Image Generation via Generative Adversarial Networks

import libraries

```
In [1]:
```

```
import torch
import torchvision
from torch.utils.data import Dataset
from os import listdir
from os.path import join
from torchvision import datasets, transforms
from torch.utils.data import DataLoader
from PIL import Image
import matplotlib.pyplot as plt
import numpy as np
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
from math import log10
from tqdm import tqdm
import os
```

load data

```
In [2]:
```

```
directory_data = './'
filename_data = 'assignment_12_data.npz'
data = np.load(os.path.join(directory_data, filename_data))
real = data['real_images']
```

```
In [3]:
```

```
print('size of real data:', real.shape)
```

```
size of real data: (4324, 32, 32)
```

plot data

In [4]:

```
def plot_image(title, image):
    nRow = 3
    nCol = 4
    size = 3

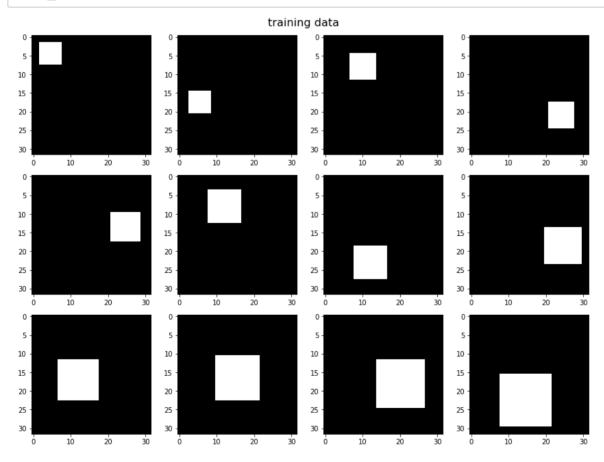
fig, axes = plt.subplots(nRow, nCol, figsize=(size * nCol, size * nRow))
    fig.suptitle(title, fontsize=16)

for r in range(nRow):
    for c in range(nCol):
        k = c * 300 + r * nCol * 300
        axes[r, c].imshow(image[k], cmap='gray')

plt.tight_layout()
    plt.show()
```

In [5]:

```
plot_image('training data', real)
```



custom data loader for the PyTorch framework

In [6]:

```
class dataset(Dataset):
    def __init__(self, image):
        self.image = image

    def __getitem__(self, index):
        sample = self.image[index]
        sample = torch.FloatTensor(sample).unsqueeze(dim=0)
    return (sample)

    def __len__(self):
        number_image = len(self.image)
    return (number_image)
```

setting device

```
In [7]:

device = torch.device('cuda' if torch.cuda.is_available() else 'mps')

In [8]:

print(device)

cuda

In [9]:

# random seed
import random
random.seed(20184757)
np.random.seed(20184757)
torch.manual_seed(20184757)
torch.cuda.manual_seed(20184757)
torch.cuda.manual_seed_all(20184757)
torch.backends.cudnn.deterministic = True
```

torch.backends.cudnn.benchmark = False

construct datasets and dataloaders

In [10]:

```
size_minibatch = 32
dim_latent = 64 # input latent vector fixed dimension

dataset_real = dataset(real)
dataloader_real = torch.utils.data.DataLoader(dataset_real, batch_size=size_mini batch, shuffle=True, drop_last=True)
```

neural networks

In [11]:

```
class Discriminator(nn.Module):
    def init (self):
        super(Discriminator, self). init ()
        self.conv = nn.Sequential(
            nn.Conv2d(in channels=1, out channels=16, kernel size=3, stride=2, p
adding=1, bias=True),
            nn.LeakyReLU(0.2),
            nn.Conv2d(in channels=16, out channels=32, kernel size=3, stride=2,
padding=1, bias=True),
            nn.LeakyReLU(0.2),
            nn.Conv2d(in channels=32, out channels=64, kernel size=3, stride=2,
padding=1, bias=True),
            nn.LeakyReLU(0.2),
            nn.Conv2d(in channels=64, out channels=128, kernel size=3, stride=2,
padding=1, bias=True),
            nn.LeakyReLU(0.2),
            nn.Conv2d(in channels=128, out channels=256, kernel size=3, stride=
2, padding=1, bias=True),
            nn.LeakyReLU(0.2),
        )
        self.fc = nn.Sequential(
            nn.Linear(256, 32),
            nn.LeakyReLU(0.2),
              nn.Linear(64, 32),
#
              nn.LeakyReLU(0.2),
#
              nn.Linear(32, 16),
#
              nn.LeakyReLU(0.2),
#
              nn.Linear(16, 8),
              nn.LeakyReLU(0.2),
            nn.Linear(32, 1)
        )
    def forward(self, x):
        out = self.conv(x)
        out = nn.Flatten()(out)
        out = self.fc(out)
        return out
```

