# **Superpixel Segmentation**

#### **Import Modules**

```
In [2]: import vigra
    from vigra import graphs
    import numpy
    import opengh
    import matplotlib
    import pylab
    try:
        from sklearn.cluster import MiniBatchKMeans, KMeans
        from sklearn import mixture
    except:
        raise RuntimeError("this examples nees sklearn")
```

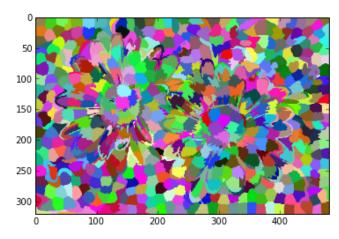
#### **Load Image**

```
In [3]: # parameter:
filepath = '124084.jpg'  # input image path
    # load image and convert to LAB
    img = vigra.impex.readImage(filepath)
    # get super-pixels with slic on LAB image
    imgLab = vigra.colors.transform_RGB2Lab(img)

vigra.imshow(img)
vigra.show()
```



### **Superpixel Segmentation and RAG**



#### **Node Features**

# In [46]: # accumulate node features from grid graph node map # which is just a plain image (with channels) nodeFeatures = rag.accumulateNodeFeatures(imgLab) nodeFeaturesImg = rag.projectNodeFeaturesToGridGraph(nodeFeatures) nodeFeaturesImg = vigra.taggedView(nodeFeaturesImg, "xyc") nodeFeaturesImgRgb = vigra.colors.transform\_Lab2RGB(nodeFeaturesImg) vigra.imshow(nodeFeaturesImgRgb)



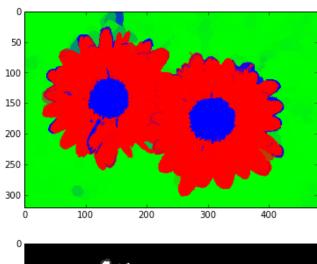
Out[46]: <matplotlib.image.AxesImage at 0xf49f9d0>

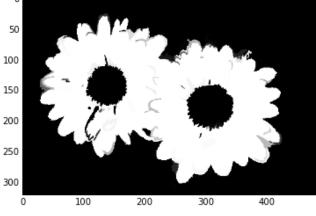
# **GMM** clustering to get cluster probabilities

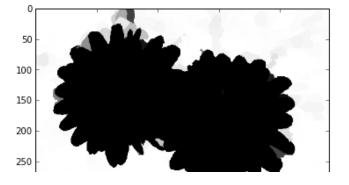
```
In [53]: nCluster = 3
    g = mixture.GMM(n_components=nCluster)
    g.fit(nodeFeatures[:,:])
    clusterProb = g.predict_proba(nodeFeatures)

clusterProbImg = rag.projectNodeFeaturesToGridGraph(clusterProb.astype(numpy.float32
    clusterProbImg = vigra.taggedView(clusterProbImg, "xyc")
    vigra.imshow(clusterProbImg)
    vigra.show()

for c in range(nCluster):
    clusterProbImg = rag.projectNodeFeaturesToGridGraph(clusterProb[:,c].astype(numpy.float32)
    clusterProbImg = vigra.taggedView(clusterProbImg, "xy")
    vigra.imshow(clusterProbImg)
    vigra.show()
```







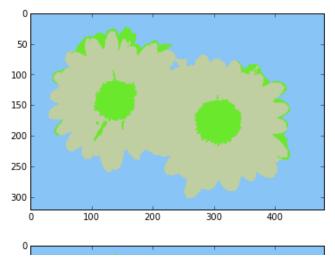
## Potts regularization

```
In [50]: # strength of potts regularizer
         beta = 10.0
         # graphical model with as many variables
         # as superpixels, each has 3 states
         gm = opengm.gm(numpy.ones(rag.nodeNum,dtype=opengm.label_type)*nCluster)
         # convert probabilites to energies
         probs = numpy.clip(clusterProb, 0.00001, 0.99999)
         costs = -1.0*numpy.log(probs)
         # add ALL unaries AT ONCE
         fids = gm.addFunctions(costs)
         gm.addFactors(fids,numpy.arange(rag.nodeNum))
         # add a potts function
         regularizer = opengm.pottsFunction([nCluster]*2,0.0,beta)
         fid = gm.addFunction(regularizer)
         # get variable indices of adjacent superpixels
         # - or "u" and "v" node id's for edges
         uvIds = rag.uvIds()
         uvIds = numpy.sort(uvIds,axis=1)
         # add all second order factors at once
         gm.addFactors(fid,uvIds)
```

Out[50]: 3184

```
In [51]: Inf = opengm.inference.BeliefPropagation
         parameter = opengm.InfParam(steps=10,damping=0.5,convergenceBound=0.001)
         inf = Inf(gm,parameter=parameter)
         class PyCallback(object):
             def __init__(self,):
                 self.labels=[]
             def begin(self,inference):
                 print "begin of inference"
             def end(self,inference):
                 self.labels.append(inference.arg())
             def visit(self,inference):
                 gm=inference.gm()
                 labelVector=inference.arg()
                 print "energy ",gm.evaluate(labelVector)
                 self.labels.append(labelVector)
         callback=PvCallback()
         visitor=inf.pythonVisitor(callback,visitNth=1)
         inf.infer(visitor)
         cmap = matplotlib.colors.ListedColormap ( numpy.random.rand ( nCluster,3))
         for arg in callback.labels:
             arg = vigra.taggedView(arg, "n")
             argImg = rag.projectNodeFeaturesToGridGraph(arg.astype(numpy.uint32))
             argImg = vigra.taggedView(argImg, "xy")
             # plot superpixels
             pylab.imshow ( argImg.swapaxes(0,1).squeeze(), cmap = cmap)
             pylab.show()
         begin of inference
         energy
                   3068.16256623
```

2699.42655288 energy 2491.21469108 energy 2431.5037594 energy 2317.53743441 energy 2274.84824561 energy 2204.10503333 energy 2181.56009123 energy energy 2183.74267258 2183.74267258 energy



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In [7]:			