2023-05-15_ChartTypes_012_VectorFields_DeformationImages

May 25, 2023

```
import numpy as np
import scipy
import imageio

import matplotlib
import matplotlib.pyplot as plt
import matplotlib.cm as cm

matplotlib.rc('image', interpolation='nearest')
matplotlib.rc('figure',facecolor='white')
matplotlib.rc('image',cmap='viridis')
colors=plt.rcParams['axes.prop_cycle'].by_key()['color']
%matplotlib inline
```

1 Vector fields

1.1 Deformation plots

```
[2]: # build some non-trivial examples of deformations for us to visualize
     def map1(x,y):
         """merely rescale radius a bit"""
         rIn=np.sqrt(x**2+y**2)
         phiIn=np.arctan2(y,x)
         rOut=rIn**0.8
         phiOut=phiIn
         u=np.cos(phiOut)*rOut
         v=np.sin(phiOut)*rOut
         return u, v
     def map1inv(x,y):
         rIn=np.sqrt(x**2+y**2)
         phiIn=np.arctan2(y,x)
         rOut=rIn**(1/0.8)
         phiOut=phiIn
         u=np.cos(phiOut)*rOut
```

```
v=np.sin(phiOut)*rOut
    return u, v
def map2(x,y):
    """rescale radius + fixed rotation"""
    rIn=np.sqrt(x**2+y**2)
    phiIn=np.arctan2(y,x)
    rOut=rIn**0.8
    phiOut=phiIn+0.25*np.pi
    u=np.cos(phiOut)*rOut
    v=np.sin(phiOut)*rOut
    return u,v
def map2inv(x,y):
    rIn=np.sqrt(x**2+y**2)
    phiIn=np.arctan2(y,x)
    rOut=rIn**(1/0.8)
    phiOut=phiIn-0.25*np.pi
    u=np.cos(phiOut)*rOut
    v=np.sin(phiOut)*rOut
    return u,v
def map3(x,y):
    """rescale radius + adaptive rotation"""
    rIn=np.sqrt(x**2+y**2)
    phiIn=np.arctan2(y,x)
    rOut=rIn**0.8
    phiOut=phiIn+0.25*np.pi*rIn
    u=np.cos(phiOut)*rOut
    v=np.sin(phiOut)*rOut
    return u, v
def map3inv(x,y):
    rIn=np.sqrt(x**2+y**2)
    phiIn=np.arctan2(y,x)
    rOut=rIn**(1/0.8)
    phiOut=phiIn-0.25*np.pi*rOut
    u=np.cos(phiOut)*rOut
    v=np.sin(phiOut)*rOut
    return u,v
usemap=map3
usemapinv=map3inv
```

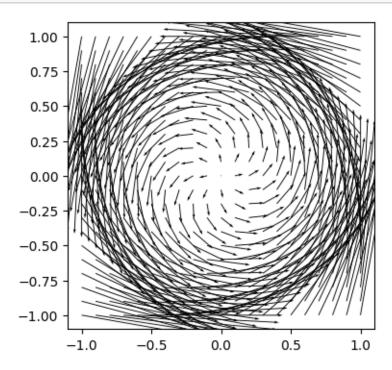
```
def getUniformData(nPts1d,rng,usemap):
    x = np.linspace(*rng,num=nPts1d)
    X, Y = np.meshgrid(x, x)
    U, V = usemap(X,Y)
    return X,Y,U,V
```

1.1.1 Visualization as quiver plot

```
[3]: X,Y,U,V=getUniformData(21,[-1,1],usemap)

fig=plt.figure(figsize=(4,4))
ax=fig.add_subplot(aspect=1.)
ax.quiver(X,Y,U-X,V-Y,pivot="tail",angles="xy",scale_units="xy",scale=1.)
plt.tight_layout()
plt.show()

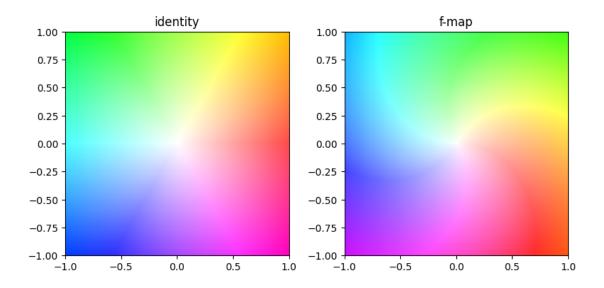
# shown earlier: this tends not to work well on actual maps with long vectors
```



1.1.2 Visualization as HSV image

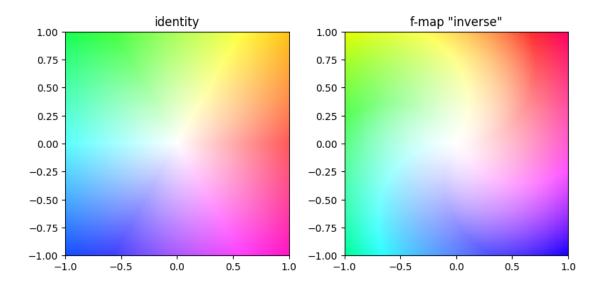
```
[4]: def HSVfromUV(U,V,magmax=None):
         colorsHSV=np.zeros(U.shape+(3,))
         # set Saturation to 1
         colorsHSV[:,:,2]=1.
         # set Value to (rescaled) magnitude
         magnitude=(U**2+V**2)**0.5
         if magmax is None:
             vmax=np.max(magnitude)
         else:
             vmax=magmax
         colorsHSV[:,:,1]=np.clip(magnitude/vmax,0.,1.)
         # set Hue to orientation
         phi=np.mod(np.arctan2(V,U)/(2*np.pi),1.)
         colorsHSV[:,:,0]=phi
         # conver to RGB
         colorsRGB=matplotlib.colors.hsv_to_rgb(colorsHSV)
         return colorsRGB
```

```
[5]: nPts1d=101
     rng=[-1,1]
     X,Y,U,V=getUniformData(nPts1d,rng,usemap)
     magmax=max(np.max((U**2+V**2)**0.5),np.max((X**2+Y**2)**0.5))
     img1=HSVfromUV(X,Y,magmax)
     # try: color each pixel by color corresponding to target position
     img2=HSVfromUV(U,V,magmax)
     fig=plt.figure(figsize=(8,4))
     fig.add_subplot(1,2,1,aspect=1.)
     plt.title("identity")
     plt.imshow(img1,extent=rng+rng,origin="lower")
     fig.add_subplot(1,2,2,aspect=1.)
     plt.title("f-map")
     plt.imshow(img2,extent=rng+rng,origin="lower")
     plt.tight_layout()
     plt.show()
```



```
[6]: # comments:
# * see how perceptually non-uniform HSV is!
# * in this way, displayed side-by-side like this, it seems like map is
# a rotation in the opposite direction!
# -> we should color each pixel by the color corresponding to its preimage
```

```
[7]: nPts1d=101
     rng = [-1, 1]
     X,Y,U,V=getUniformData(nPts1d,rng,usemapinv)
     magmax=max(np.max((U**2+V**2)**0.5),np.max((X**2+Y**2)**0.5))
     img1=HSVfromUV(X,Y,magmax)
     # try: color each pixel by color corresponding to target position
     img3=HSVfromUV(U,V,magmax)
     fig=plt.figure(figsize=(8,4))
     fig.add_subplot(1,2,1,aspect=1.)
     plt.title("identity")
     plt.imshow(img1,extent=rng+rng,origin="lower")
     fig.add_subplot(1,2,2,aspect=1.)
     plt.title("f-map \"inverse\"")
     plt.imshow(img3,extent=rng+rng,origin="lower")
     plt.tight_layout()
     plt.show()
```



1.1.3 Visualization as warped grid

```
[9]: rng=[-1,1]
nLines=15
nPtsPerLine=20
X,Y=getLines(rng,nLines,nPtsPerLine)
U,V=usemap(X,Y)

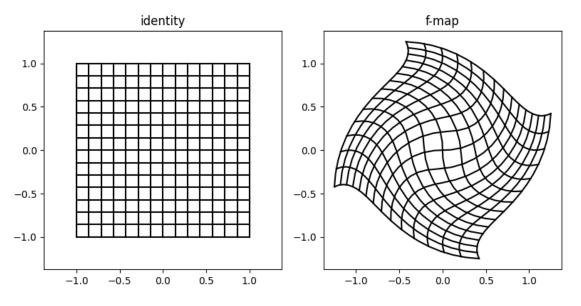
# get consistent plot range in both images
magmax=max([np.max(np.abs(Z)) for Z in [X,Y,U,V]])*1.1

fig=plt.figure(figsize=(8,4))

fig.add_subplot(1,2,1,aspect=1.)
plt.title("identity")
for x,y in zip(X,Y):
```

```
plt.plot(x,y,c="k")
plt.xlim([-magmax,magmax])
plt.ylim([-magmax,magmax])

fig.add_subplot(1,2,2,aspect=1.)
plt.title("f-map")
for x,y in zip(U,V):
    plt.plot(x,y,c="k")
plt.xlim([-magmax,magmax])
plt.ylim([-magmax,magmax])
plt.tight_layout()
plt.show()
```



```
[10]: # experiment: add color to lines to make identification a bit more obvious in
    # ambiguous cases
    # try also: line style (solid vs dashed)
    # make lines a bit thicker, so that these features (color, dash pattern)
    # can be seen better

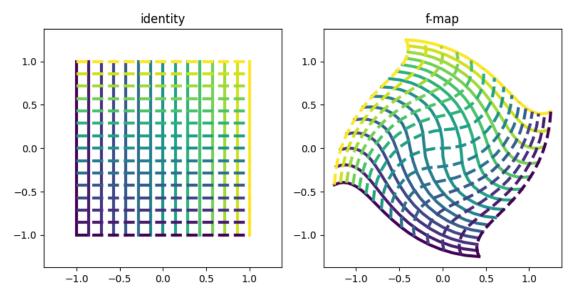
colors=np.linspace(0,1,num=nLines)
    colors=np.concatenate((colors,colors))
    linestyle=["solid" for _ in range(nLines)]+["dashed" for _ in range(nLines)]

fig=plt.figure(figsize=(8,4))

fig.add_subplot(1,2,1,aspect=1.)
    plt.title("identity")
```

```
for x,y,c,ls in zip(X,Y,colors,linestyle):
    plt.plot(x,y,c=cm.viridis(c),ls=ls,lw=3)
plt.xlim([-magmax,magmax])
plt.ylim([-magmax,magmax])

fig.add_subplot(1,2,2,aspect=1.)
plt.title("f-map")
for x,y,c,ls in zip(U,V,colors,linestyle):
    plt.plot(x,y,c=cm.viridis(c),ls=ls,lw=3)
plt.xlim([-magmax,magmax])
plt.ylim([-magmax,magmax])
plt.tight_layout()
plt.show()
```

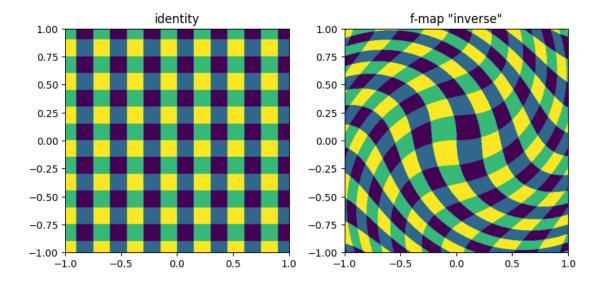


1.1.4 Visualization as warped grid, variant 2

This is a fusion of the two previous approaches. Visualize deformation as colored image, but instead of encoding full position as color, only encode position in a grid.

```
[11]: def colorFromPos(X,Y,res):
    # divide coords into square grid of edge length res
    # determine color of "which square" pixel is in
    indexX=np.clip(np.floor(np.mod((X/res),2)),0,1)
    indexY=np.clip(np.floor(np.mod((Y/res),2)),0,1)
    return indexX*0.67+indexY*0.33
```

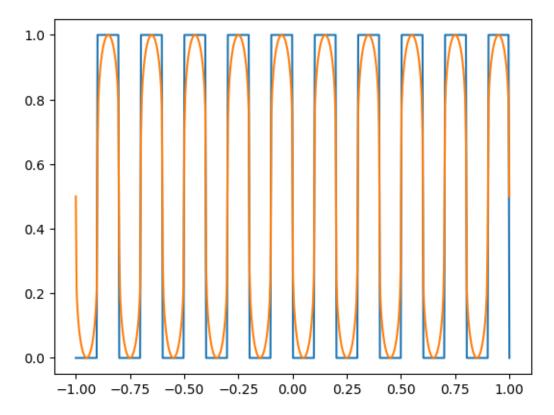
```
[12]: nPts1d=500
      rng=[-1,1]
      res=0.15
      X,Y,U,V=getUniformData(nPts1d,rng,usemapinv)
      img1=colorFromPos(X,Y,res)
      img3=colorFromPos(U,V,res)
      fig=plt.figure(figsize=(8,4))
      # here an interpolation mode that is not "nearest" is more appropriate
      # to dampen discretization artefacts
      fig.add_subplot(1,2,1,aspect=1.)
      plt.title("identity")
      plt.imshow(img1,extent=rng+rng,origin="lower",interpolation="bilinear")
      fig.add_subplot(1,2,2,aspect=1.)
      plt.title("f-map \"inverse\"")
      plt.imshow(img3,extent=rng+rng,origin="lower",interpolation="bilinear")
      plt.tight_layout()
      plt.show()
```



```
[13]: # another alternative: use a smoother pattern function
```

```
[14]: # visualize discrete grid vs smoth oscillation def comb(x,res):
```

```
return np.floor(np.mod(x/res,2))
def smooth_comb(x,res):
    cosdat=-np.sin(x/res*np.pi)
    return 0.5*(1+np.sign(cosdat)*np.abs(cosdat)**0.25)
x=np.linspace(-1,1,num=500)
plt.plot(x,comb(x,0.1))
plt.plot(x,smooth_comb(x,0.1))
plt.show()
```



```
[15]: def colorFromPos2(X,Y,res):
    # divide coords into square grid of edge length res
    # determine color of "which square" pixel is in
    indexX=smooth_comb(X,res)
    indexY=smooth_comb(Y,res)
    return indexX*0.67+indexY*0.33
```

```
[16]: nPts1d=500
    rng=[-1,1]
    res=0.15

X,Y,U,V=getUniformData(nPts1d,rng,usemapinv)
```

```
img1=colorFromPos2(X,Y,res)
img3=colorFromPos2(U,V,res)

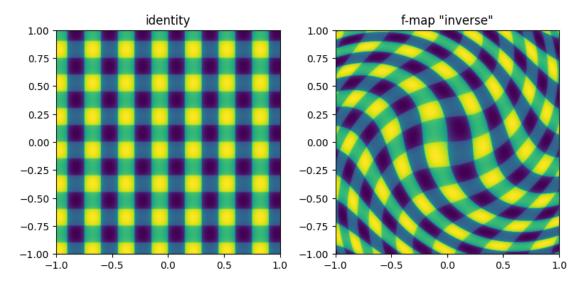
fig=plt.figure(figsize=(8,4))

fig.add_subplot(1,2,1,aspect=1.)
plt.title("identity")
plt.imshow(img1,extent=rng+rng,origin="lower",interpolation="bilinear")

fig.add_subplot(1,2,2,aspect=1.)
plt.title("f-map \"inverse\"")
plt.title("f-map \"inverse\"")
plt.imshow(img3,extent=rng+rng,origin="lower",interpolation="bilinear")

plt.tight_layout()

plt.show()
```



[]: