任务 1:

计算每个字母的概率,分别取最大值为结果

任务 2:

1)计算每个字母的识别结果概率,并联合两个字母的连续概率,使用 ICM 进行迭代求最大值。使用任务 1 中的结果作为迭代起始值。

2) 使用维特比算法求解最大值

```
numChar=length(images);
pred=zeros([numChar,1]);
K=imageModel.K;
p1=ones([numChar,K]);
for n=1:numChar
     for k=1:K-1 p1(n,k)=exp(imageModel.W(k,:)*images(n).img(:)+imageModel.bias(k)); end
     p1(n,:)=p1(n,:)./sum(p1(n,:));
if numChar<3
return
end
pred_m=zeros([K,numChar]);
for n=1:K
    pred_m(n,1)=n;
end
p_m=log(p1(1,:));
for n=2:numChar
          p(k)=log(p1(n,k))+log(pairwiseModel(pred_m(k_m,n-1),k));
end
          [P,M]=max(p);
         p_m(k_m)=p_m(k_m)+P;
pred_m(k_m,n)=M;
[~,M]=max(p_m);
pred=pred_m(M,:);
```

任务 3:

1) 计算每个字母的识别结果概率,并联合两个字母和三个字母的连续概率,使用 ICM 进行迭代求最大值。使用任务 1 中的结果作为迭代起始值。

```
numChar=length(images);
K=imageModel.K;
pred=randi(K,[numChar,1]);
p1=ones([numChar,K]);
for n=1:numChar
       p1(n,k)=exp(imageModel.W(k,:)*images(n).img(:)+imageModel.bias(k));
    p1(n,:)=p1(n,:)./sum(p1(n,:));
[~,pred]=max(p1,[],2); %迭代初始值|
if numChar<4
   return
end
max_p=zeros([numChar,1]);
while 1
    for n=1:numChar
        p=zeros([K,1]);
        for k =1:K
            pred(n)=k;
            p(k) = \log(p1(numChar, pred(numChar))) + \log(p1(numChar-1, pred(numChar-1))) + \log(pairwiseModel(pred(numChar-1), pred(numChar)));
            for w=1:numChar-2
                                 %计算单词条件概率
                p(k) = p(k) + \log(p1(w, pred(w))) + \log(pairwiseModel(pred(w), pred(w+1))) + \log(tripletModel(pred(w), pred(w+1), pred(w+2)));
        end
        [max_p(n),pred(n)]=max(p);
    end
    if all (~(diff(max_p)))
        break
    end
```

2) 使用任务2中维特比算法求得结果作为初始值。

```
numChar=length(images);
K=imageModel.K;
pred=randi(K,[numChar,1]);
p1=ones([numChar,K]);
for n=1:numChar
    for k=1:K-1
       p1(n,k)=exp(imageModel.W(k,:)*images(n).img(:)+imageModel.bias(k));
    p1(n,:)=p1(n,:)./sum(p1(n,:));
pred = RunInference2w(images, imageModel, pairwiseModel, tripletModel); %迭代初始值
if numChar<4
    return
max_p=zeros([numChar,1]);
while 1
    for n=1:numChar
        p=zeros([K,1]);
            pred(n)=k;
             p(k) = \log(p1(numChar, pred(numChar))) + \log(p1(numChar-1, pred(numChar-1))) + \log(pairwiseModel(pred(numChar-1), pred(numChar)));
             for w=1:numChar-2
                p(k) = p(k) + \log(p1(w, pred(w))) + \log(pairwiseModel(pred(w), pred(w+1))) + \log(tripletModel(pred(w), pred(w+1), pred(w+2)));
            end
        [\max_p(n), pred(n)] = \max(p);
    end
    if all (~(diff(max_p)))
        break
    end
```

任务 4:

1) 计算每个字母的识别结果概率,并联合两个字母和三个字母的连续概率、以及图像相似度最大的两个图像,使用 ICM 进行迭代求最大值。使用任务 1 中的结果作为迭代起始值。

2) 使用任务 2 中维特比算法求得结果作为初始值。

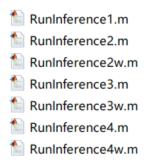
```
numChar=length(images);
K=imageModel.K;
pred=randi(K,[numChar,1]);
p1=ones([numChar,K]);
for n=1:numChar
     for k=1:K-1

p1(n,k)=exp(imageModel.W(k,:)*images(n).img(:)+imageModel.bias(k));

end
    p1(n,:)=p1(n,:)./sum(p1(n,:));
pred = RunInference2w(images, imageModel, pairwiseModel, tripletModel);
if numChar<4
    return
max_p=zeros([numChar,1]);
while 1
     for n=1:numChar
         p=zeros([K,1]);
for k =1:K
pred(n)=k;
              p(k)=log(p1(numChar,pred(numChar)))+log(p1(numChar-1,pred(numChar-1)))+log(pairwiseModel(pred(numChar-1),pred(numChar)));
              p(k)=p(k)+log(p1(w,pred(w)))+log(pairwiseModel(pred(w),pred(w+1)))+log(tripletModel(pred(w),pred(w+1),pred(w+2))); end c=zeros([numChar*(numChar*(numChar*1)/2,1]);
              c0=1;
for i =1:numChar-1
                   for j =i+1:numChar
    if pred(i)~=pred(j)
        c(c0)=1;
                            c(c0)=ImageSimilarity(images(i).img,images(j).img);
                       c0=c0+1;
              p(k)=p(k)+log(c(length(c)))+log(c(length(c)-1));
         end
[max_p(n),pred(n)]=max(p);
    end
if all (~(diff(max_p)))
```

结果分析:

不同的算法分别存在不同的函数中,其中带有 w 的函数为使用维特比算法或使用维特比路径作为初始值进行 ICM 迭代。



通过修改 wordPredictions.m 中使用的识别函数,可以得到不同的结果。

```
numWords = length(allWords);
wordPredictions = cell(numWords, 1);

for i = 1:numWords
     wordPredictions{i} = RunInference4w(allWords{i}, imageModel, pairwiseModel, tripletModel);
end
end
```

运行的结果如下所示:

	ICM	维特比
任 务 1	530 / 691 characters (76.70% accuracy) 22 / 100 words (22.00% accuracy)	_
任 务 2	529 / 691 characters (76.56% accuracy) 25 / 100 words (25.00% accuracy)	531 / 691 characters (76.85% accuracy) 28 / 100 words (28.00% accuracy)
任 务 3	535 / 691 characters (77.42% accuracy) 30 / 100 words (30.00% accuracy)	547 / 691 characters (79.16% accuracy) 34 / 100 words (34.00% accuracy)
任 务 4	540 / 691 characters (78.15% accuracy) 32 / 100 words (32.00% accuracy)	548 / 691 characters (79.31% accuracy) 35 / 100 words (35.00% accuracy)