In [12]:

```
class Generator(nn.Module):
   def init (self):
        super(Generator, self). init ()
        self.conv = nn.Sequential(
            nn.Upsample(scale factor=2, mode='bilinear', align corners=False),
            nn.Conv2d(in channels=dim latent, out channels=128, kernel size=3, s
tride=1, padding=1, bias=True),
            nn.BatchNorm2d(128),
            nn.LeakyReLU(0.2),
            nn.Upsample(scale factor=2, mode='bilinear', align corners=False),
            nn.Conv2d(in channels=128, out channels=64, kernel size=3, stride=1,
padding=1, bias=True),
           nn.BatchNorm2d(64),
            nn.LeakyReLU(0.2),
            nn.Upsample(scale factor=2, mode='bilinear', align corners=False),
            nn.Conv2d(in channels=64, out channels=32, kernel size=3, stride=1,
padding=1, bias=True),
            nn.BatchNorm2d(32),
            nn.LeakyReLU(0.2),
            nn.Upsample(scale factor=2, mode='bilinear', align corners=False),
            nn.Conv2d(in channels=32, out channels=16, kernel size=3, stride=1,
padding=1, bias=True),
            nn.BatchNorm2d(16),
            nn.LeakyReLU(0.2),
            nn.Upsample(scale factor=2, mode='bilinear', align corners=False),
            nn.Conv2d(in channels=16, out channels=1, kernel size=3, stride=1, p
adding=1, bias=True),
            nn.Sigmoid()
        )
   def forward(self, x):
        out = self.conv(x)
        return out
```

build network

```
In [13]:
```

```
def weights_init(m):
    classname = m.__class__.__name__
    if classname.find("Conv") != -1:
        nn.init.xavier_uniform_(m.weight)
        nn.init.constant_(m.bias, 1.)
    elif classname.find("BatchNorm") != -1:
        nn.init.constant_(m.weight, 1.)
        nn.init.constant_(m.bias, 1.)
    elif classname.find("Linear") != -1:
        nn.init.xavier_uniform_(m.weight)
        nn.init.constant_(m.bias, 1.)
```

In [14]:

```
generator = Generator().to(device)
discriminator = Discriminator().to(device)

generator = generator.apply(weights_init)
discriminator = discriminator.apply(weights_init)

optimizer_generator = torch.optim.Adam(generator.parameters(), lr=0.001, betas=(0.5, 0.999))
optimizer_discriminator = torch.optim.Adam(discriminator.parameters(), lr=0.001, betas=(0.5, 0.999))
```

compute the prediction

```
In [15]:
```

```
def compute_prediction(model, input):
    prediction = model(input)
    return prediction
```

compute the accuracy

In [16]:

```
def compute_bounding_mask(prediction_binary):
    bounding_mask = torch.zeros_like(prediction_binary)

for i in range(prediction_binary.shape[0]):
    prediction = prediction_binary[i]
    (idx_row, idx_col) = torch.nonzero(prediction, as_tuple=True)

if len(idx_row) > 0:
    idx_row_min = idx_row.min()
    idx_row_max = idx_row.max()

idx_col_min = idx_col.min()
    idx_col_max = idx_col.max()

    bounding_mask[i, idx_row_min:idx_row_max+1, idx_col_min:idx_col_max+1] = 1

return bounding_mask
```

In [17]:

```
def compute_accuracy(prediction):
    binary = (prediction >= 0.5)
    mask = compute_bounding_mask(binary)

intersection = (binary & mask).float().sum((1, 2))
    union = (binary | mask).float().sum((1, 2))

eps = 1e-8
    correct = (intersection + eps) / (union + eps)
    accuracy = correct.mean() * 100.0
    accuracy = accuracy.cpu()
```

train

In [18]:

```
def train(generator, discriminator, dataloader):
    loss generator epoch = []
   loss discriminator epoch = []
   for index batch, (real) in enumerate(dataloader):
                  = real.to(device)
       real
       size batch = len(real)
                  = torch.randn(size batch, dim latent, device=device)
                  = torch.reshape(latent, [size batch, dim latent, 1, 1])
       latent
       # update the generator
       generator.train()
       discriminator.eval()
       optimizer generator.zero grad()
       fake = compute prediction(generator, latent)
       prediction fake = compute prediction(discriminator, fake)
       label fake = torch.ones like(prediction fake)
       loss label = nn.BCEWithLogitsLoss()(prediction fake, label fake)
       loss generator = loss label
       loss generator.backward()
       optimizer generator.step()
        # -----
       # update the discriminator
       generator.eval()
       discriminator.train()
       optimizer discriminator.zero grad()
       fake = compute prediction(generator, latent)
       prediction real = compute prediction(discriminator, real)
       prediction fake = compute prediction(discriminator, fake)
       loss real = nn.BCEWithLogitsLoss()(prediction real, torch.ones like(pred
iction real, device=device))
       loss_fake = nn.BCEWithLogitsLoss()(prediction_fake, torch.zeros_like(pre
diction fake, device=device))
       loss discriminator = (loss real + loss fake) / 2.
       loss_discriminator.backward()
       optimizer discriminator.step()
       loss generator epoch.append(loss generator.item())
       loss_discriminator_epoch.append(loss_discriminator.item())
   return np.mean(loss generator epoch), np.mean(loss discriminator epoch)
```

test

```
In [19]:
```

```
def test(generator, latent):
    generator.eval()

    prediction = generator(latent)
    prediction = prediction.squeeze(axis=1)
    accuracy = compute_accuracy(prediction)

return accuracy
```

training epoch

In [20]:

```
number epoch
               = 2000
accuracy epoch = np.zeros(number epoch)
loss generator = np.zeros(number epoch)
loss discriminator = np.zeros(number epoch)
num latent test = 100
latent test = torch.randn(num latent test, dim latent, device=device)
latent test = torch.reshape(latent test, [num latent test, dim latent, 1, 1])
for i in range(number epoch):
   loss generator epoch, loss discriminator epoch = train(generator, discrimina
tor, dataloader real)
   accuracy = test(generator, latent test)
   accuracy epoch[i] = accuracy
   loss generator[i] = loss generator epoch
   loss discriminator[i] = loss discriminator epoch
   print(f'epoch {i}, loss_generator = {loss_generator_epoch}, loss_discriminat
or={loss discriminator epoch}, accuracy={accuracy}')
   # if accuracy>98 stop
   if i>10 and accuracy>98:
        accuracy_epoch = accuracy_epoch[:i+1]
        break
```

epoch 0, loss generator = 0.8082629900839594, loss discriminator=0.7 643318666352166, accuracy=84.20942687988281 epoch 1, loss generator = 0.7027878527288084, loss discriminator=0.6 972652969536958, accuracy=87.51338958740234 epoch 2, loss generator = 0.7522311550599557, loss discriminator=0.7 334282446790624, accuracy=91.7977294921875 epoch 3, loss generator = 0.6983382472285518, loss discriminator=0.6 929080552524991, accuracy=90.97225189208984 epoch 4, loss generator = 0.6983237134085761, loss discriminator=0.6 935412967646564, accuracy=92.1667251586914 epoch 5, loss generator = 0.7589374214962676, loss discriminator=0.7 358458081881205, accuracy=100.0 epoch 6, loss generator = 0.6913192913488105, loss discriminator=0.6 869762160159923, accuracy=89.08001708984375 epoch 7, loss generator = 0.9522485311660501, loss discriminator=0.6 915906588236491, accuracy=93.10673522949219 epoch 8, loss generator = 1.1243618664918122, loss discriminator=0.5 409262560032032, accuracy=80.57636260986328 epoch 9, loss generator = 1.3604767507976956, loss discriminator=0.4 761901090542475, accuracy=76.34229278564453 epoch 10, loss generator = 1.3620896498362223, loss discriminator=0. 474625191644386, accuracy=77.29827117919922 epoch 11, loss generator = 1.5573338956744582, loss discriminator=0. 4249078617051796, accuracy=81.41233825683594 epoch 12, loss_generator = 1.7527882244851853, loss_discriminator=0. 3811194009251065, accuracy=76.98038482666016 epoch 13, loss generator = 1.960030887303529, loss discriminator=0.3 5012611548105876, accuracy=78.46669006347656 epoch 14, loss generator = 2.273727814356486, loss discriminator=0.2 787886160943243, accuracy=80.28314971923828 epoch 15, loss generator = 2.4653858259872155, loss discriminator=0. 29219485813820806, accuracy=81.42350006103516 epoch 16, loss generator = 3.0391117316705207, loss discriminator=0. 20908179531494778, accuracy=82.75890350341797 epoch 17, loss generator = 3.580859975461607, loss discriminator=0.1 7330280393362046, accuracy=84.18598937988281 epoch 18, loss generator = 3.7706255082730893, loss discriminator=0. 15167678067529644, accuracy=83.10103607177734 epoch 19, loss generator = 4.622254577389469, loss discriminator=0.1 2233960833400488, accuracy=85.0777359008789 epoch 20, loss generator = 4.426264007003219, loss discriminator=0.1 2268856842484739, accuracy=84.14549255371094 epoch 21, loss generator = 5.024055301922339, loss discriminator=0.1 3281221559478176, accuracy=87.44444274902344 epoch 22, loss generator = 4.31471023912783, loss discriminator=0.18 808373814379728, accuracy=87.71733093261719 epoch 23, loss generator = 4.307887722386254, loss discriminator=0.1 9419753308649415, accuracy=87.39325714111328 epoch 24, loss generator = 4.003738592289112, loss discriminator=0.2 3217686889348207, accuracy=82.76016998291016 epoch 25, loss generator = 3.4664252936840056, loss discriminator=0. 24246207078297932, accuracy=88.41580200195312 epoch 26, loss generator = 3.649142809708913, loss discriminator=0.2 4145819340591076, accuracy=89.46354675292969 epoch 27, loss generator = 3.585045764181349, loss discriminator=0.2 379020240019869, accuracy=89.93412017822266 epoch 28, loss generator = 3.9421526149467185, loss discriminator=0. 19966096276486361, accuracy=89.08478546142578 epoch 29, loss generator = 3.8899481417956174, loss discriminator=0.

24355100558863746, accuracy=91.60093688964844 epoch 30, loss generator = 3.9328899423281354, loss discriminator=0. 2145877868488983, accuracy=86.38325500488281 epoch 31, loss generator = 4.266624908756326, loss discriminator=0.2 3212754037921077, accuracy=85.96870422363281 epoch 32, loss generator = 3.3968014997464637, loss discriminator=0. 3183640140074271, accuracy=87.832763671875 epoch 33, loss generator = 4.094814695252312, loss discriminator=0.2 476238193611304, accuracy=91.5387954711914 epoch 34, loss generator = 4.4703655852211845, loss discriminator=0. 21463038879964086, accuracy=92.55770111083984 epoch 35, loss generator = 4.263258313028901, loss discriminator=0.2 3003059148236557, accuracy=90.92512512207031 epoch 36, loss generator = 5.015959831520363, loss discriminator=0.1 8908967309527927, accuracy=91.64525604248047 epoch 37, loss generator = 5.68901817533705, loss discriminator=0.16 798959203340388, accuracy=90.32850646972656 epoch 38, loss generator = 6.077758192133021, loss discriminator=0.1 6120823980481536, accuracy=91.03966522216797 epoch 39, loss generator = 5.553621039567171, loss discriminator=0.1 7233437316285238, accuracy=90.04877471923828 epoch 40, loss generator = 4.9196132456814805, loss discriminator=0. 1660916232124523, accuracy=88.65951538085938 epoch 41, loss generator = 4.924588997275741, loss discriminator=0.1 9302286279422265, accuracy=89.90026092529297 epoch 42, loss generator = 4.635873254140218, loss discriminator=0.1 8597089537867792, accuracy=91.55152130126953 epoch 43, loss generator = 4.7808431263875075, loss discriminator=0. 2811540625161595, accuracy=90.78208923339844 epoch 44, loss generator = 3.728647326540064, loss discriminator=0.2 914534136101052, accuracy=90.42366790771484 epoch 45, loss generator = 4.289696999170162, loss discriminator=0.2 782743180239642, accuracy=91.83181762695312 epoch 46, loss generator = 4.804126509472176, loss discriminator=0.2 2151669672241917, accuracy=91.35775756835938 epoch 47, loss generator = 4.353231819470723, loss discriminator=0.2 963639687332842, accuracy=92.21605682373047 epoch 48, loss generator = 5.590406799316407, loss discriminator=0.1 981130360453217, accuracy=89.69986724853516 epoch 49, loss generator = 5.27590650540811, loss discriminator=0.21 120277803253246, accuracy=92.37712097167969 epoch 50, loss generator = 5.833990669250488, loss discriminator=0.1 79158312689375, accuracy=93.57933044433594 epoch 51, loss generator = 3.155111863160575, loss discriminator=0.3 649993650891163, accuracy=92.63752746582031 epoch 52, loss generator = 3.0244055911346717, loss discriminator=0. 31205638729863694, accuracy=92.34168243408203 epoch 53, loss generator = 5.279647572614529, loss discriminator=0.2 300964398516549, accuracy=91.22415161132812 epoch 54, loss_generator = 5.565741634810412, loss_discriminator=0.2 2188065703268403, accuracy=92.83433532714844 epoch 55, loss_generator = 5.2192006691738415, loss_discriminator=0. 2503238366709815, accuracy=91.41524505615234 epoch 56, loss generator = 5.930626337616532, loss discriminator=0.2 011133623343927, accuracy=91.63658142089844 epoch 57, loss generator = 6.2713178407262875, loss discriminator=0. 2225825733884617, accuracy=92.82865905761719 epoch 58, loss generator = 5.032997703552246, loss discriminator=0.2 6264107718511864, accuracy=91.29759979248047

epoch 59, loss generator = 6.137228870391846, loss discriminator=0.2 047358697211301, accuracy=90.33395385742188 epoch 60, loss generator = 6.218489473837393, loss discriminator=0.1 9614289689947056, accuracy=90.59857177734375 epoch 61, loss generator = 5.963494848321985, loss discriminator=0.2 0990066448295558, accuracy=90.28412628173828 epoch 62, loss generator = 6.02463938748395, loss discriminator=0.16 730702530454705, accuracy=92.13375091552734 epoch 63, loss generator = 5.801383698428118, loss discriminator=0.1 7522610006509004, accuracy=93.39067077636719 epoch 64, loss generator = 5.856387018274378, loss discriminator=0.1 4415361299007026, accuracy=91.99081420898438 epoch 65, loss_generator = 5.34918668711627, loss discriminator=0.15 64818392197291, accuracy=93.04478454589844 epoch 66, loss generator = 5.329869121975369, loss discriminator=0.1 614993942556558, accuracy=91.05958557128906 epoch 67, loss_generator = 5.037941749007613, loss_discriminator=0.1 7389597372600327, accuracy=92.12467193603516 epoch 68, loss generator = 5.211588804810136, loss discriminator=0.1 6734876577500943, accuracy=94.98899841308594 epoch 69, loss generator = 3.6006228937043083, loss discriminator=0. 3027732202300319, accuracy=92.1993637084961 epoch 70, loss generator = 4.557053037926003, loss discriminator=0.2 173115169008573, accuracy=91.87950134277344 epoch 71, loss generator = 4.818367957185816, loss discriminator=0.2 548279994063907, accuracy=91.03377532958984 epoch 72, loss generator = 5.621056161103425, loss discriminator=0.2 550302648985827, accuracy=93.7872085571289 epoch 73, loss generator = 5.623464893411707, loss discriminator=0.2 4457846929629642, accuracy=90.66010284423828 epoch 74, loss generator = 5.483881175959552, loss discriminator=0.2 4715743285638314, accuracy=89.63140869140625 epoch 75, loss generator = 9.533017230916906, loss discriminator=0.0 9703522027056251, accuracy=81.46532440185547 epoch 76, loss_generator = 7.446395310649166, loss discriminator=0.1 6692125524083773, accuracy=92.42424011230469 epoch 77, loss generator = 6.684698507520888, loss discriminator=0.1 9940119280859275, accuracy=90.4032974243164 epoch 78, loss_generator = 5.672221038959645, loss_discriminator=0.2 2358151420398995, accuracy=90.1744384765625 epoch 79, loss generator = 5.9573861863878035, loss discriminator=0. 19885778973499935, accuracy=92.44161224365234 epoch 80, loss generator = 5.484625477261013, loss discriminator=0.2 3409820270759088, accuracy=91.92988586425781 epoch 81, loss generator = 6.105573742477982, loss discriminator=0.1 9224909612977947, accuracy=91.90402221679688 epoch 82, loss generator = 7.325678240811383, loss discriminator=0.1 552903743253814, accuracy=92.7491455078125 epoch 83, loss generator = 5.024997290417, loss discriminator=0.2188 1565617190468, accuracy=92.99227142333984 epoch 84, loss generator = 5.635455205705431, loss discriminator=0.1 818334893495948, accuracy=91.79354858398438 epoch 85, loss_generator = 5.889366103984692, loss discriminator=0.1 7044834858841365, accuracy=93.52947998046875 epoch 86, loss generator = 5.585920856617115, loss discriminator=0.1 8281907226752353, accuracy=92.6313247680664 epoch 87, loss_generator = 5.721853271237126, loss_discriminator=0.2 0001092503468196, accuracy=91.81695556640625 epoch 88, loss_generator = 5.315864877789108, loss discriminator=0.2 3878052565786573, accuracy=91.72125244140625 epoch 89, loss generator = 6.284599477273447, loss discriminator=0.2 1627099215984344, accuracy=93.88388061523438 epoch 90, loss generator = 6.62302657145041, loss discriminator=0.19 913109593369344, accuracy=93.50154113769531 epoch 91, loss generator = 6.25126999175107, loss discriminator=0.21 687889479928546, accuracy=94.61326599121094 epoch 92, loss generator = 6.694562228520711, loss discriminator=0.1 8581741744721378, accuracy=92.25569152832031 epoch 93, loss generator = 4.810653353178942, loss discriminator=0.3 114408669234426, accuracy=92.96257781982422 epoch 94, loss generator = 5.9433038446638315, loss discriminator=0. 23577634670116282, accuracy=92.69552612304688 epoch 95, loss generator = 6.514171702331967, loss discriminator=0.2 4001952227618958, accuracy=92.31243896484375 epoch 96, loss generator = 6.042556881463086, loss discriminator=0.2 3588223137237407, accuracy=94.51895141601562 epoch 97, loss generator = 6.617071040471395, loss discriminator=0.1 99194867439844, accuracy=92.70243072509766 epoch 98, loss generator = 6.011558528299685, loss discriminator=0.1 9210915424757533, accuracy=91.1521987915039 epoch 99, loss generator = 6.222189050250583, loss discriminator=0.1 8822810981008742, accuracy=92.8996353149414 epoch 100, loss generator = 6.57770035178573, loss discriminator=0.1 854261209015493, accuracy=92.46334075927734 epoch 101, loss generator = 5.719110386459915, loss discriminator=0. 23581910028501793, accuracy=91.81687927246094 epoch 102, loss generator = 6.509247504340278, loss discriminator=0. 2147956114124369, 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epoch 413, loss generator = 12.456725410178855, loss discriminator= 0.09347267719330611, accuracy=96.39059448242188 epoch 414, loss generator = 12.166135349980108, loss discriminator= 0.08476737354088712, accuracy=96.75149536132812 epoch 415, loss generator = 10.345053799947102, loss discriminator= 0.08799597434699535, accuracy=95.91967010498047 epoch 416, loss generator = 10.246588975411875, loss discriminator= 0.09255077438773933, accuracy=97.20438385009766 epoch 417, loss generator = 10.561509086467602, loss discriminator= 0.08939268718715067, accuracy=96.66439819335938 epoch 418, loss generator = 10.805800497973406, loss discriminator= 0.09398343841786738, accuracy=96.75873565673828 epoch 419, loss generator = 9.095091952217949, loss discriminator=0. 10131709048593486, accuracy=95.87114715576172 epoch 420, loss generator = 10.049164206893355, loss discriminator= 0.10610545185153131, accuracy=95.82950592041016 epoch 421, loss generator = 11.332076436501962, loss discriminator= 0.09342819537829469, accuracy=96.0008316040039 epoch 422, loss generator = 11.72525279433639, loss discriminator=0. 07864015901944152, accuracy=95.62948608398438 epoch 423, loss generator = 10.404781853711164, loss discriminator= 0.07393311888531402, accuracy=95.25654602050781 epoch 424, loss generator = 9.713429108372441, loss discriminator=0. 07630198423026337, accuracy=95.5407943725586 epoch 425, loss generator = 9.821238351751257, loss discriminator=0. 08221289543952379, accuracy=95.05409240722656 epoch 426, loss generator = 9.064047212953922, loss discriminator=0. 0810910703200433, accuracy=96.32391357421875 epoch 427, loss generator = 9.231933268794307, loss discriminator=0. 09100220259968882, accuracy=96.47509765625 epoch 428, loss generator = 8.991849048049362, loss discriminator=0. 08433043653904287, accuracy=96.72195434570312 epoch 429, loss generator = 8.329735925462511, loss discriminator=0. 08126918431716385, accuracy=96.56846618652344 epoch 430, loss_generator = 8.732350527798689, loss discriminator=0. 11077650403948845, accuracy=96.54327392578125 epoch 431, loss generator = 8.207335502130015, loss discriminator=0. 11611758425003953, accuracy=96.72345733642578 epoch 432, loss_generator = 8.470511750821714, loss_discriminator=0. 11458271630108356, accuracy=96.59849548339844 epoch 433, loss generator = 9.404518692581743, loss discriminator=0. 10621802311528612, accuracy=96.65312957763672 epoch 434, loss generator = 8.632697886007804, loss discriminator=0. 11494878363554124, accuracy=96.7145767211914 epoch 435, loss generator = 9.05750480581213, loss discriminator=0.1 2631638975882972, accuracy=96.65581512451172 epoch 436, loss generator = 9.234907174993443, loss discriminator=0. 11981539982888434, accuracy=95.6423110961914 epoch 437, loss_generator = 9.45547728008694, loss discriminator=0.1 084246199440073, accuracy=96.67656707763672 epoch 438, loss generator = 10.605786644970928, loss discriminator= 0.10255656856353636, accuracy=96.56919860839844 epoch 439, loss generator = 10.430407467594852, loss discriminator= 0.09958719935406138, accuracy=96.66259002685547 epoch 440, loss generator = 8.49193620858369, loss discriminator=0.1 0700371011539742, accuracy=97.63046264648438 epoch 441, loss generator = 8.707506003203216, loss discriminator=0. 09962157871988085, accuracy=96.463134765625 epoch 442, loss_generator = 8.22848542884544, loss discriminator=0.0

9357521358739447, accuracy=96.66783905029297 epoch 443, loss generator = 8.06623924749869, loss discriminator=0.0 9965895658565892, accuracy=95.5280990600586 epoch 444, loss generator = 8.720835777565284, loss discriminator=0. 10143446060656397, accuracy=96.80402374267578 epoch 445, loss generator = 9.26607496296918, loss discriminator=0.0 9536481177503311, accuracy=95.65274810791016 epoch 446, loss generator = 8.831186072031658, loss discriminator=0. 08059510329255351, accuracy=97.14743041992188 epoch 447, loss generator = 8.302925233487729, loss discriminator=0. 07906765890894113, accuracy=97.15553283691406 epoch 448, loss generator = 7.351055060492621, loss discriminator=0. 0912532354739529, accuracy=96.93804931640625 epoch 449, loss generator = 7.571500262507686, loss discriminator=0. 0733762727153522, accuracy=97.4209976196289 epoch 450, loss generator = 8.232349621808087, loss discriminator=0. 07115782224055793, accuracy=96.664794921875 epoch 451, loss generator = 7.89321002430386, loss discriminator=0.0 7739289608542566, accuracy=97.15860748291016 epoch 452, loss generator = 7.676506793057477, loss discriminator=0. 08204296540330958, accuracy=97.41244506835938 epoch 453, loss generator = 7.24060208355939, loss discriminator=0.0 8605912666491888, accuracy=96.89875793457031 epoch 454, loss generator = 7.181978578920718, loss discriminator=0. 0832837748237782, accuracy=96.98011016845703 epoch 455, loss generator = 7.378570055078577, loss discriminator=0. 07867942597992994, accuracy=97.68498229980469 epoch 456, loss generator = 7.132302446718569, loss discriminator=0. 0719336797448772, accuracy=97.19715881347656 epoch 457, loss generator = 6.872718079884847, loss discriminator=0. 07589828855340817, accuracy=96.20040893554688 epoch 458, loss generator = 7.2190486166212295, loss discriminator= 0.07028240280363847, accuracy=97.955810546875 epoch 459, loss generator = 7.206185837145205, loss discriminator=0. 07257438414833611, accuracy=96.6709976196289 epoch 460, loss generator = 6.5808637901588725, loss discriminator= 0.07250777750655457, accuracy=97.5972900390625 epoch 461, loss generator = 7.257132754502473, loss discriminator=0. 07622050230080883, accuracy=97.06265258789062 epoch 462, loss generator = 7.371435115955494, loss discriminator=0. 06951218930038588, accuracy=97.70030975341797 epoch 463, loss generator = 7.22562922724971, loss discriminator=0.0 683253218350863, accuracy=97.66573333740234 epoch 464, loss generator = 6.889823698114466, loss discriminator=0. 06356216711509559, accuracy=97.81315612792969 epoch 465, loss generator = 6.860033349637632, loss discriminator=0. 058198691016339045, accuracy=97.22891235351562 epoch 466, loss generator = 6.594887260154442, loss discriminator=0. 056943534202528776, accuracy=97.06239318847656 epoch 467, loss generator = 6.656566434436375, loss discriminator=0. 05548320973498954, accuracy=97.46192932128906 epoch 468, loss generator = 5.94052049142343, loss discriminator=0.0 5712720895772455, accuracy=97.1368637084961 epoch 469, loss generator = 5.947152160715174, loss discriminator=0. 0660790933503045, accuracy=98.07032012939453

functions for presenting the results

In [21]:

```
def function result 01():
    num latent visualize = 16
    latent visualize = torch.randn(num latent visualize, dim latent, device=devi
ce)
    latent_visualize = torch.reshape(latent_visualize, [num_latent_visualize, di
m latent, 1, 1])
    prediction = compute prediction(generator, latent visualize)
    prediction = prediction.squeeze(axis=1)
    prediction_binary = (prediction >= 0.5)
    bounding mask = compute bounding mask(prediction binary)
    fake = prediction.detach().cpu()
    mask = bounding mask.detach().cpu()
    nRow = 4
    nCol = 4
    size = 3
    title = 'fake image'
    fig, axes = plt.subplots(nRow, nCol, figsize=(size * nCol, size * nRow))
    fig.suptitle(title, fontsize=16)
    for r in range(nRow):
        for c in range(nCol):
            k = c + r * nCol
            axes[r, c].imshow(fake[k], cmap='gray', vmin=0, vmax=1)
            axes[r, c].xaxis.set_visible(False)
            axes[r, c].yaxis.set visible(False)
    plt.tight layout()
    plt.show()
    title = 'bounding mask'
    fig, axes = plt.subplots(nRow, nCol, figsize=(size * nCol, size * nRow))
    fig.suptitle(title, fontsize=16)
    for r in range(nRow):
        for c in range(nCol):
            k = c + r * nCol
            axes[r, c].imshow(mask[k], cmap='gray', vmin=0, vmax=1)
            axes[r, c].xaxis.set_visible(False)
            axes[r, c].yaxis.set visible(False)
    plt.tight layout()
    plt.show()
```

In [22]:

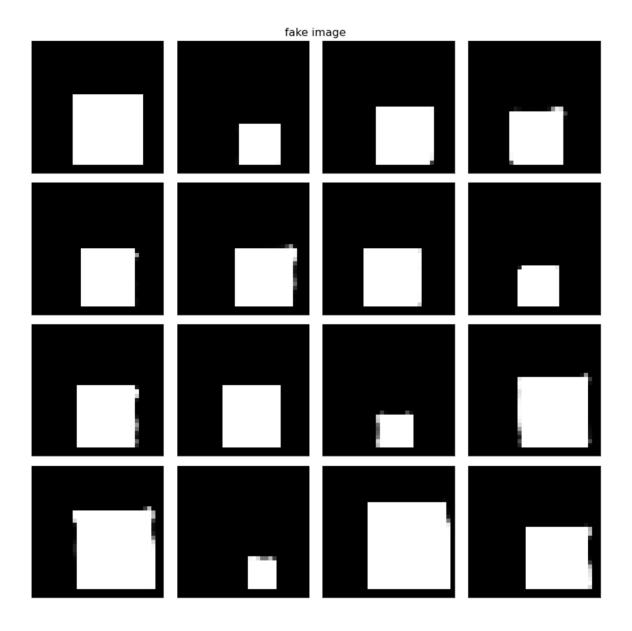
In [23]:

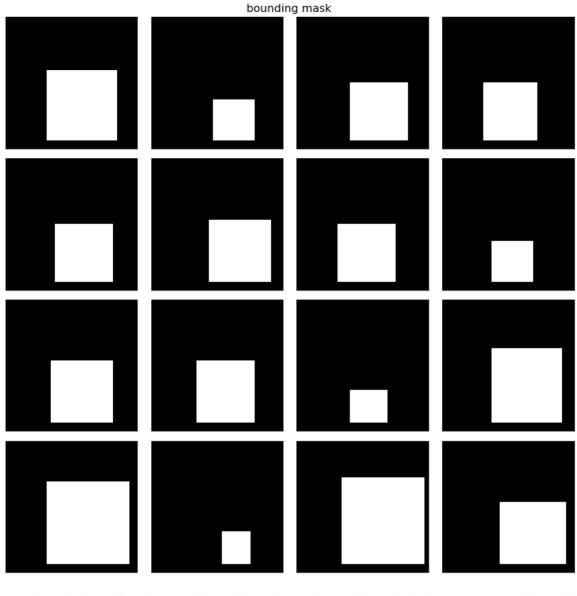
```
def function_result_03():
    print('final accuracy = %9.8f' % (accuracy_epoch[-1]))
```

results

In [24]:

```
number result = 3
for i in range(number result):
            = '# RESULT # {:02d}'.format(i+1)
  title
  name function
           = 'function result {:02d}()'.format(i+1)
  print('')
  ########" )
  print('#')
  print(title)
  print('#')
  ########" )
  print('')
  eval(name function)
```



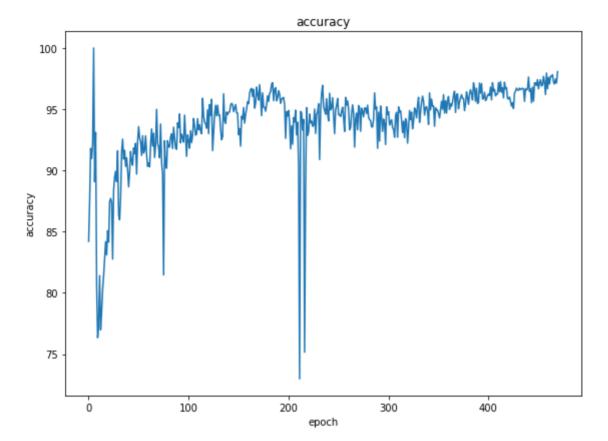


#

RESULT # 02

#

############



#

RESULT # 03

#

final accuracy = 98.07032013

In []: