0.0	2	0	2	1
17	66	0	3	150	226	0	1	114	0	2.6	0	0	2	1

A preliminary analysis of the data reveals that the 14 variables have 13 independent variables and 1 dependent variable. 13 independent variables have categorical, dummy, discrete and continuous variables. The independent variables were divided into two main

categories: demographic data of the sample and medical indicators.

This dataset is very clean, no null values and no dirty data. The data types are uniform. Therefore, there is no need to clean up the data.

Code

EX1

```
In [12]: import requests
                                     import xml.dom.minidom as m
                                      import time as time
                                       import json
                                      import xml. etree. ElementTree as ET
In [19]: def get_PubMedIds(query, retmax=100):
                                                    get_rubmediac(query, retmax=100):
search_term = '&term=' + query
search_retmax = '&tertmax=" + str(retmax)
search_rettype = '&tetmode=json'
search_tritype = '&tetmode=json'
search_url = ''https://eutils.ncbi.nlm.nih.gov/entrez/eutils/esearch.fcgi?db=pubmed"+search_term+search_retmax+search_retty
r = requests.get(search_url)
                                                      return r. json()['esearchresult']['idlist']
                                                      AlzheimerTerm = 'Alzheimers+AND+2022[pdat]'
                                                     retmax = 1000
alz_ids = get_PubMedIds(AlzheimerTerm, retmax)
                                                    CancerTerm = 'Cancer+AND+2022[pdat]'
cancer_ids = get_PubMedIds(CancerTerm, retmax)
return alz_ids, cancer_ids
                                      alz_ids, cancer_ids = get_ids()
In [20]: print(len(alz_ids))
                                     print(len(cancer_ids))
                                       print(alz_ids)
                                     print(cancer_ids)
                                       1000
                                   1000
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In [21]: overlap=(set(cancer_ids)&set(alz_ids))
    print("overlap is:", overlap)
                                      overlap is: {'36321615', '36321363'}
```

```
In [16]: def get_info(pmid, deep = 0):
    time.sleep(1)
                      r = requests.get(
                       https://eutils.ncbi.nlm.nih.gov/entrez/eutils/efetch.fcgi?db=pubmed&retmode=xml&id=" + pmid"
                      tree = ET.fromstring(r.text)
                      try:
    titles = tree.find('PubmedArticle/MedlineCitation/Article/ArticleTitle')
    title = ET.tostring(titles, method="text").decode()
                      except:
                                 print("Error: {} No tittle max deep reached".format(pmid))
title = ""
                            else:
                      return get_info(pmid, deep + 1)
abstracts = tree.find('PubmedArticle/MedlineCitation/Article/Abstract/AbstractText')
                      try:
                            abstract = ET.tostring(abstracts, method="text").decode()
                            if deep > 5:
                                 print("Error: {} No abstract max deep reached".format(pmid))
                                  abstract =
                      else:
    return get_info(pmid, deep + 1)
return title, abstract
In [17]: def json_output()
                      dictionary = {}
for i in alz_ids:
                            title, abstract = get_info(i)
dictionary[i] = {'ArticleTitle': title, 'ArticleAbstract': abstract, 'query': 'Alzheimer'}
                     for i in cancer_ids:
    title, abstract = get_info(i)
    dictionary[i] = {'ArticleTitle': title, 'ArticleAbstract': abstract, 'query': 'Cancer'}
json_obj = json.dumps(dictionary, indent=4)
                      print(dictionary)
                     with open("alz&cancer.json", "w") as outfile:
    outfile.write(json_obj)
In [18]: json_output()
                Error: 36319136 No abstract max deep reached
               Error: 36316783 No abstract max deep reached Error: 36314232 No abstract max deep reached
                Error: 36310167 No abstract max deep reached
               Error: 36306458 No abstract max deep reached
Error: 36304998 No abstract max deep reached
Error: 36299613 No abstract max deep reached
                Error: 36284665 No abstract max deep reached
               Error: 36284252 No abstract max deep reached
Error: 36284251 No abstract max deep reached
Error: 36281688 No abstract max deep reached
                Error: 36281687 No abstract max deep reached
                Error: 36281660 No abstract max deep reached
                Error: 36281659 No abstract max deep reached
               Error: 36281659 No abstract max deep reached Error: 36281658 No abstract max deep reached Error: 36278375 No abstract max deep reached Error: 36278016 No abstract max deep reached Error: 36275013 No abstract max deep reached Error: 36271398 No abstract max deep reached
```

```
In [1]: pip install transformers
                Requirement already satisfied: transformers in d:\anaconda\lib\site-packages (4.24.0)
                Requirement already satisfied: transformers in d:\anaconda\lib\site-packages (4.4.0)
Requirement already satisfied: huggingface-hub\cl. 0, >=0.10.0 in d:\anaconda\lib\site-packages (from transformers) (0.10.1)
Requirement already satisfied: requests in d:\anaconda\lib\site-packages (from transformers) (2.27.1)
Requirement already satisfied: tokenizers!=0.11.3 \cdot 0.14.>=0.11.1 in d:\anaconda\lib\site-packages (from transformers) (6.0)
Requirement already satisfied: puypaml>=5.1 in d:\anaconda\lib\site-packages (from transformers) (6.0)
Requirement already satisfied: numpy>=1.17 in d:\anaconda\lib\site-packages (from transformers) (1.21.5)
Requirement already satisfied: tqdm>=4.27 in d:\anaconda\lib\site-packages (from transformers) (4.64.0)
                 Requirement already satisfied: regex!=2019.12.17 in d:\anaconda\lib\site-packages (from transformers) (2022.3.15)
                Requirement already satisfied: filelock in d:\anaconda\lib\site-packages (from transformers) (3.6.0)
                Requirement already satisfied: packaging>=20.0 in d:\anaconda\lib\site-packages (from transformers) (21.3)
Requirement already satisfied: typing-extensions>=3.7.4.3 in d:\anaconda\lib\site-packages (from huggingface-hub<1.0,>=0.10.0-
                Requirement already satisfied: pyparsing!=3.0.5, >=2.0.2 in d:\anaconda\lib\site-packages (from packaging>=20.0->transformers)
                 (3.0.4)
                Requirement already satisfied: colorama in d:\anaconda\lib\site-packages (from tqdm>=4.27->transformers) (0.4.4)
                Requirement already satisfied: idna(4,)=2.5 in d:\anaconda\lib\site=packages (from requests->transformers) (3.3)
Requirement already satisfied: charset-normalizer=2.0.0 in d:\anaconda\lib\site=packages (from requests->transformers) (2.0.
                Requirement already satisfied: certifi>=2017.4.17 in d:\anaconda\lib\site-packages (from requests->transformers) (2021.10.8)
                 Requirement already satisfied: urllib3<1.27,>=1.21.1 in d:\anaconda\lib\site-packages (from requests->transformers) (1.26.9)
                Note: you may need to restart the kernel to use updated packages.
                import torch
  In [1]: import json
                import tqdm
                import pandas as pd
                import matplotlib.pyplot as plt
  In [2]: from transformers import AutoTokenizer, AutoModel
                # load model and tokenizer
                tokenizer = AutoTokenizer.from_pretrained('allenai/specter')
                model = AutoModel.from_pretrained('allenai/specter')
                Downloading:
                                       0% [
                                                             | 0.00/440M [00:00<?, ?B/s]
                D:\Anaconda\lib\site-packages\huggingface_hub\file_download.py:123: UserWarning: `huggingface_hub` cache-system uses symlinks
                by default to efficiently store duplicated files but your machine does not support them in C:\USers\weipa\.cache\huggingface\hub. Caching files will still work but in a degraded version that might require more space on your disk. This warning can be disabled by setting the \u00e4HF_HUB_DISABLE_SYMLINKS_WARNING environment variable. For more details, see https://huggingface.co/docs/huggingface.hub/how-to-cache#limitations.
                To support symlinks on Windows, you either need to activate Developer Mode or to run Python as an administrator. In order to see activate developer mode, see this article: https://docs.microsoft.com/en-us/windows/apps/get-started/enable-your-device-for
                   warnings.warn(message)
In [32]: alz_cancer_papers = json.load(open('alz&cancer.json'))
print(len(alz_cancer_papers))
```

1000

```
In [33]: papers = alz_cancer_papers.copy()
                             # we can use a persistent dictionary (via shelve) so we can stop and restart if needed # alternatively, do the same but with embeddings starting as an empty dictionary embeddings = {}
                            for pmid, paper in tqdm.tqdm(papers.items()):
    data = [paper["ArticleTitle"] + tokenizer.sep_token + paper["ArticleAbstract"]]
    inputs = tokenizer(
                                                  data, padding=True, truncation=True, return_tensors="pt", max_length=512
                                        result = model(**inputs)
                                        * take the first token in the batch as the embedding embeddings[pmid] = result.last_hidden_state[:, 0, :].detach().numpy()[0]
                             # turn our dictionary into a list
embeddings = [embeddings[pmid] for pmid in papers.keys()]
                             100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 10
In [34]: from sklearn import decomposition pca = decomposition.PCA(n_components=3)
                             embeddings_pca = pd.DataFrame(
    pca.fit_transform(embeddings),
    columns=['PCO', 'PC1', 'PC2']
                             embeddings_pca["query"] = [paper["query"] for paper in papers.values()]
In [36]: embeddings_pca
 Out[36]:
                                   PC0
                                                                           PC1 PC2 query
                              0 -1.040141 -5.184434 -0.222884 Alzheimer
                                    1 -5.402351 4.374268 0.928445 Alzheimer
                              2 -5.520820 4.610436 -1.998771 Alzheimer
                                    3 -5.684639 -4.256922 -0.069922 Alzheimer
                               4 -5.909279 -2.143645 0.723237 Alzheimer
                               1993 1.974529 -4.315622 5.541237 Cancer
                               1994 5.472513 6.085035 3.709071 Cancer
                               1995 3.539508 -4.110456 2.859843 Cancer
                               1996 2.834004 -2.746227 4.597492 Cancer
                               1997 7.089354 -7.298180 2.489772 Cancer
```

1998 rows × 4 columns

```
In [54]: alz_PC0 = embeddings_pca[embeddings_pca['query']=='Alzheimer']['PC0']
alz_PC1 = embeddings_pca[embeddings_pca['query']=='Alzheimer']['PC1']
alz_PC2 = embeddings_pca[embeddings_pca['query']=='Alzheimer']['PC1']
cancer_PC0 = embeddings_pca[embeddings_pca['query']=='Cancer']['PC0']
cancer_PC1 = embeddings_pca[embeddings_pca['query']=='Cancer']['PC1']
cancer_PC2 = embeddings_pca[embeddings_pca['query']=='Cancer']['PC2']
                          plt.scatter(alz_PCO, alz_PCI, label="Alzheimer")
plt.scatter(cancer_PCO, cancer_PCI, label="Cancer")
plt.title("PCO#PCI")
plt.xlabel("PCO")
plt.ylabel("PCO")
plt.legend()
plt.show()
                                                                                                PC0&PC1
                                       5.0
                                       2.5
                               Q 0.0
                                    -2.5
                                    -5.0
                                              -10
In [55]: plt.scatter(alz_PC0, alz_PC2, label="Alzheimer")
   plt.scatter(cancer_PC0, cancer_PC2, label="Cancer")
   plt.title("PC0&PC2")
                           plt.xlabel("PCO")
plt.ylabel("PC2")
plt.legend()
                            plt.show()
                                                                                                PC0&PC2
                                       7.5
                                      2.5
                               0.0
                                    -2.5
                                     -5.0
                                    -7.5
  In [56]: plt.scatter(alz_PC1, alz_PC2, label="Alzheimer")
   plt.scatter(cancer_PC1, cancer_PC2, label="Cancer")
   plt.title("PC1&PC2")
                            plt.xlabel("PC1")
plt.ylabel("PC2")
                            plt.legend()
plt.show()
                                                                                                PC1&PC2
                                        7.5
                                        5.0
                               Q 0.0
                                     -2.5
                                     -7.5
                                                        -7.5
```

```
In [61]: import matplotlib.pyplot as plt
import numpy as np

def derivative(x, beta, gamma, N):
    s, i, r = x
    return np. array([-beta * s * i / N, beta * s * i / N - gamma * i, gamma * i])

# define standard explicit euler function
def explicit_euler(x0, t0, dt, T, beta, gamma, N):
    x = [x0]
    t = t0 + dt
    while t < T:
        x. append(x[-1] + dt * derivative(x[-1], beta, gamma, N))
        t += dt
    return x</pre>
```

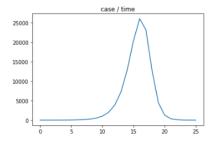
```
In [62]: # apply explicit euler to solve the SIR model
def solve(s0, i0, r0, beta, gamma, T):
    x0 = np. array([s0, i0, r0])
    t0 = 0
    N = s0 + i0 + r0
    dt = 1 # time step = 1 day
    x = explicit_euler(x0, t0, dt, T, beta, gamma, N)
    t = np. array(s). T
    return s, i, r = np. array(x). T
    return s, i, r, t

def SIR_model(s0, i0, r0, beta, gamma, T):
    s, i, r, t = solve(s0, i0, r0, beta, gamma, T)
    plt.plot(t, i)
```

```
In [86]:
N = 134000
s, i, r, t = solve(N - 1, 1, 0, 2, 1, 30)
infected_drop = 0
while (i[infected_drop] >= 1):
    infected_drop += 1
print("The number of infected individuals drop below 1 at day: ", infected_drop)
plt.plot(t[:infected_drop], i[:infected_drop])
plt.title("case / time")
```

The number of infected individuals drop below 1 at day: 26

Out[86]: Text(0.5, 1.0, 'case / time')



```
In [89]: N = 134000
s, i, r, t = solve(N - 1, 1, 0, 2, 1, 30)
day = 1
                         day = 1
while (1):
   if (i[day] > i[day - 1]) and (i[day] > i[day + 1]):
                        break
day += 1
print ("Peak is day: ", day, "; Infected cases are: ", i[day])
                          Peak is day: 16 ; Infected cases are: 26033.391521237274
In [84]: def heatmap():
    beta = np. arange(1.5, 2.5, 0.1)
    gamma = np. arange(0.5, 1.5, 0.1)
    peak = np. zeros((len(gamma), len(beta)))
    time = np. zeros((len(gamma), len(beta)))
    for i in range(len(beta)):
        for j in range(len(gamma)):
            ind, inf = find_peak(beta[i], gamma[j])
            peak[j][i] = inf
            time[j][i] = ind
        return peak, time
    peak, time = heatmap()
In [85]: beta = np.arange(1.5, 2.5, 0.1)
gamma = np.arange(0.5, 1.5, 0.1)
plt.pcolormesh(beta, gamma, peak)
plt.colorbar()
plt.xlabel("beta")
plt.ylabel("gamma")
plt.title("Peak")
plt.show()
                        plt.pcolormesh(beta, gamma, time)
plt.colorbar()
                        plt.colorbar()
plt.xlabel("beta")
plt.ylabel("gamma")
plt.title("Time")
plt.show()
                                                                          Peak
                                                                                                                                   80000
                                                                                                                                   70000
                                1.2
                                                                                                                                   50000
                            g 1.0
                                                                                                                                   40000
                                                                                                                                   30000
                                0.8
                                                                                                                                   20000
                                                                                                2.2
                                                                                                                 2.4
                                                1.6
                                                                 1.8
                                                                            2.0
beta
                                                                           Time
                                14
                                1.2
                            e 10
                                0.8
                                                                1.8
                                                                                                2.2
```

```
In [3]: import pandas as pd
                     df = pd.read_csv('heart.csv')
                     print(df.to_string())
                                                                 trestbps
145
130
130
120
120
140
140
140
172
172
150
140
130
130
110
150
                                                                                                                  restecg
0
1
                                                                                                                                        thalach
150
187
                                                                                                                                                                            oldpeak
2.3
3.5
1.4
0.8
0.6
0.4
1.3
0.0
0.5
1.6
1.2
0.2
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1.8
1.0
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0.0
2.6
                                                                                                                                                                                                thal
                                                                                                                                                                                                                                        target
                                                                                        chol fbs
233 1
250 0
204 0
236 0
354 0
192 0
263 0
199 1
168 0
239 0
275 0
266 0
211 0
283 1
219 0
340 0
                                                                                                                                                            age
63
37
41
56
57
57
56
44
52
57
54
48
49
64
58
50
58
66
                                                                                                                                                                                                                 1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
                                                 0
1
0
1
0
                                                          1
1
0
0
1
1
2
2
0
2
1
3
3
2
2
3
                                                                                                                                                                                                                                   2
2
1
2
3
                                                                                                                                                 163
                                                                                                                                                 148
153
173
162
174
160
                                                                                                                                                 139
171
                                                                                                                                                 144
162
                                                                              120
120
150
                                                 0
                                                                                                                                                 158
172
114
In [4]: df.isnull().sum()
 Out[4]: age sex
                                                 00000000
                      ср
                     cp
trestbps
chol
fbs
restecg
thalach
                      exang
oldpeak
                      slope
                                                 0 0 0
                      ca
thal
                      target 0
dtype: int64
In [5]: df.dtypes
                                                      int64
int64
int64
int64
int64
int64
int64
 Out[5]: age
                      sex
                     cp
trestbps
                     chol
fbs
restecg
thalach
                                                 int64
float64
int64
int64
                      exang
oldpeak
                      slope
ca
thal
                                                       int64
                      target
dtype: object
                                                       int64
```