Video Object Detection

July 21, 2019

1 Preamble

L'objectif de ce document est de présenter un algorithme de détermination de l'angle d'une remorque à l'arrière d'un véhicule équipé d'une caméra vidéo.

2 import

Le programme écrit en python s'appuie sur plusieurs bibliothèques existantes standard.

3 File selection

fileName = 'P473_Arizona_Day_Asphalt_Close_To_Sunset_dry_Nominal_8300lx.avi'

4 Test Video Loop

5 Variables globales

Pour des raisons de lisibilité du code, l'ensemble des variables locales seront précédés du préfixe 'l' afin de les différencier des variables globales qui n'ont pas de préfixe. ## Type de données de position

```
In [7]: posType = np.dtype([('x', 'u1'), ('y', 'u2')])
```

5.1 Type de données de Neurones

5.1.1 Neurone sensoriel à champs récepteur

5.2 Taille des champs récepteurs neuronaux

```
In [328]: tailleField = 7
```

6 Fonctions

6.1 Calcul d'un neurone champ moyen

A partir d'une liste de neurones, il retourne le neurone moyen

6.2 Matrice des directions

Afin de faciliter le calcul des angles des pixels, une matrice de poids est générée afin d'appliquer à chaque pixel centré sur un champs récepteur un poids correspondant à l'angle d'une ligne passant par ce centre. Voici comment les angles sont représentés IMAGE

6.3 Fonction d'activation des neurones

Chaque neurone retourne une valeur comprise entre 0 et 255 qui reflète son niveau d'activation. Cette activation reflète le niveau de confiance que le neurone a sur le lien existant entre sa fonction de base et les pixels reçus dans son champs récepteur. Plus les pixels sont organisés de façon à former une ligne avec l'angle correspondant à la fonction de base du neurone et plus ce dernier sera activé. Comme on ne souhaite pas obtenir une activation de valeur infinie, on utilise donc une fonction sigmoide qui s'applique à l'écart-type des angles supposés.

```
In [12]: @jit(nopython=True, parallel=True)
    def sigmoidActivationFctN1(lActivationVector):
        lDenom = (1 + np.exp(0.1 * (np.abs(np.std(lActivationVector)) - 30)))
        return 255 / lDenom
```

6.4 Création d'une liste de neurones à champs récepteurs

```
In [256]: #@jit(nopython=True, parallel=True)
          def getNeuronActivationList(idxX, idxY, size, frameE, nbPixelPts, layer=1, lVerbose=
              #commencer par créer le tableau de neurones
              lNeuronType = np.dtype([('longueur', 'u1'), ('angle', 'f4'),
                                       ('weight', 'f4'), ('precision', 'f4'),
                                       ('xPos', 'u1'), ('yPos', 'u2'), ('group', 'u1')])
              lCriterion = nbPixelPts >= size
              nbNeurons = sum(lCriterion)
              1Neurons = np.zeros(nbNeurons, dtype=1NeuronType)
              lpNeurons = pd.DataFrame(1Neurons)
              lpNeurons['longueur'] = size
              lpNeurons['layer'] = layer
              offsetField = int(np.floor(size / 2))
              lAngleMat = fillAngleMat(size)
              newX = idxX[lCriterion]
              newY = idxY[lCriterion]
              if lVerbose:
                  print("size :" + str(len(newX)))
                  print("newX")
                  print(np.min(newX))
                  print(np.max(newX))
                  print("newY")
                  print(np.min(newY))
                  print(np.max(newY))
                  print()
              pos = 0
              lnPos = 0
              for lintX in newX:
                  lintY = newY[pos]
                  if (lintX - offsetField)<0 or (lintY - offsetField)<0:</pre>
                      print("exceed the limit of the matrix")
                      pos += 1
                      continue
                  lNeuronFieldFrame = frameE[
                      int(lintX - offsetField):int(lintX + offsetField + 1),
                      int(lintY - offsetField):int(lintY + offsetField + 1)] / 255
                  try:
                      tmp = np.multiply(lAngleMat, lNeuronFieldFrame)
                  except:
                      print("error 10 : ")
```

```
print("lAngleMat")
        print(lAngleMat)
        print("lNeuronFieldFrame")
        print(lNeuronFieldFrame)
        print("lintX")
        print(lintX)
        print("lintY")
        print(lintY)
        print("offsetField")
        print(offsetField)
        continue
    lNeuronFieldValues = tmp[np.nonzero(tmp)]
    if lVerbose:
        print("lNeuronFieldFrame :")
        print(lNeuronFieldFrame)
        print("np.multiply(lAngleMat, lNeuronFieldFrame)")
        print(tmp)
        print("lNeuronFieldValues")
        print(lNeuronFieldValues)
    if (np.mean(lNeuronFieldValues))<0:</pre>
        1NeuronFieldValues[1NeuronFieldValues>89]=-90
    elif np.std(lNeuronFieldValues)>45:
        1NeuronFieldValues[1NeuronFieldValues>89]=-90
    if (lNeuronFieldValues.size > 0):
        lpNeurons.loc[pos, ['angle']] = np.mean(lNeuronFieldValues)
        lpNeurons.loc[pos, ['weight']] = sigmoidActivationFctN1(
            lNeuronFieldValues)
        lpNeurons.loc[pos, ['precision']] = np.std(lNeuronFieldValues)
        lpNeurons.loc[pos, ['xPos']] = lintX
        lpNeurons.loc[pos, ['yPos']] = lintY
        lnPos += 1
    else:
        True #print ("error it shouldn't be zero")
    pos += 1
if lVerbose:
    print("nb de positions couvertes : " + str(lnPos) + " sur " + str(pos))
return lpNeurons
```

6.5 Nombre de pixels actifs dans chaque champs récepteur

A partir des coordonnées des centres supposés de chaque champs récepteurs et de la taille du champs récepteur, recherche sur la frame bitmap passée en paramètres, retourne un tableau contenant le nombre de pixels allumés à l'intérieur de chacun de ces champs.

```
In [14]: @jit(nopython=True, parallel=True)
         def nbPixelField(lTableX, lTableY, lFrameEdge, lintTailleField=3):
             1Idx = 0
             lResults = np.zeros(lTableX.size)
             lRayon = np.floor(lintTailleField / 2)
             lTailleMaxX = lFrameEdge.shape[0]
             \#lTailleMaxY = lFrameEdge.shape[1]
             lHalfX = lTailleMaxX / 3
             for lPosX in lTableX:
                 lPosY = lTableY[lIdx]
                 if lPosX > lHalfX and lPosX >= lRayon and (lPosX +
                                                             lRayon) < lTailleMaxX:</pre>
                     lResults[lIdx] = np.sum(
                         1FrameEdge[int(1PosX - 1Rayon):int(1PosX + 1Rayon + 1),
                                    int(lPosY - lRayon):int(lPosY + lRayon + 1)] / 255)
                 1Idx += 1
             return lResults
In [15]: #@jit(nopython=True, parallel=True)
         def getNonZero(LImg):
             return np.where(LImg != [0])
6.6 Coordonnées de la fonction de base (ligne)
In [101]: def getNFCoordinate(lNeurone,lVerbose=False):
              try:
                  lintDist = int(np.floor(lNeurone.longueur / 2))
              except:
                  1P1 = (0, 0)
                  1P2 = (0, 0)
                  return (1P1, 1P2)
              if np.abs(lNeurone.angle) < 45:
                  lAlpha = lNeurone.angle / 180 * pi
                  lintY1 = np.around(lNeurone.yPos - lintDist * np.tan(lAlpha))
                  lintX1 = lNeurone.xPos + lintDist
                  lintY2 = np.around(lNeurone.yPos + lintDist * np.tan(lAlpha))
                  lintX2 = lNeurone.xPos - lintDist
              else:
                  lAlpha = (90 - lNeurone.angle) / 180 * pi
                  if lVerbose:
                      print("Angle : "+str(lNeurone.angle))
                      print("yPos = "+str(lNeurone.yPos)+"xPos = "+str(lNeurone.xPos))
                  lintX1 = np.around(lNeurone.xPos - lintDist * np.tan(lAlpha))
                  lintY1 = lNeurone.yPos + lintDist
                  lintX2 = np.around(lNeurone.xPos + lintDist * np.tan(lAlpha))
                  lintY2 = lNeurone.yPos - lintDist
              1P1 = (int(lintY1), int(lintX1))
```

```
1P2 = (int(lintY2), int(lintX2))
if lVerbose:
    print("point 1: "+str(lP1))
    print("point 2: "+str(lP2))
    print("")
return lP1, lP2
```

6.7 Calcule la distance entre deux points

6.8 Retourne les neurones les plus proches d'un point

6.9 Crée un neurone avec les paramètres passés

6.10 Dessine les fonctions de base des neurones sur un bitmap

```
if lVerbose:
    print(lNeuron)
    print(lCoord)

if lIndexPassOver > 7:
    lIndexPassOver = lInitShow
    try:
        cv2.line(lBitmap, lCoord[0], lCoord[1],
        # (255, 255, 255), 1)
        (int(lNeuron.weight), int(lNeuron.weight), int(lNeuron.weight)), 1)
    except:
        True

if lVerbose:
    lIndexPassOver += 1

return lBitmap
```

6.11 Find neuronal groups

Un groupe neuronal est un ensemble de neurone dont les champs récepteurs sont complémentaires les uns des autres. Pour faire partie d'un champs récepteur, deux conditions doivent être réunies. (A compléter) ### Translation Retourne les coordonnées d'un point translaté d'une certaine distance avec un certain angle. Cette fonction demande un angle, une distance et les coordonnées d'un point de départ. Il retourne ensuite les coordonnées après translation.

Effectue le même calcul que la fonction moveCoordDeg mais prend comme paramètre un neurone. Il effectue la translation en prenant comme point de départ le centre du champs récepteur et effectue un déplacement de la taille de ce champs dans la direction de la fonction de base.

6.11.1 Calcul des groupes à partir d'une liste de neurones à champs récepteurs

```
lintCurrentGroupID = 0
lintNbGroups = 0
1Index = 0
##DEBUG
lnbNeuron = 0
##DEBUG
# liste des neurones sans groupe
lNoGroupList = lneuronList[lneuronList.group == 0]
while lNoGroupList.shape[0] > 0:
    #Sélection d'un neurone dans la liste (ceux sans groupID ou groupID=0)
    lMoyenNeuron = lNoGroupList.iloc[0]
    lIndex = lNoGroupList.head().index.values[0]
    while True:
        #Assignation d'un nouveau numéro de GroupID en cours
        lintNbGroups += 1
        lintCurrentGroupID += 1
        if lneuronList[lneuronList.group ==
                       lintCurrentGroupID].shape[0] == 0:
            break
    lneuronList.loc[lIndex, ['group']] = lintCurrentGroupID
    #déplacement
    lnPos = getNextPosition(lMoyenNeuron, lVerbose)
    #recherche de neurones proches
    lClosestNeurons = closestFieldNeurons(
        lneuronList, lnPos[0], lnPos[1],
        int(np.floor(lMoyenNeuron.longueur / 2)))
    if lVerbose:
        print("")
        print("")
        print("Coordonnées en cours : (" + str(lnPos[0]) + "," +
              str(lnPos[1]) + ")")
        lnbNeuron += 1
        if lClosestNeurons.shape[0] == 0:
            print("Aucun neurone a proximité pour le neurone #" +
                  str(lnbNeuron) + " aux coordonnées : (" + str(lnPos[0]) +
                  "," + str(lnPos[1]) + str(") a la distance :") +
                  str(int(np.floor(lMoyenNeuron.longueur / 2))))
        #Oui ==> retour étape 1
        1NbFindGroup = 0
```

```
while lClosestNeurons.shape[0] != 0:
    #recherche des groupID dans cette sous-sélection
    if lClosestNeurons[lClosestNeurons.group > 0].shape[0] == 0:
        #Non => Assigner à tous les neurones de la sous-sélection
        #le groupID en cours => aller directement à l'étape 7
        if lVerbose:
            print("Aucun neurone dans le groupe : " +
                  str(lintCurrentGroupID))
        for lintIdx in lClosestNeurons.head().index.values:
            lneuronList.loc[lintIdx, ['group']] = lintCurrentGroupID
    else:
        #Oui
        if lVerbose:
            ##DEBUG
            #lNbFindGroup += 1
            print("Trouvé " + str(lClosestNeurons[
                lClosestNeurons.group > 0].shape[0]) +
                  " neurone(s) déja dans des groupes :")
            print("Groupe en cours : " + str(lintCurrentGroupID))
        #Récupération de la liste de tous les groupID utilisés
        #Sélection du groupID le plus petit
        #(en comparant aussi avec le groupID en cours)
        lintPreviousGroupID = lintCurrentGroupID
        lintCurrentGroupID = np.min(
            lClosestNeurons[lClosestNeurons.group > 0].group)
        if lVerbose:
            print("Change pour le groupe #" + str(lintCurrentGroupID))
            print("-")
        #Assigner au neurone en cours le nouveau groupe
        lneuronList.loc[lIndex, ['group']] = lintCurrentGroupID
        #Assigner à tous les neurones de la sous-sélection ce nouveau groupID
        for lintIdx in lClosestNeurons.head().index.values:
            lneuronList.loc[lintIdx, ['group']] = lintCurrentGroupID
            #remplacer dans la liste globale,
            #pour chaque groupID présent dans la liste par le nouveau groupID
            for lintGroupID in lClosestNeurons[
                    lClosestNeurons.group > 0].group:
                lneuronList.loc[lneuronList.group == lintGroupID,
                                'group'] = lintCurrentGroupID
        if lintPreviousGroupID == lintCurrentGroupID:
            #si tous les neurones
            if lClosestNeurons[lClosestNeurons.group >
                               0].shape[0] == lClosestNeurons[
                                   lClosestNeurons.group ==
                                   lintPreviousGroupID].shape[0]:
```

```
if lVerbose:
                         #Calcul du neurone Field moyen
                         print("Neurones trouvé :")
                         print(lClosestNeurons)
                     lMoyenNeuron = getAvgFieldNeuron(lClosestNeurons)
                     if lVerbose:
                         print("neurone Moyen")
                         print(lMoyenNeuron)
                     #déplacement
                     lnPos = getNextPosition(lMoyenNeuron, lVerbose)
                     #recherche de neurones proches
                     lClosestNeurons = closestFieldNeurons(
                         lneuronList, lnPos[0], lnPos[1],
                         int(np.floor(lMoyenNeuron.longueur / 2)))
                 lNoGroupList = lneuronList[lneuronList.group == 0]
             return lneuronList
6.12 Get the Weighted Average of the group Angle (WAGA)
In [424]: def getWAGA(lNeuronList, lGroupID):
              lnl = lNeuronList[lNeuronList.group==lGroupID]
              return float(np.sum((lnl.angle * lnl.weight) /
                         np.sum(lnl.weight)))
6.13 Get two Neuron line spread
In [ ]: def getTNLSpread(lNeuron1, lNeuron2):
            return 0;
   Video Loop
In [24]: kernelSize = 21 # Kernel Bluring size
         # Edge Detection Parameter
         parameter1 = 20
         parameter2 = 40
         intApertureSize = 1
         \#cap = cv2.VideoCapture(0)
         cap = cv2.VideoCapture(fileName)
         1Counter = 0
         while (cap.isOpened()):
             # Capture frame-by-frame
             ret, Cannyframe = cap.read()
```

break # sortie de la boucle while

```
if ret == True:
        # Our operations on the frame come here
        if lCounter == 1:
            Cannyframe = cv2.GaussianBlur(Cannyframe, (kernelSize, kernelSize), 0, 0)
            Cannyframe = cv2.Canny(Cannyframe, parameter1, parameter2,
                              intApertureSize) # Canny edge detection
            1Counter = 0
            break
        lCounter += 1
        # Display the resulting frame
        cv2.imshow('Edges Video', Cannyframe)
        if cv2.waitKey(1) & OxFF == ord('q'): # press q to quit
            break
    else:
        break
# When everything done, release the capture
cap.release()
cv2.destroyAllWindows()
```

8 Sandbox

8.1 Toy data Generator

```
In [339]: def generateToy(lType=1, lHauteur=80, lLargeur=128,lepaisseur=1):
              lFrame = 0
              if lType == 1:
                  1Frame = np.zeros((lHauteur, lLargeur))
                  1Frame[:, int((lLargeur-lepaisseur) / 2):int((lLargeur+lepaisseur) / 2)] = 2
              elif lType == 2:
                  1Frame = np.zeros((lHauteur, lLargeur))
                  1Frame[int((lHauteur-lepaisseur) / 2):int((lHauteur+lepaisseur)/2), :] = 255
              elif lType == 3:
                  lFrame = np.zeros((lHauteur, lLargeur))
                  cv2.line(lFrame, (int(lLargeur / 3), lHauteur),
                           (int(2 * 1Largeur / 3), 0), (255, 255, 255), lepaisseur)
              elif lType == 4:
                  1Frame = np.zeros((lHauteur, lLargeur))
                  cv2.rectangle(1Frame,
                                 (int(lLargeur / 128 * 10), int(lHauteur / 80 * 30)),
                                 (int(lLargeur / 128 * 30), int(lHauteur / 80 * 50)),
                                 (255, 255, 255), lepaisseur)
                  pts = np.array([[int(lLargeur / 128 *64),
                                   int(lHauteur / 80 * 30)],
                                  [int(lLargeur / 128 * 76),
                                   int(lHauteur / 80 * 50)],
                                   [int(lLargeur / 128 * 53),
```

```
int(lHauteur / 80 *50)]], np.int32)
   ts = pts.reshape((-1, 1, 2))
   cv2.polylines(lFrame, [pts], True, (255, 255, 255), lepaisseur)
    cv2.circle(1Frame,
               (int(lLargeur / 128 * 107), int(lHauteur / 80 * 40)),
               int(lHauteur / 80 * 10), (255, 255, 255), lepaisseur)
elif lType == 5:
   1Frame = np.zeros((lHauteur, lLargeur))
    #createNeuron(llong, langle, lXpos, lYpos,
   llN = int(lLargeur / 256 * 50)
    llXTmp = int(lHauteur / 160 * 100)
   lNeuronTest2 = createNeuron(llN, -75, llXTmp, int(lLargeur / 256 * 25))
    drawFieldNeurons(1NeuronTest2, 1Frame)
    lNeuronTest2 = createNeuron(llN, -60, llXTmp, int(lLargeur / 256 * 75))
   drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(11N, 0, 11XTmp, int(1Largeur / 256 * 125))
   drawFieldNeurons(1NeuronTest2, 1Frame)
   lNeuronTest2 = createNeuron(llN, 60, llXTmp, int(lLargeur / 256 * 175))
   drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(11N, 75, 11XTmp, int(1Largeur / 256 * 225))
    drawFieldNeurons(lNeuronTest2, lFrame)
elif lType == 6:
   1Frame = np.zeros((lHauteur, lLargeur))
    #createNeuron(llong, langle, lXpos, lYpos,
    11N = int(lLargeur / 256 * 50)
    llXTmp = int(lHauteur / 160 * 100)
   lNeuronTest2 = createNeuron(llN, 0, llXTmp, int(lLargeur / 256 * 25))
    drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(llN, 0, llXTmp, int(lLargeur / 256 * 75))
   drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(11N, 0, 11XTmp, int(1Largeur / 256 * 125))
    drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(11N, 0, 11XTmp, int(1Largeur / 256 * 175))
   drawFieldNeurons(1NeuronTest2, 1Frame)
   1NeuronTest2 = createNeuron(11N, 0, 11XTmp, int(1Largeur / 256 * 225))
    drawFieldNeurons(lNeuronTest2, lFrame)
elif lType == 7:
   1Frame = np.zeros((lHauteur, lLargeur))
    #createNeuron(llong, langle, lXpos, lYpos,
   llN = int(lLargeur / 256 * 50)
   llXTmp = int(lHauteur / 160 * 100)
    lNeuronTest2 = createNeuron(11N, 5, 11XTmp, int(1Largeur / 256 * 25))
    drawFieldNeurons(lNeuronTest2, lFrame)
   lNeuronTest2 = createNeuron(llN, 5, llXTmp, int(lLargeur / 256 * 75))
   drawFieldNeurons(1NeuronTest2, 1Frame)
    lNeuronTest2 = createNeuron(11N, 0, 11XTmp+2, int(1Largeur / 256 * 125))
    drawFieldNeurons(lNeuronTest2, lFrame)
    lNeuronTest2 = createNeuron(llN, 0, llXTmp+2, int(lLargeur / 256 * 175))
```

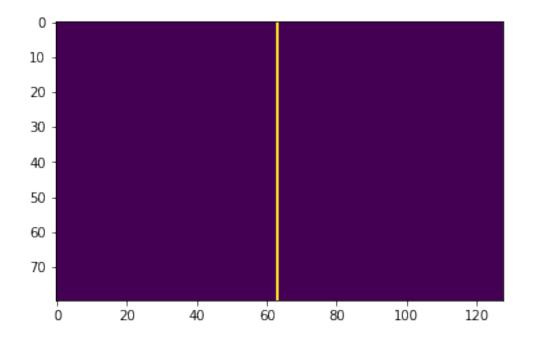
```
drawFieldNeurons(lNeuronTest2, lFrame)
    lNeuronTest2 = createNeuron(llN, 0, llXTmp+2, int(lLargeur / 256 * 225))
    drawFieldNeurons(lNeuronTest2, lFrame)
else:
    lFrame = np.zeros((lHauteur, lLargeur))
    print("First parameter should be between 1 to 7")
return lFrame
```

8.2 Playground

8.2.1 Test 1

Generate data of type 1

```
In [26]: frame = generateToy(1,80,128,1)
        imgplot = plt.imshow(frame)
```



Génération des neurones à champs récepteur

In [28]: titi.describe()

```
Out [28]:
                                                                                 group \
                longueur
                                  angle
                                             weight
                                                     precision
                                                                    xPos
                                                                           yPos
                    50.0
                          5.000000e+01
                                          50.000000
                                                          50.0
                                                                50.00000
                                                                           50.0
                                                                                  50.0
         count
                     7.0
                         9.999996e-03
                                         242.906311
                                                           0.0
                                                                51.50000
                                                                           63.0
                                                                                   0.0
         mean
         std
                     0.0 3.763112e-09
                                           0.000092
                                                           0.0
                                                                14.57738
                                                                            0.0
                                                                                   0.0
                     7.0 1.000000e-02
                                         242.906403
                                                           0.0 27.00000
                                                                           63.0
                                                                                   0.0
         min
         25%
                     7.0 1.000000e-02
                                         242.906403
                                                           0.0 39.25000
                                                                           63.0
                                                                                   0.0
         50%
                     7.0 1.000000e-02
                                         242.906403
                                                           0.0 51.50000
                                                                           63.0
                                                                                   0.0
                                                                63.75000
                          1.000000e-02
         75%
                     7.0
                                         242.906403
                                                           0.0
                                                                           63.0
                                                                                   0.0
                     7.0
                         1.000000e-02
                                         242.906403
                                                           0.0 76.00000
                                                                           63.0
                                                                                   0.0
         max
                layer
                 50.0
         count
                  1.0
         mean
         std
                  0.0
         min
                  1.0
         25%
                  1.0
         50%
                  1.0
         75%
                  1.0
                  1.0
         max
In [29]: titi[0:4]
Out [29]:
            longueur
                      angle
                                  weight precision xPos
                                                           yPos
                                                                 group
                                                                         layer
                       0.01 242.906403
                                                0.0
         0
                   7
                                                       27
                                                             63
                                                                      0
                                                                             1
         1
                   7
                       0.01
                             242.906403
                                                0.0
                                                       28
                                                             63
                                                                      0
                                                                             1
```

0.0

0.0

29

30

63

63

0

0

1

1

Affichage graphique du champs récepteur des neurones

0.01

0.01

7

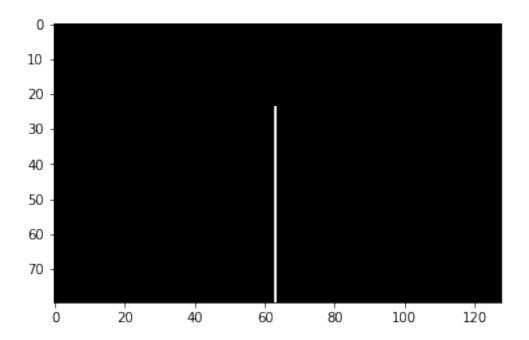
7

2

3

242.906403

242.906403

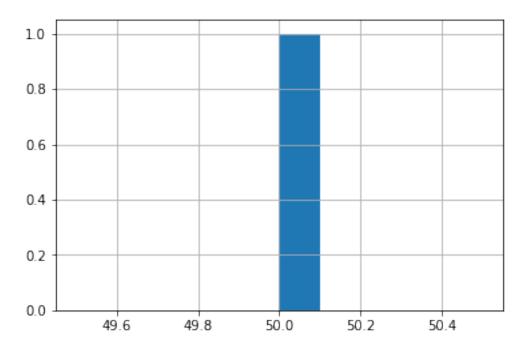


Génération des groupes

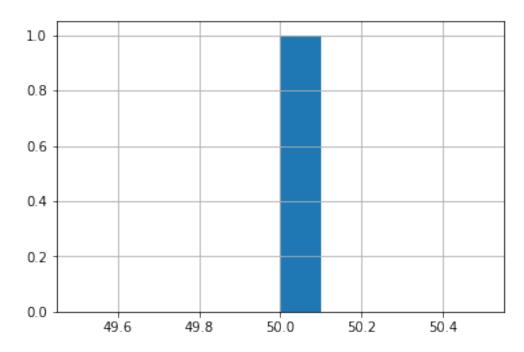
```
In [33]: findGroups(titi);
In [34]: titi.groupby('group').agg(['mean', 'count'])[0:5]
Out[34]:
               longueur
                              angle
                                               weight
                                                            precision
                                                                              xPos \
                   mean count mean count
                                                 mean count
                                                                  mean count
                                                                              mean
         group
                      7
         1
                           50 0.01
                                       50 242.906403
                                                          50
                                                                   0.0
                                                                          50 51.5
                     yPos
                                layer
               count mean count mean count
         group
         1
                  50
                       63
                             50
                                    1
                                         50
```

In [35]: titi.groupby('group').size().hist()

Out[35]: <matplotlib.axes._subplots.AxesSubplot at 0x12e2c5ba8>



Out[36]: <matplotlib.axes._subplots.AxesSubplot at 0x12de5aa90>



In [37]: titi.describe()

Out[37]:		longueur	angle	weight	precision	xPos	yPos	group	\
	count	50.0	5.000000e+01	50.000000	50.0	50.00000	50.0	50.0	
	mean	7.0	9.999996e-03	242.906311	0.0	51.50000	63.0	1.0	
	std	0.0	3.763112e-09	0.000092	0.0	14.57738	0.0	0.0	
	min	7.0	1.000000e-02	242.906403	0.0	27.00000	63.0	1.0	
	25%	7.0	1.000000e-02	242.906403	0.0	39.25000	63.0	1.0	
	50%	7.0	1.000000e-02	242.906403	0.0	51.50000	63.0	1.0	
	75%	7.0	1.000000e-02	242.906403	0.0	63.75000	63.0	1.0	
	max	7.0	1.000000e-02	242.906403	0.0	76.00000	63.0	1.0	

	layer
count	50.0
mean	1.0
std	0.0
min	1.0
25%	1.0
50%	1.0
75%	1.0
max	1.0

In [38]: titi[0:4]

Out[38]:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	7	0.01	242.906403	0.0	27	63	1	1
1	7	0.01	242.906403	0.0	28	63	1	1
2	7	0.01	242.906403	0.0	29	63	1	1
3	7	0.01	242.906403	0.0	30	63	1	1

Simplify

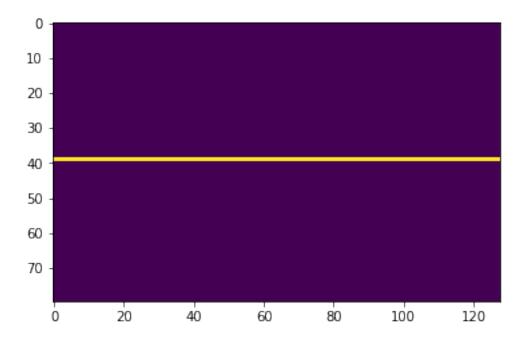
In [455]: np.sum(titi.memory_usage())

Out[455]: 5020

8.2.2 Test 2

Generate data of type 2

```
In [39]: frame = generateToy(2,80,128,1)
    imgplot = plt.imshow(frame)
```



Génération des neurones à champs récepteur

```
In [40]: indices = np.where(frame != [0])
        nbPixelsAll = nbPixelField(indices[0], indices[1], frame, tailleField)
        titi = getNeuronActivationList(indices[0], indices[1], tailleField, frame,
                                       nbPixelsAll)
exceed the limit of the matrix
exceed the limit of the matrix
exceed the limit of the matrix
error 10 :
lAngleMat
[[-4.499e+01 -3.368e+01 -1.842e+01  1.000e-02  1.844e+01  3.370e+01]
  4.501e+01]
 [-5.630e+01 -4.499e+01 -2.656e+01 1.000e-02 2.658e+01 4.501e+01
  5.632e+01]
 [-7.156e+01 -6.342e+01 -4.499e+01 1.000e-02 4.501e+01 6.344e+01
  7.158e+01]
 [ 9.000e+01 9.000e+01 9.000e+01 0.000e+00 9.000e+01 9.000e+01
  9.000e+01]
 [ 7.158e+01 6.344e+01 4.501e+01 1.000e-02 -4.499e+01 -6.342e+01
 -7.156e+01]
 [ 5.632e+01 4.501e+01 2.658e+01 1.000e-02 -2.656e+01 -4.499e+01
 -5.630e+01]
 [ 4.501e+01 3.370e+01 1.844e+01 1.000e-02 -1.842e+01 -3.368e+01
 -4.499e+01]]
```

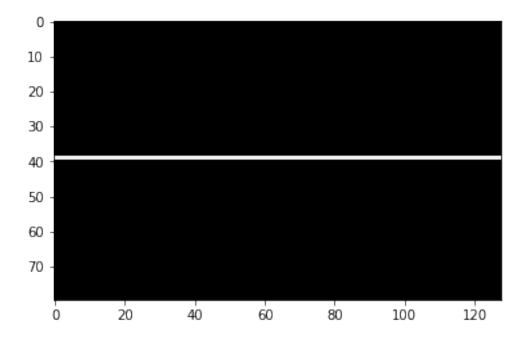
```
lNeuronFieldFrame
[[0. 0. 0. 0. 0. 0.]
[0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [1. 1. 1. 1. 1. 1.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]]
lintX
39
lintY
125
offsetField
3
error 10 :
lAngleMat
[[-4.499e+01 -3.368e+01 -1.842e+01 1.000e-02 1.844e+01 3.370e+01
   4.501e+01]
 [-5.630e+01 -4.499e+01 -2.656e+01 1.000e-02 2.658e+01 4.501e+01
   5.632e+01]
 [-7.156e+01 -6.342e+01 -4.499e+01 1.000e-02 4.501e+01 6.344e+01
   7.158e+01]
 [ 9.000e+01 9.000e+01 9.000e+01 0.000e+00 9.000e+01 9.000e+01
   9.000e+01]
 [ 7.158e+01 6.344e+01 4.501e+01 1.000e-02 -4.499e+01 -6.342e+01
 -7.156e+01]
 [ 5.632e+01 4.501e+01 2.658e+01 1.000e-02 -2.656e+01 -4.499e+01
 -5.630e+01]
 [ 4.501e+01 3.370e+01 1.844e+01 1.000e-02 -1.842e+01 -3.368e+01
 -4.499e+01]]
lNeuronFieldFrame
[[0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [1. 1. 1. 1. 1. 1.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]]
lintX
39
lintY
125
offsetField
3
error 10 :
lAngleMat
 \begin{bmatrix} [-4.499e+01 & -3.368e+01 & -1.842e+01 & 1.000e-02 & 1.844e+01 & 3.370e+01 \end{bmatrix} 
   4.501e+01]
```

```
[-5.630e+01 -4.499e+01 -2.656e+01 1.000e-02 2.658e+01 4.501e+01
   5.632e+01]
 [-7.156e+01 -6.342e+01 -4.499e+01 1.000e-02 4.501e+01 6.344e+01
   7.158e+01]
 [ 9.000e+01 9.000e+01 9.000e+01 0.000e+00 9.000e+01 9.000e+01
   9.000e+01]
 [ 7.158e+01 6.344e+01 4.501e+01 1.000e-02 -4.499e+01 -6.342e+01
 -7.156e+01]
 [ 5.632e+01 4.501e+01 2.658e+01 1.000e-02 -2.656e+01 -4.499e+01
 -5.630e+017
 [ 4.501e+01 3.370e+01 1.844e+01 1.000e-02 -1.842e+01 -3.368e+01
  -4.499e+01]]
lNeuronFieldFrame
[[0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [1. 1. 1. 1. 1. 1.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0.]]
lintX
39
lintY
125
offsetField
3
In [41]: titi.describe()
Out [41]:
                longueur
                                          weight precision
                                                                               yPos \
                               angle
                                                                   xPos
                   128.0 128.000000 128.000000
                                                      128.0 128.000000
                                                                         128.000000
         count
        mean
                     7.0
                           85.781250
                                      231.520050
                                                        0.0
                                                              37.171875
                                                                          60.523438
                                       51.545200
         std
                     0.0
                           19.098122
                                                        0.0
                                                               8.275863
                                                                          37.053901
        min
                     7.0
                           0.000000
                                        0.000000
                                                        0.0
                                                               0.000000
                                                                           0.000000
        25%
                     7.0
                           90.000000 242.906403
                                                        0.0
                                                              39.000000
                                                                          28.750000
                     7.0
                                                        0.0
        50%
                           90.000000 242.906403
                                                              39.000000
                                                                          60.500000
         75%
                     7.0
                           90.000000 242.906403
                                                        0.0
                                                              39.000000
                                                                          92.250000
                     7.0
                           90.000000 242.906403
                                                        0.0
                                                              39.000000 124.000000
        max
                group
                      layer
               128.0
                      128.0
         count
                  0.0
                         1.0
        mean
         std
                  0.0
                         0.0
                         1.0
        min
                  0.0
         25%
                  0.0
                         1.0
        50%
                  0.0
                         1.0
        75%
                  0.0
                         1.0
                  0.0
                         1.0
        max
```

In [42]: titi[0:4]

Out[42]:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	7	0.0	0.000000	0.0	0	0	0	1
1	7	0.0	0.000000	0.0	0	0	0	1
2	7	0.0	0.000000	0.0	0	0	0	1
3	7	90.0	242.906403	0.0	39	3	0	1

Affichage graphique du champs récepteur des neurones



Génération des groupes

```
In [44]: findGroups(titi);
```

In [45]: titi.groupby('group').agg(['mean', 'count'])[0:5]

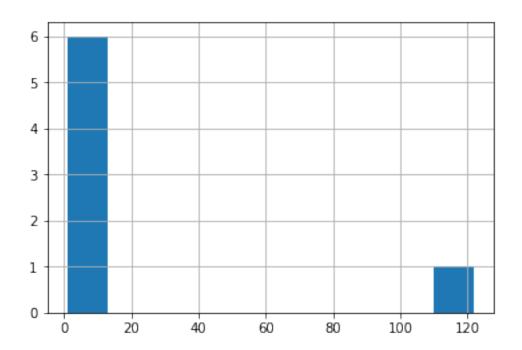
Out[45]:		longueur		angle		weight		precision		xPos		١
		mean	count	mean	count	mean	count	mean	${\tt count}$	${\tt mean}$	count	
	group											
	1	7	1	0.0	1	0.000000	1	0.0	1	0	1	
	2	7	1	0.0	1	0.000000	1	0.0	1	0	1	
	3	7	1	0.0	1	0.000000	1	0.0	1	0	1	
	4	7	122	90.0	122	242.906403	122	0.0	122	39	122	

5 7 1 0.0 1 0.000000 1 0.0 1 0 1

	yPos		layer	
	mean	count	mean	${\tt count}$
group				
1	0.0	1	1	1
2	0.0	1	1	1
3	0.0	1	1	1
4	63.5	122	1	122
5	0.0	1	1	1

In [46]: titi.groupby('group').size().hist()

Out[46]: <matplotlib.axes._subplots.AxesSubplot at 0x12dd8cb38>

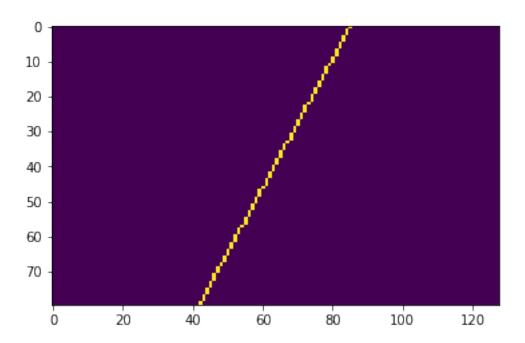


In [47]: titi[0:4]

Out $[47]$:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	7	0.0	0.000000	0.0	0	0	1	1
1	7	0.0	0.000000	0.0	0	0	2	1
2	7	0.0	0.000000	0.0	0	0	3	1
3	7	90.0	242.906403	0.0	39	3	4	1

8.2.3 Test 3

Generate data of type 3



Génération des neurones à champs récepteur

In [50]: titi.describe()

Out[50]:	longueur	angle	weight	precision	xPos	yPos	group	\
count	50.0	50.000000	50.000000	50.000000	50.00000	50.000000	50.0	
mean	7.0	26.755207	213.043991	13.634507	51.50000	56.980000	0.0	
std	0.0	2.806284	5.832972	1.934354	14.57738	7.924362	0.0	
min	7.0	25.053333	205.570328	7.588554	27.00000	44.000000	0.0	
25%	7.0	25.053333	212.746216	13.835940	39.25000	50.250000	0.0	
50%	7.0	25.053333	212.746216	13.835940	51.50000	57.000000	0.0	
75%	7.0	27.596666	212.746216	13.835940	63.75000	63.750000	0.0	
max	7.0	35.096668	230.490402	15.747627	76.00000	70.000000	0.0	

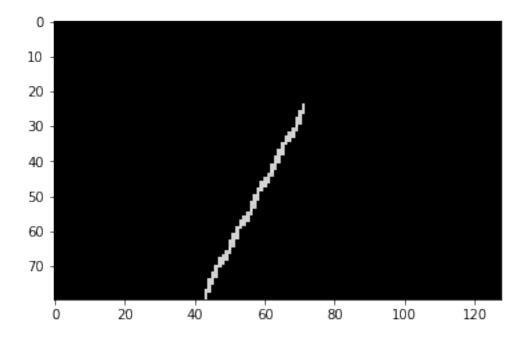
	layer
count	50.0
mean	1.0
std	0.0

```
min 1.0
25% 1.0
50% 1.0
75% 1.0
max 1.0
```

In [51]: titi[0:4]

Out[51]:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	7	25.053333	212.746216	13.83594	27	70	0	1
1	7	25.053333	212.746216	13.83594	28	70	0	1
2	7	25.053333	212.746216	13.83594	29	69	0	1
3	7	25.053333	212.746216	13.83594	30	69	0	1

Affichage graphique du champs récepteur des neurones



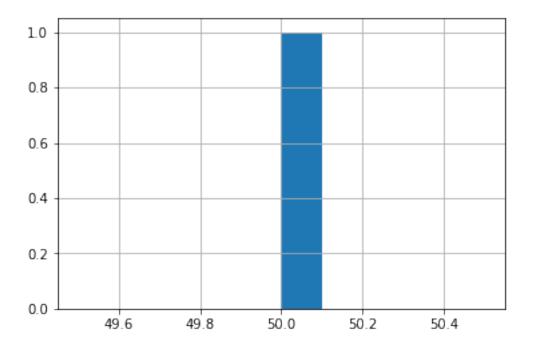
Génération des groupes

```
In [53]: findGroups(titi);
In [54]: titi.groupby('group').agg(['mean', 'count'])[0:5]
```

```
Out[54]:
               longueur
                                    angle
                                                     weight
                                                                   precision
                   mean count
                                    mean count
                                                       mean count
                                                                         mean count
         group
         1
                      7
                           50 26.755199
                                             50
                                                 213.044052
                                                               50
                                                                   13.634507
                                                                                 50
                xPos
                             yPos
                                         layer
                                         mean count
                mean count
                             mean count
         group
                51.5
                        50 56.98
                                      50
                                                  50
```

In [55]: titi.groupby('group').size().hist()

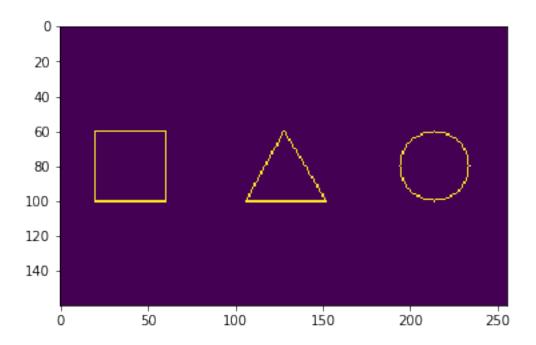
Out[55]: <matplotlib.axes._subplots.AxesSubplot at 0x12df24b38>



8.2.4 Test 4

Generate data of type 4

```
In [372]: frame = generateToy(4,160,256,1)
    imgplot = plt.imshow(frame)
```



Génération des neurones à champs récepteur

In [374]: titi.describe()

Out[374]:	longueur	angle	weight	precision	xPos	yPos	\
count	398.0	398.000000	398.000000	398.000000	398.000000	398.000000	
mean	7.0	22.983759	214.483398	9.913938	82.311558	117.037688	
std	0.0	52.803646	44.123013	11.387993	15.140764	72.832620	
min	7.0	-86.926666	10.970448	0.000000	60.000000	20.000000	
25%	7.0	-7.785227	205.570328	0.000000	67.000000	50.000000	
50%	7.0	13.627500	230.490402	7.588554	84.000000	120.000000	
75%	7.0	90.000000	242.906403	15.747627	99.000000	195.000000	
max	7.0	90.000000	242.906403	61.020844	100.000000	234.000000	

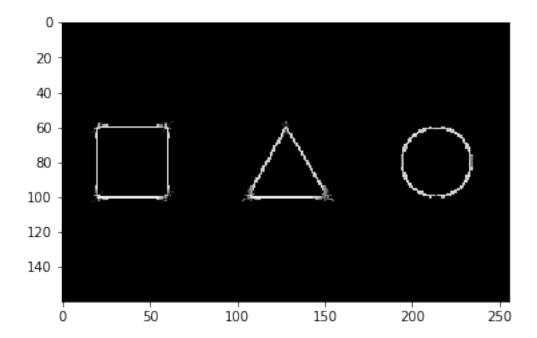
	group	layer
count	398.0	398.0
mean	0.0	1.0
std	0.0	0.0
min	0.0	1.0
25%	0.0	1.0
50%	0.0	1.0

```
75% 0.0 1.0 max 0.0 1.0
```

In [375]: titi[0:4]

Out[375]:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	7	45.005001	46.537529	44.994999	60	20	0	1
1	7	64.290001	123.892136	30.566092	60	21	0	1
2	7	74.018753	176.191147	21.954557	60	22	0	1
3	7	79.212219	202.501602	16.500349	60	23	0	1

Affichage graphique du champs récepteur des neurones

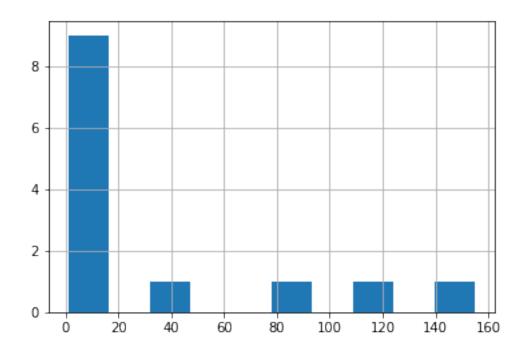


Génération des groupes

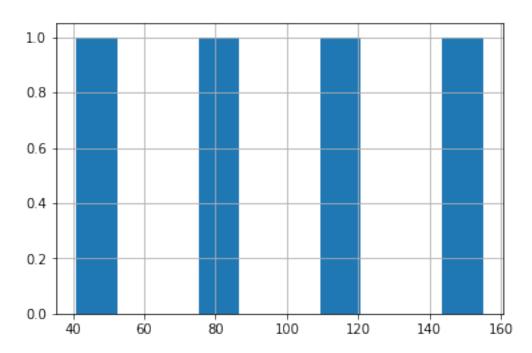
In [377]: findGroups(titi);

In [378]: titi.groupby('group').size().hist()

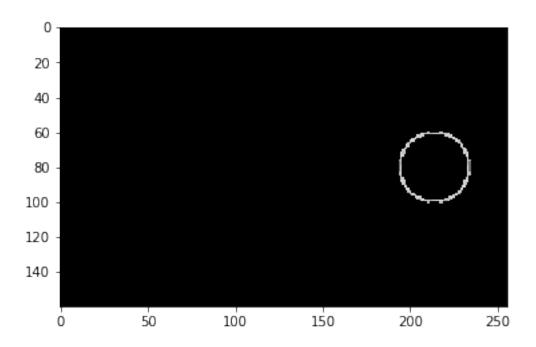
Out[378]: <matplotlib.axes._subplots.AxesSubplot at 0x134986a58>



Out[379]: <matplotlib.axes._subplots.AxesSubplot at 0x132711da0>

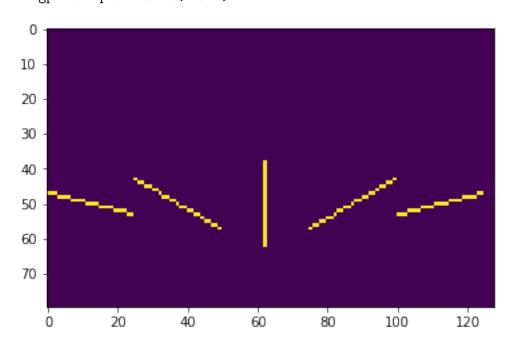


```
In [396]: titi.groupby('group').agg(['mean', 'count'])[resultGroup>10]
Out [396]:
                longueur
                                     angle
                                                       weight
                                                                      precision
                    mean count
                                      mean count
                                                         mean count
                                                                           mean count
          group
                                                                       3.791248
          1
                        7
                            155
                                40.768021
                                              155
                                                   229.658386
                                                                155
                                                                                  155
          6
                        7
                             83
                                 48.416748
                                              83 211.846710
                                                                 83
                                                                      10.593689
                                                                                   83
          9
                        7
                             41 -21.370747
                                               41 187.712830
                                                                      18.333818
                                                                                   41
                                                                 41
          10
                        7
                            110
                                  0.008000
                                              110 214.484055
                                                                      12.641945
                                                                                   110
                                                                110
                       xPos
                                         yPos
                                                     layer
                       mean count
                                         mean count mean count
          group
                 80.638710
                              155
                                    39.393548
                                                             155
          1
                                                 155
                                                         1
          6
                 90.156627
                               83 122.337349
                                                  83
                                                         1
                                                              83
          9
                 80.902439
                               41
                                   140.390244
                                                  41
                                                         1
                                                              41
          10
                 80.000000
                              110 214.000000
                                                 110
                                                             110
In [423]: for lidx, langle in titi.groupby('group').count().sort_values('angle', ascending=Fale
              print (lidx)
1
10
6
9
2
3
4
5
7
8
11
12
13
In [426]: getWAGA(titi,6)
Out [426]: 54.005035400390625
In [401]: testBitmap = np.zeros((frame.shape[0],frame.shape[1],3), np.uint8)
          testBitmap = drawFieldNeurons(titi, testBitmap,0,10)
          imgplot = plt.imshow(testBitmap)
```



8.2.5 Test 5

Generate data of type 5



Génération des neurones à champs récepteur

Out[429]:		longueur	angle	weight	precision	xPos	yPos	\
	count	95.0	95.000000	95.000000	95.000000	95.000000	95.000000	
	mean	7.0	0.007439	218.687790	10.881714	50.000000	62.000000	
	std	0.0	62.906307	14.875514	6.040379	3.313416	35.882491	
	min	7.0	-79.423332	200.650269	0.000000	41.000000	3.000000	
	25%	7.0	-62.404999	205.570328	7.588554	48.000000	32.500000	
	50%	7.0	0.010000	212.923187	13.785652	50.000000	62.000000	
	75%	7.0	62.421665	230.490402	15.747627	52.000000	91.500000	
	max	7.0	79.433334	242.906403	16.938763	59.000000	121.000000	

	group	layer
count	95.0	95.0
mean	0.0	1.0
std	0.0	0.0
min	0.0	1.0
25%	0.0	1.0
50%	0.0	1.0
75%	0.0	1.0
max	0.0	1.0

In [430]: titi.groupby('angle').agg(['mean', 'count'])

Out[430]:		longueur		weight		precision		xPos		\
		mean	count	mean	count	mean	count	mean	count	
	angle									
	-79.423332	7	10	222.054947	10	10.919160	10	50.000000	10	
	-74.995003	7	9	200.650269	9	16.938763	9	49.777778	9	
	-64.948334	7	4	212.735580	4	13.838958	4	48.500000	4	
	-62.404999	7	6	212.899750	6	13.792321	6	50.333333	6	
	-61.876667	7	6	205.558472	6	15.750603	6	50.333333	6	
	-54.903332	7	3	230.490402	3	7.588554	3	50.333333	3	
	0.010000	7	19	242.906403	19	0.000000	19	50.000000	19	
	54.923332	7	3	230.490402	3	7.588554	3	49.666667	3	
	61.893333	7	6	205.582184	6	15.744652	6	49.666667	6	
	62.421665	7	6	212.923187	6	13.785652	6	49.666667	6	
	64.964996	7	4	212.756851	4	13.832924	4	51.500000	4	
	75.004997	7	9	200.688141	9	16.929905	9	50.222222	9	
	79.433334	7	10	222.082733	10	10.909474	10	50.000000	10	

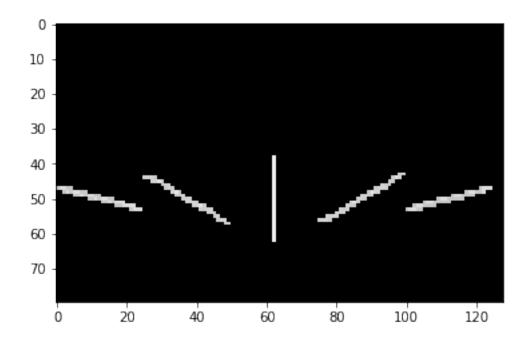
layer

group

yPos

	mean	count	mean	count	mean	count
angle						
-79.423332	12.500000	10	0	10	1	10
-74.995003	11.444444	9	0	9	1	9
-64.948334	34.500000	4	0	4	1	4
-62.404999	37.666667	6	0	6	1	6
-61.876667	37.666667	6	0	6	1	6
-54.903332	37.666667	3	0	3	1	3
0.010000	62.000000	19	0	19	1	19
54.923332	87.666667	3	0	3	1	3
61.893333	87.666667	6	0	6	1	6
62.421665	87.666667	6	0	6	1	6
64.964996	84.500000	4	0	4	1	4
75.004997	111.444444	9	0	9	1	9
79.433334	112.500000	10	0	10	1	10

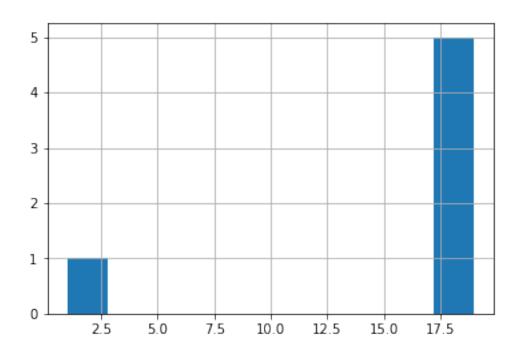
Affichage graphique du champs récepteur des neurones



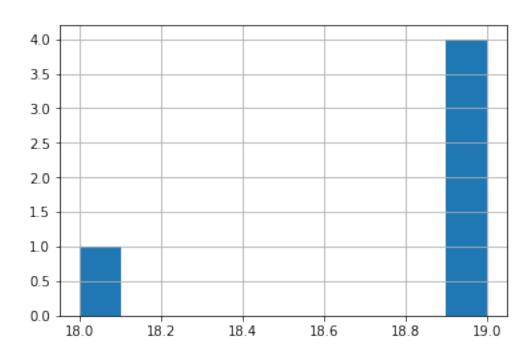
Génération des groupes

```
In [432]: findGroups(titi);
In [433]: titi.groupby('group').size().hist()
```

Out[433]: <matplotlib.axes._subplots.AxesSubplot at 0x12e882ba8>



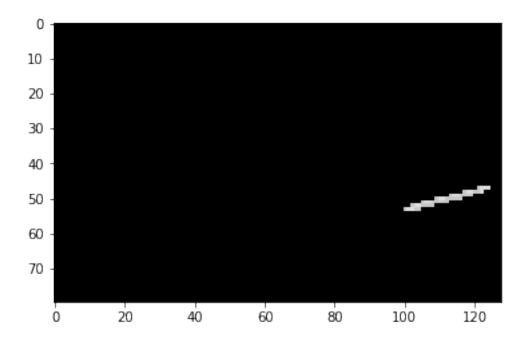
Out[434]: <matplotlib.axes._subplots.AxesSubplot at 0x13546e048>



In [435]: titi.groupby('group').agg(['mean', 'count'])[resultGroup>10]

Out[435]:	: longueur		angle	angle wei			precision		\	
		mean	count	mean	count	mean	count	mean	count	
	group									
	1	7	19	0.010000	19	242.906403	19	0.000000	19	
	2	7	19	-61.589123	19	213.324356	19	13.441002	19	
	3	7	18	61.419720	18	213.376328	18	13.413681	18	
	4	7	19	-77.325699	19	211.915894	19	13.770551	19	
	5	7	19	77.335701	19	211.948456	19	13.761257	19	
		vΡ	2	77	208	laver				

	xPos		yPos		layer	
	mean	count	mean	${\tt count}$	mean	count
group						
1	50.000000	19	62.000000	19	1	19
2	49.947368	19	37.000000	19	1	19
3	49.777778	18	87.44444	18	1	18
4	49.894737	19	12.000000	19	1	19
5	50.105263	19	112.000000	19	1	19



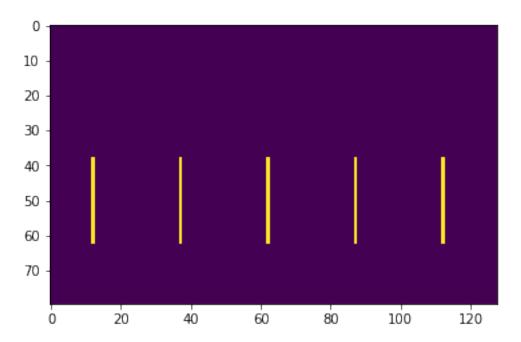
In [440]: getWAGA(titi,5)

Out [440]: 77.44712829589844

8.2.6 Test 6

Generate data of type 6

In [354]: frame = generateToy(6,80,128,1)
 imgplot = plt.imshow(frame)

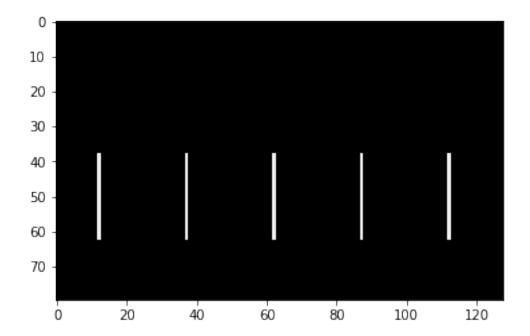


Génération des neurones à champs récepteur

In [356]: titi.groupby('angle').agg(['mean', 'count'])

Out [356]: longueur xPos yPos weight precision mean count mean count mean count mean count mean count angle 0.01 7 95 242.906403 95 0.0 95 50 95 62 95

Affichage graphique du champs récepteur des neurones

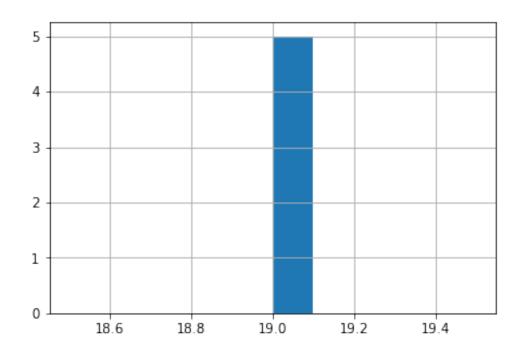


Génération des groupes

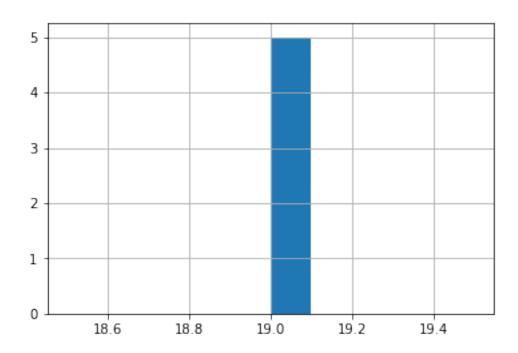
```
In [358]: findGroups(titi);
```

In [359]: titi.groupby('group').size().hist()

Out[359]: <matplotlib.axes._subplots.AxesSubplot at 0x133f20908>



Out[361]: <matplotlib.axes._subplots.AxesSubplot at 0x135c3b390>



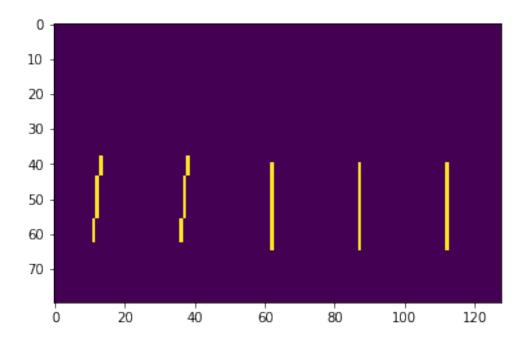
In [362]: titi.groupby('group').agg(['mean', 'count'])[resultGroup>10]

Out[362]:		longueur		angle		weight		precision		xPos		\
		mean	count	mean	count	mean	${\tt count}$	mean	${\tt count}$	${\tt mean}$	count	
	group											
	1	7	19	0.01	19	242.906403	19	0.0	19	50	19	
	2	7	19	0.01	19	242.906403	19	0.0	19	50	19	
	3	7	19	0.01	19	242.906403	19	0.0	19	50	19	
	4	7	19	0.01	19	242.906403	19	0.0	19	50	19	
	5	7	19	0.01	19	242.906403	19	0.0	19	50	19	

	yPos		layer	
	mean	count	mean	${\tt count}$
group				
1	12	19	1	19
2	37	19	1	19
3	62	19	1	19
4	87	19	1	19
5	112	19	1	19

8.2.7 Test 7

Generate data of type 7

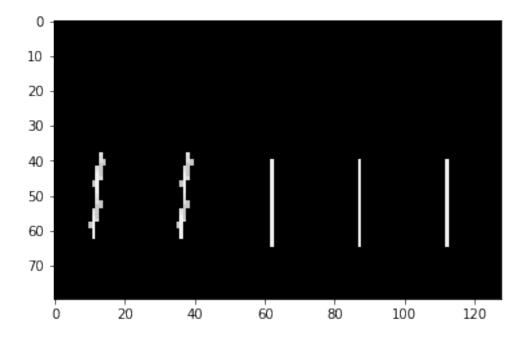


Génération des neurones à champs récepteur

In [443]: titi.groupby('angle').agg(['mean', 'count'])

Out[443]:	longueur			weight p			precision xPos				\	
		mean c	ount	1	nean	count	unt mean		count mean		count	
	angle											
	0.010000	7	71	242.90	6403	71	_	0.000000	71	51.774648	71	
	3.081667	7	8	232.03	9963	8	3	6.868455	8	49.500000	8	
	7.510000	7	8	222.21	3516	8	3 1	10.863776	8	49.500000	8	
	15.010000	7	8	200.669	9205	8	3 1	16.934334	8	49.500000	8	
		yPos		group		lay	er					
		mean	coun	t mean	coun	nt me	an	count				
	angle											
	0.010000	74.647887	7	1 0	7	1	1	71				
	3.081667	24.500000	;	8 0		8	1	8				
	7.510000	24.500000	;	8 0		8	1	8				
	15.010000	24.500000	;	8 0		8	1	8				

Affichage graphique du champs récepteur des neurones

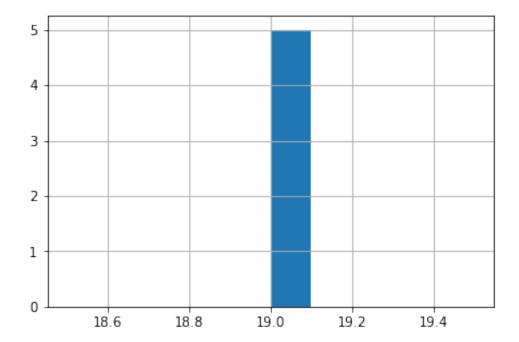


Génération des groupes

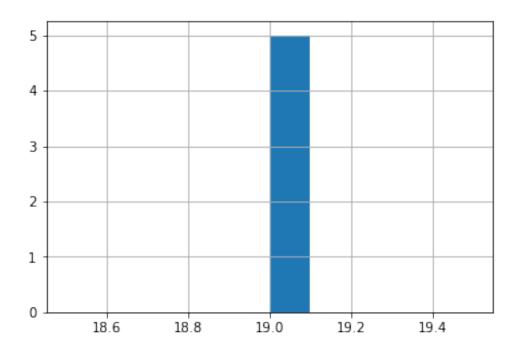
In [445]: findGroups(titi);

In [446]: titi.groupby('group').size().hist()

Out[446]: <matplotlib.axes._subplots.AxesSubplot at 0x136c1d358>



Out[447]: <matplotlib.axes._subplots.AxesSubplot at 0x132eea0f0>



In [448]: titi.groupby('group').agg(['mean', 'count'])[resultGroup>10]

Out[448]:		longueur		angle		weight		precision		xPos	\
		mean	count	mean	count	mean	count	mean	${\tt count}$	mean	
	group										
	1	7	19	5.393509	19	227.370285	19	7.298224	19	50	
	2	7	19	5.393509	19	227.370285	19	7.298224	19	50	
	3	7	19	0.010000	19	242.906403	19	0.000000	19	52	
	4	7	19	0.010000	19	242.906403	19	0.000000	19	52	
	5	7	19	0.010000	19	242.906403	19	0.000000	19	52	

		yPos		layer	
	count	mean	${\tt count}$	mean	count
group					
1	19	11.947368	19	1	19
2	19	36.947368	19	1	19
3	19	62.000000	19	1	19
4	19	87.000000	19	1	19
5	19	112.000000	19	1	19

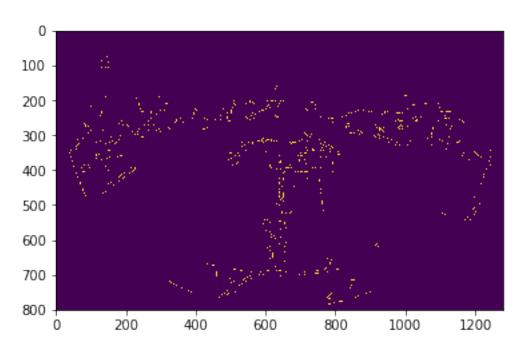
In [450]: getWAGA(titi,2)

Out[450]: 5.000133991241455

8.2.8 Test 8 : video frame

Get Video frame

In [302]: frame = Cannyframe
 imgplot = plt.imshow(frame)



Génération des neurones à champs récepteur

In [304]: titi.describe()

Out[304]:		longueur	${\tt angle}$	weight	precision	xPos	\
	count	6352.0	6352.000000	6352.000000	6352.000000	6352.000000	
	mean	9.0	3.500042	155.807388	25.445107	444.441751	
	std	0.0	49.856220	75.384323	17.663477	153.220702	
	min	9.0	-88.243752	1.932616	0.000000	267.000000	
	25%	9.0	-35.947117	106.064775	13.851732	325.000000	
	50%	9.0	0.010000	184.319527	20.414985	388.000000	
	75%	9.0	47.406427	212.690514	33.394614	547.000000	
	max	9.0	90.000000	242.906403	78.747810	784.000000	

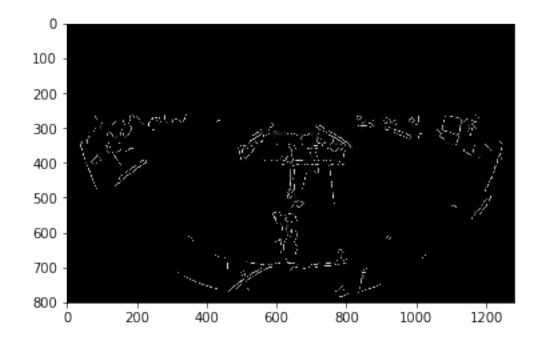
	yPos	group	layer
count	6352.000000	6352.0	6352.0
mean	649.877204	0.0	1.0
std	296.119564	0.0	0.0

```
39.000000
                        0.0
                                 1.0
min
25%
        523.750000
                        0.0
                                 1.0
50%
        653.000000
                        0.0
                                 1.0
75%
        788.000000
                        0.0
                                 1.0
                        0.0
                                 1.0
       1246.000000
max
```

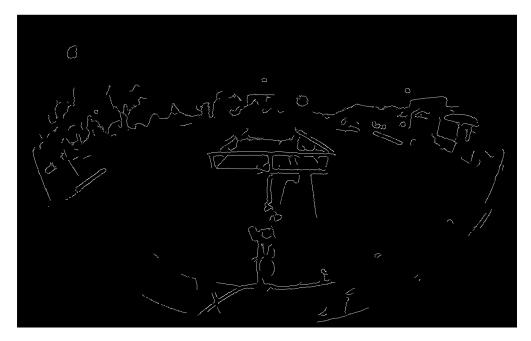
In [305]: titi[0:4]

Out[305]:	longueur	angle	weight	precision	xPos	yPos	group	layer
0	9	-14.750000	209.718018	14.671453	267	82	0	1
1	9	-18.597778	178.442535	21.537748	267	180	0	1
2	9	-40.653751	139.374527	28.131916	267	191	0	1
3	9	-2.295000	151.148193	26.247042	267	266	0	1

Affichage graphique du champs récepteur des neurones



```
In [307]: lintI = 0
    while (lintI < 10):
        cv2.imshow('testBitmap', testBitmap)
        if cv2.waitKey(1) & OxFF == ord('q'): # press q to quit
            break
        lintI += 1</pre>
```



Edge detection frame image, with the Canny Algorithm

Affichage de la frame :

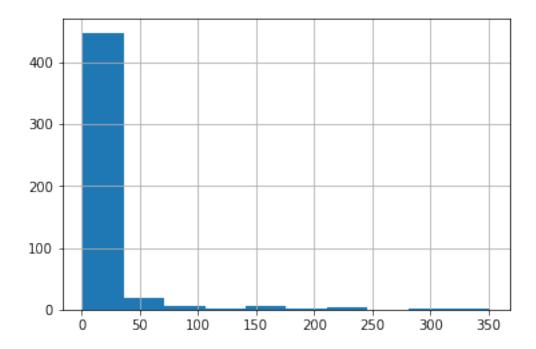
Affichage des champs récepteurs des neurones

Génération des groupes

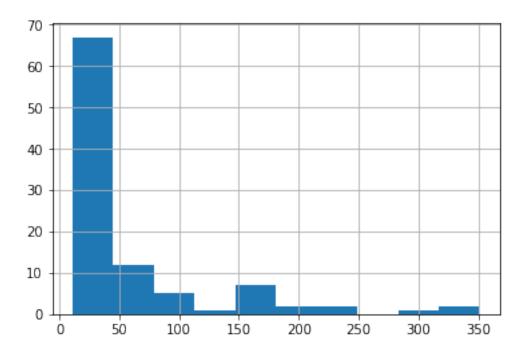
```
In [308]: findGroups(titi);
In [309]: titi.groupby('group').size().hist()
Out[309]: <matplotlib.axes._subplots.AxesSubplot at 0x13460b908>
```



NeuronField Image: Le niveau de gris correspond au niveau d'activation du neurone



Out[310]: <matplotlib.axes._subplots.AxesSubplot at 0x13464f630>



In [371]: titi.groupby('group').agg(['mean', 'count'])[resultGroup>130]

```
KeyError
```

Traceback (most recent call last)

```
<ipython-input-371-c1c777f13774> in <module>
----> 1 titi.groupby('group').agg(['mean', 'count'])[resultGroup>130].sort_values(by=['count'])
```

```
/anaconda3/lib/python3.7/site-packages/pandas/core/frame.py in sort_values(self, by, at 4717 by = by[0]
```

if isinstance(ascending, (tuple, list)):

KeyError: 'count'

9 Errors list

9.1 Error 10

Problème dans la fonction Section ??

In []: