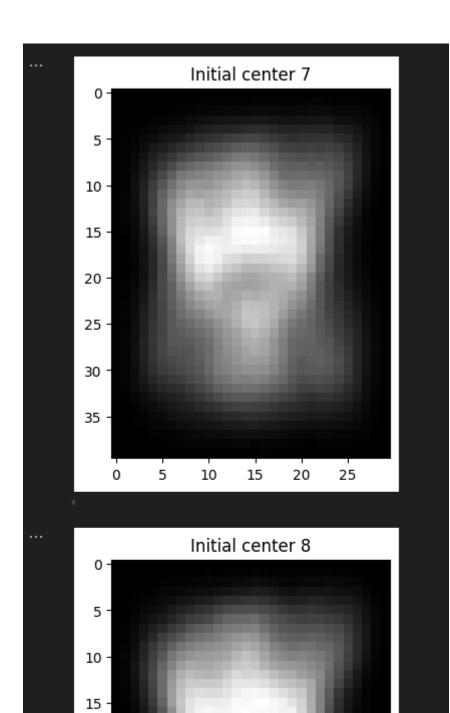


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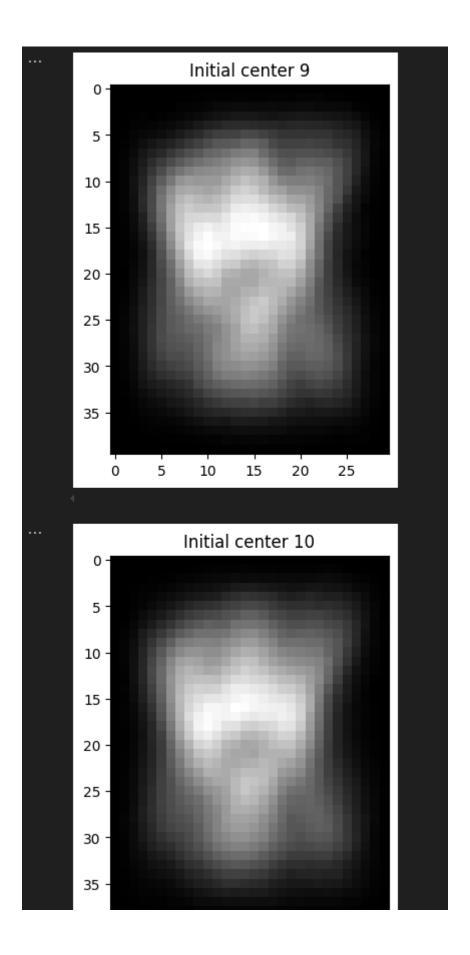


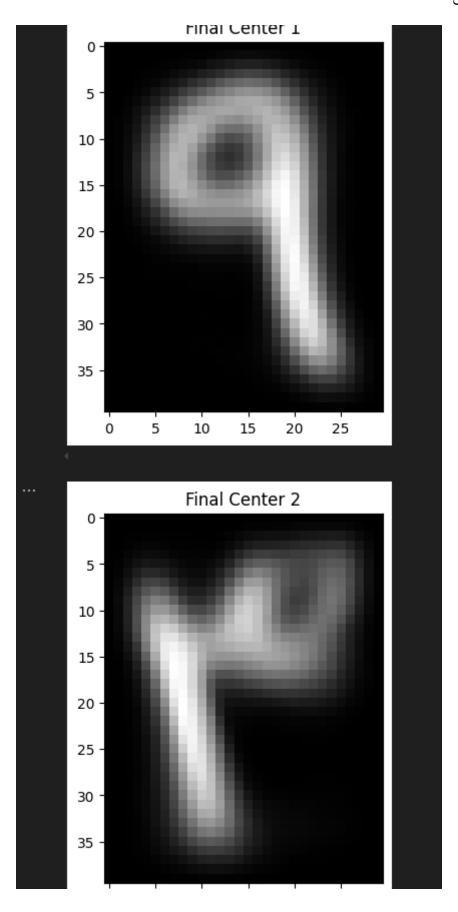
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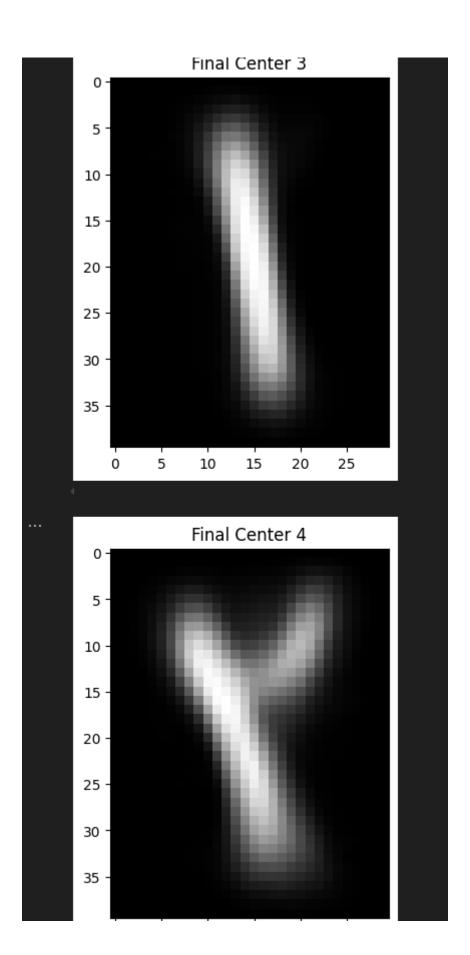
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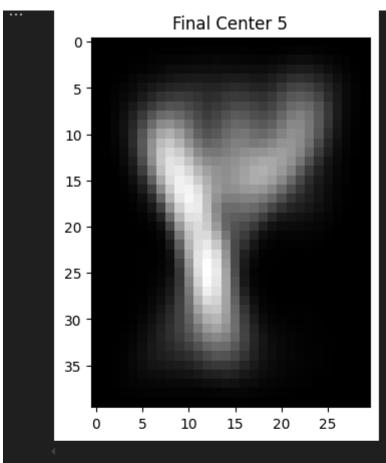
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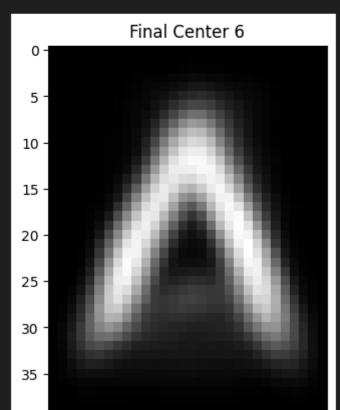
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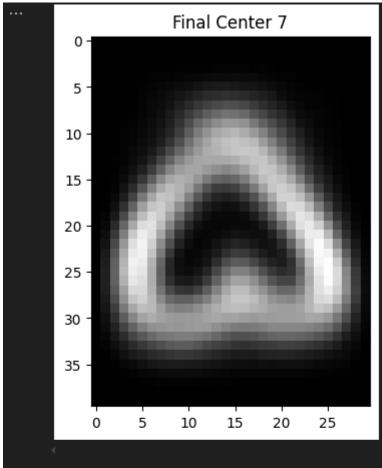


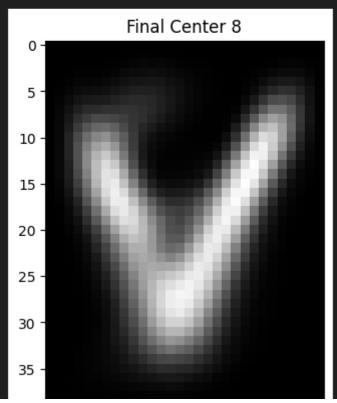


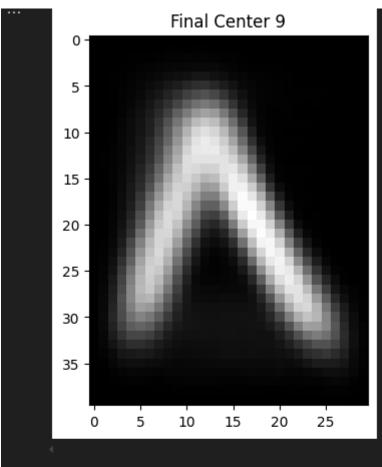


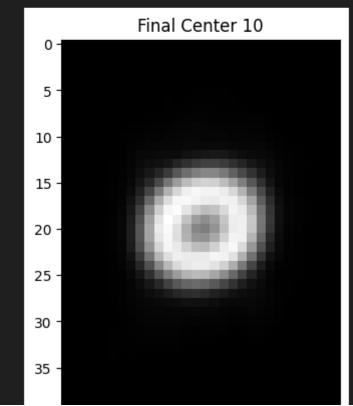


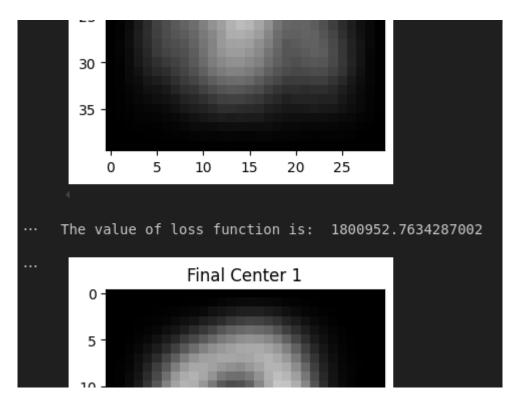


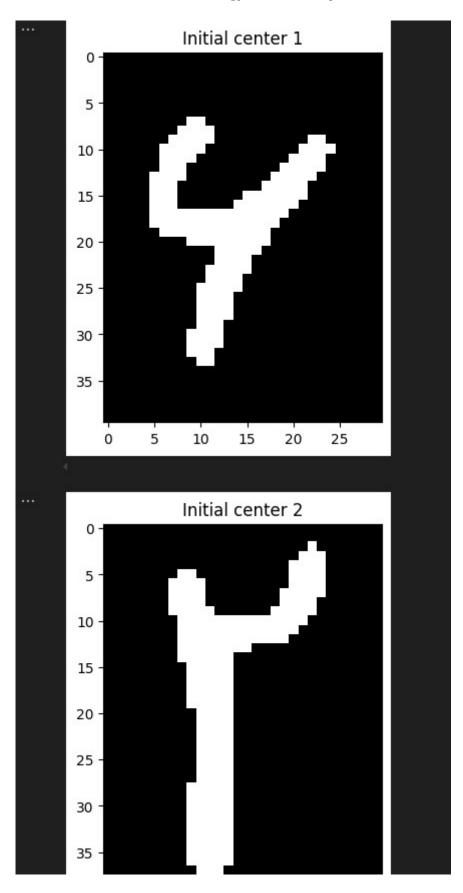


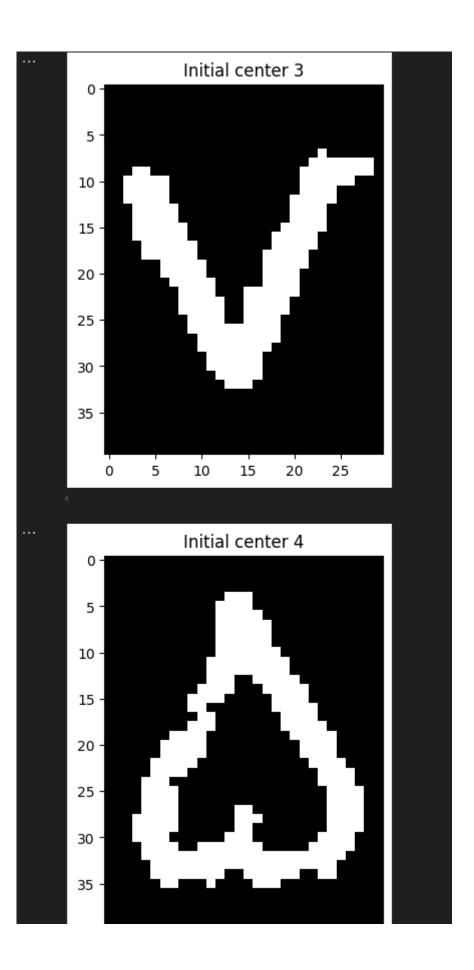


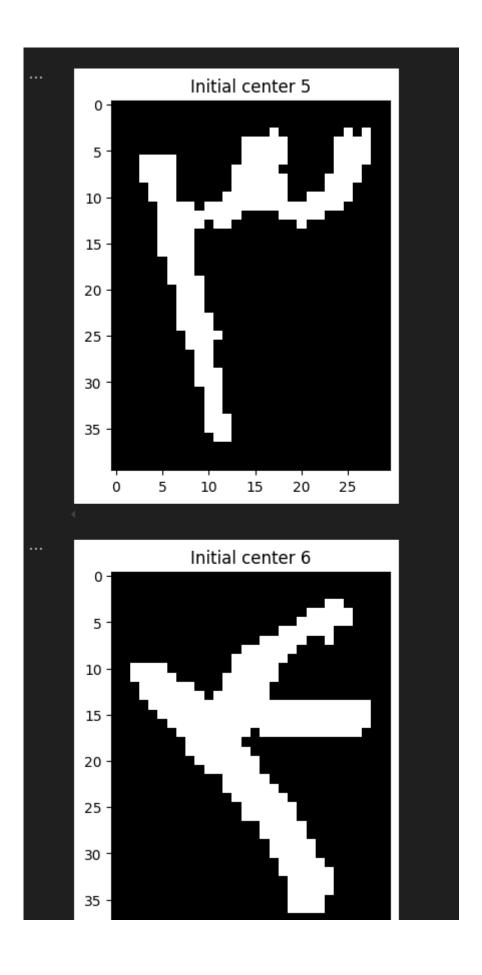


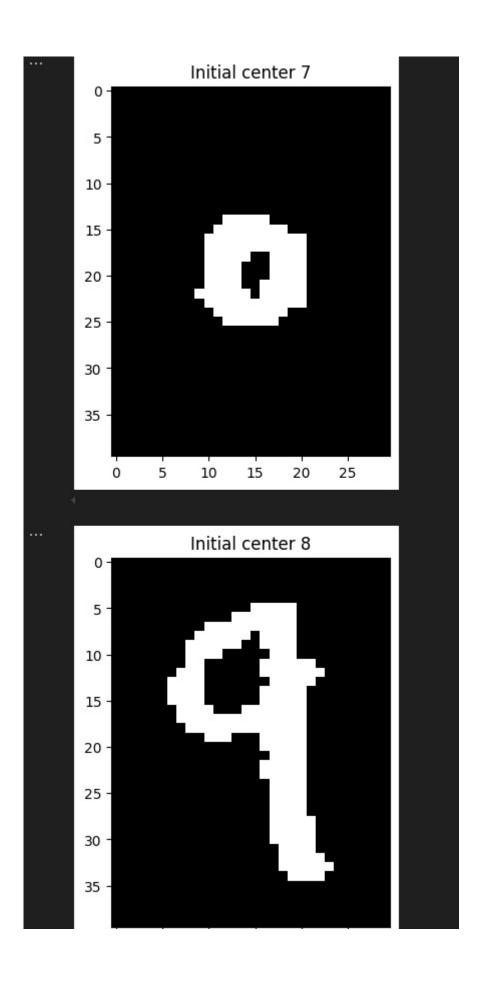


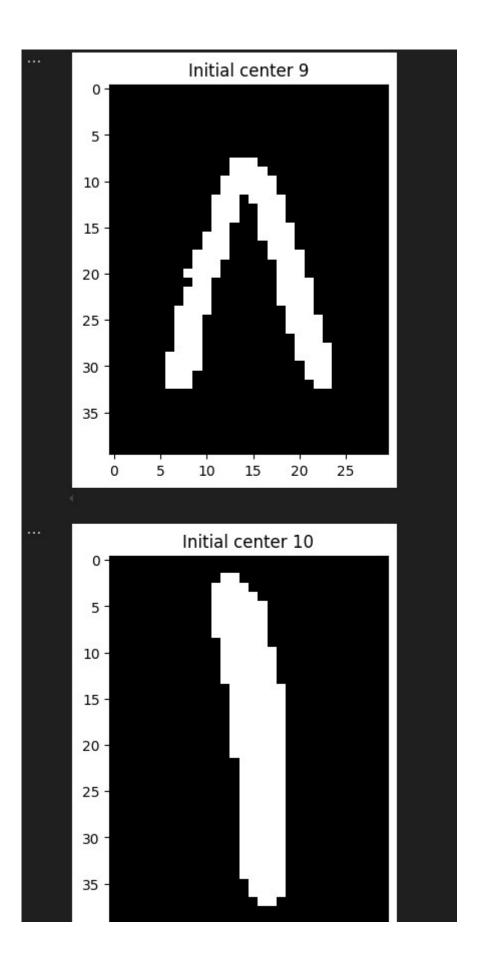


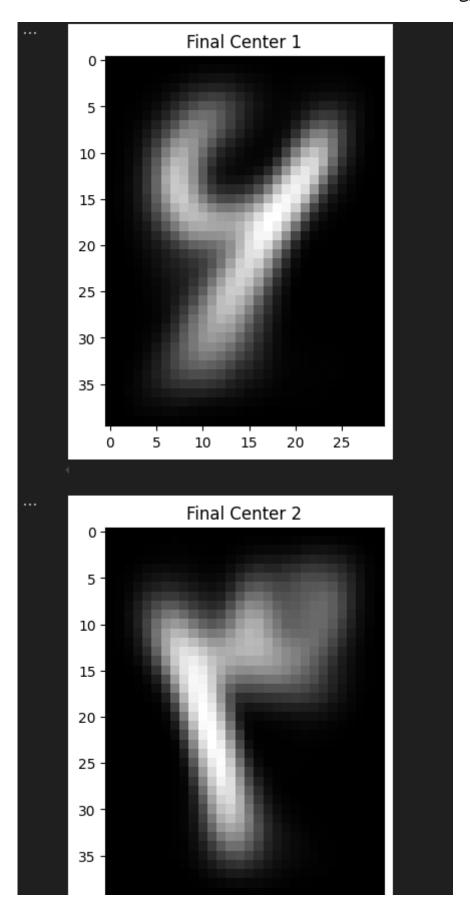


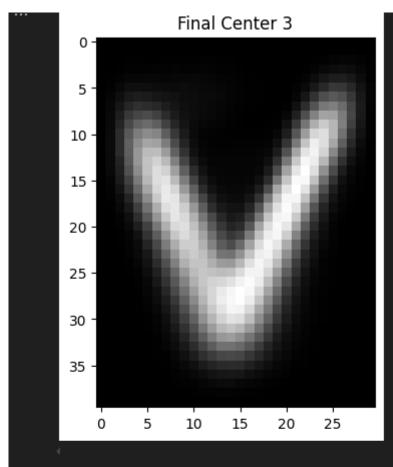


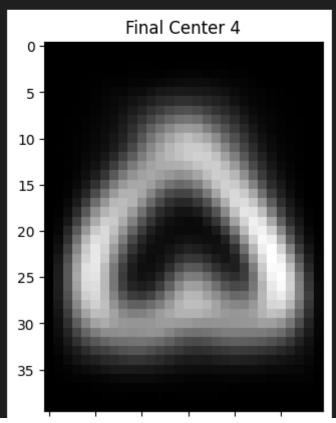


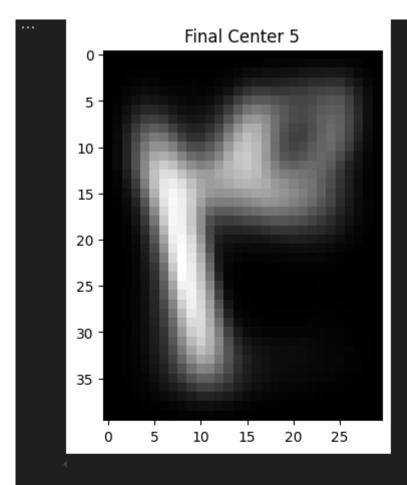


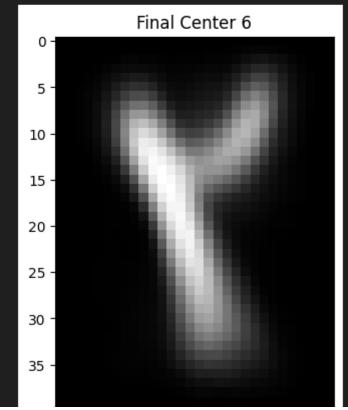


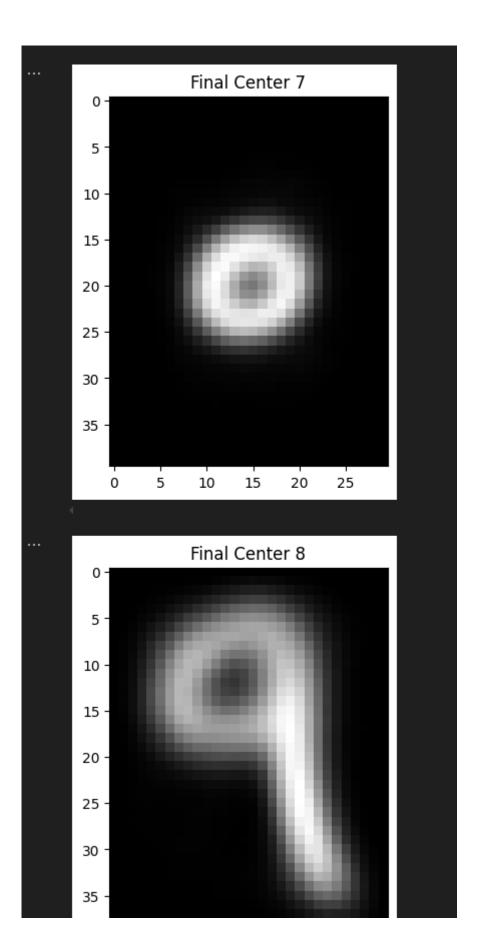


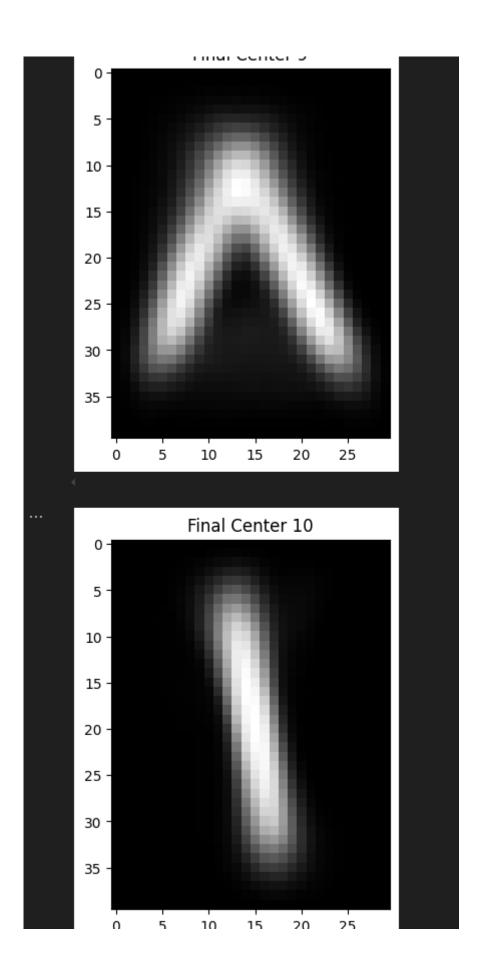


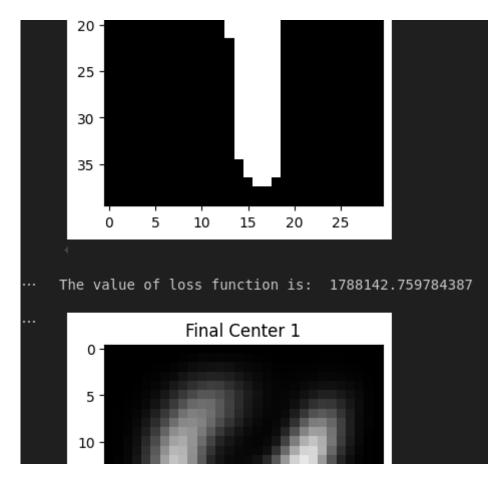


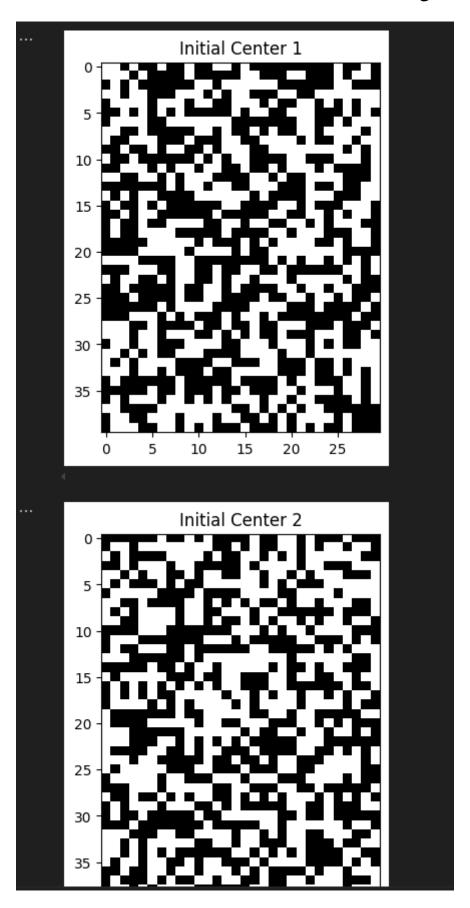


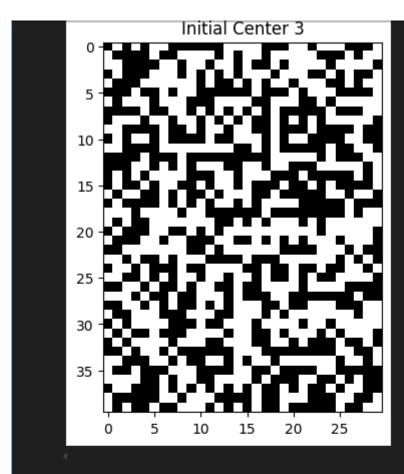


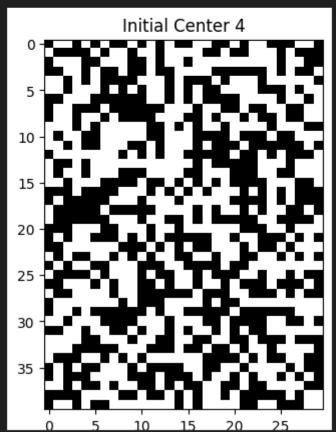


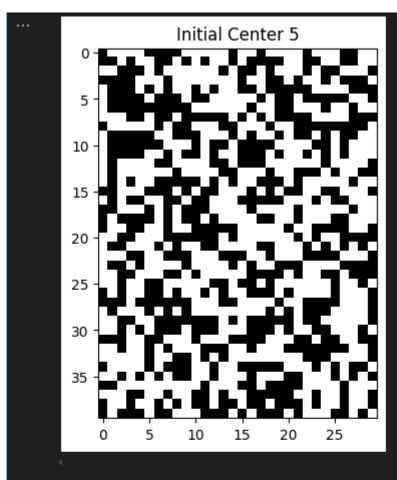


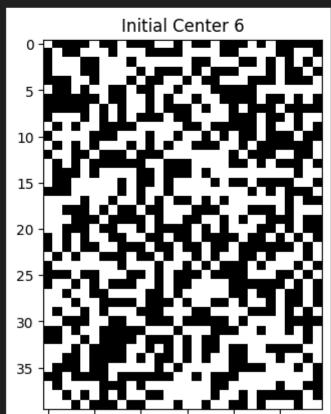


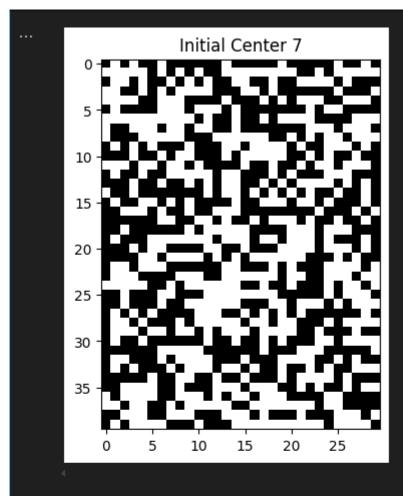


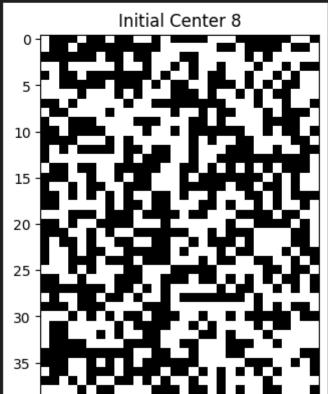


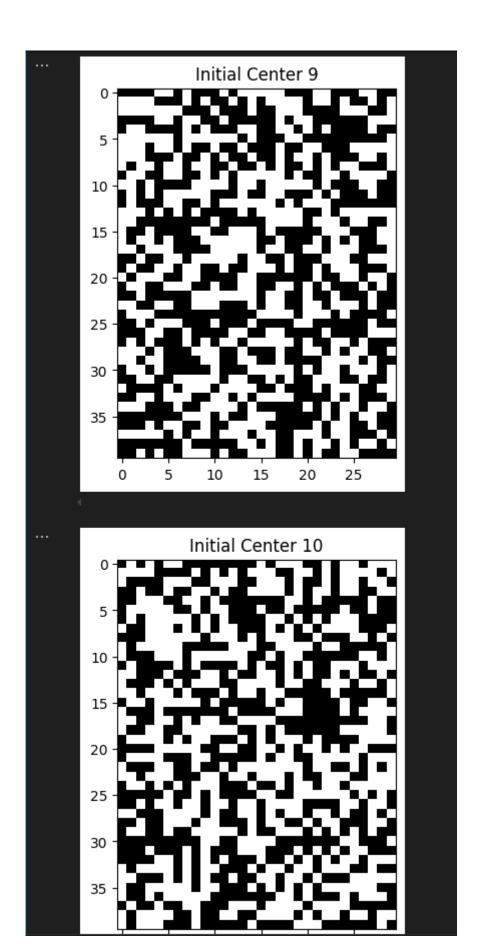


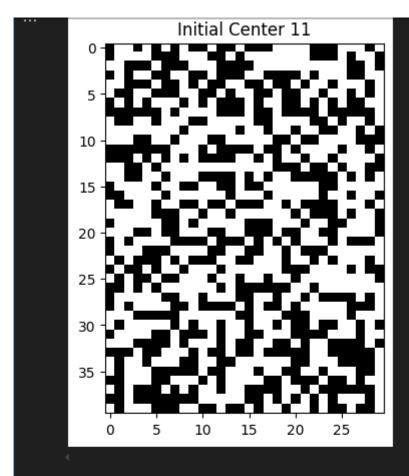


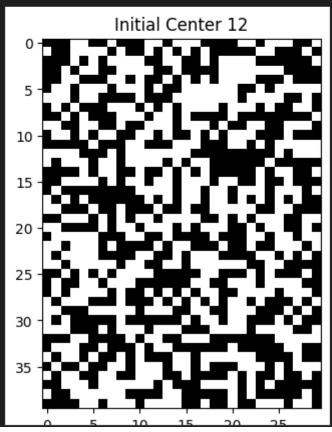


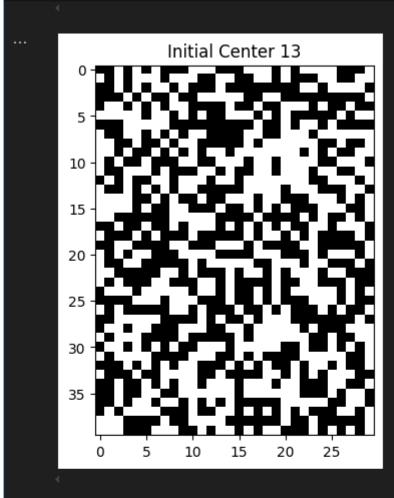


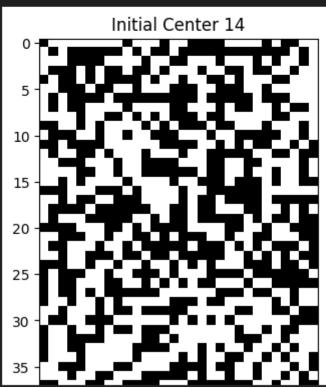


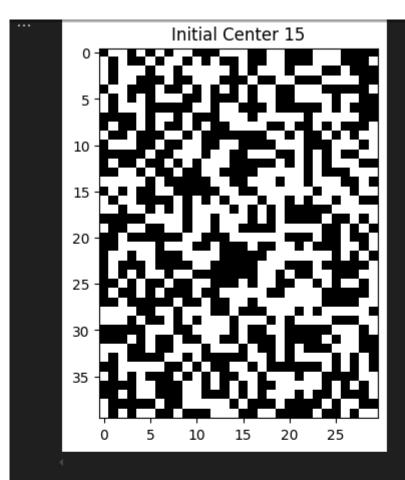


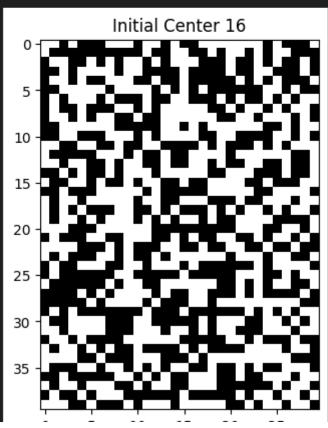


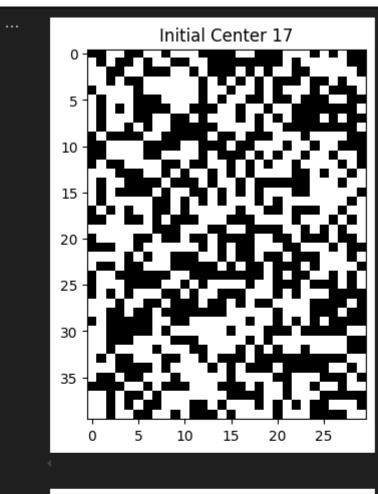


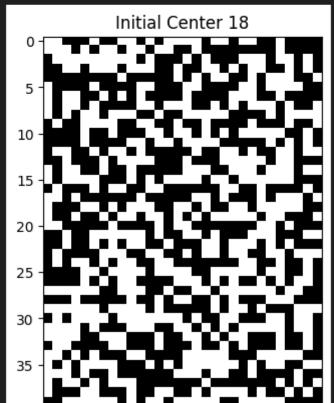


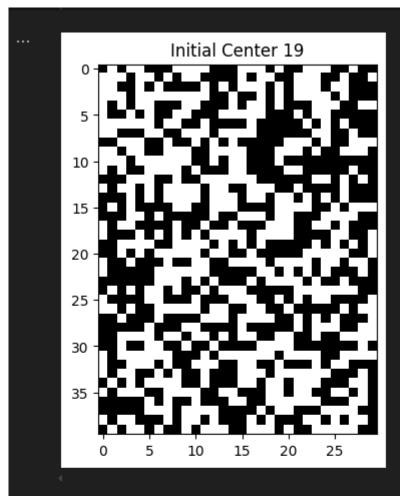


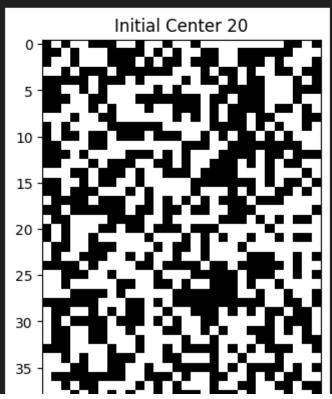


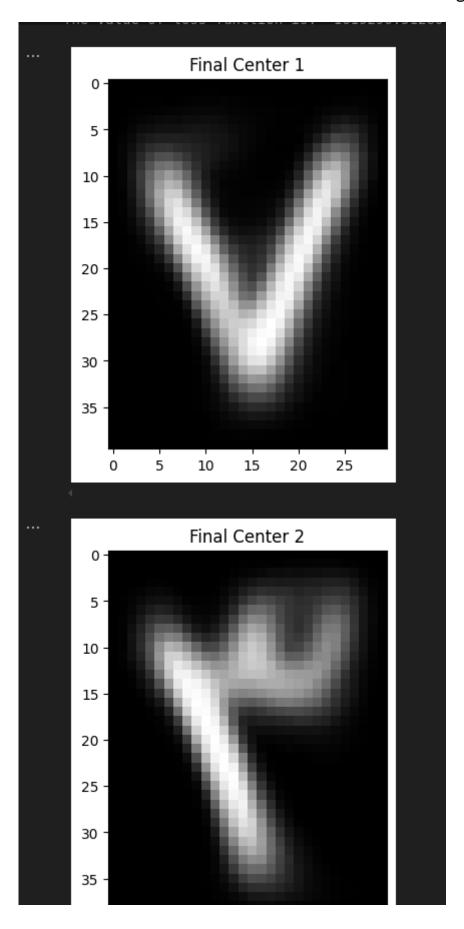


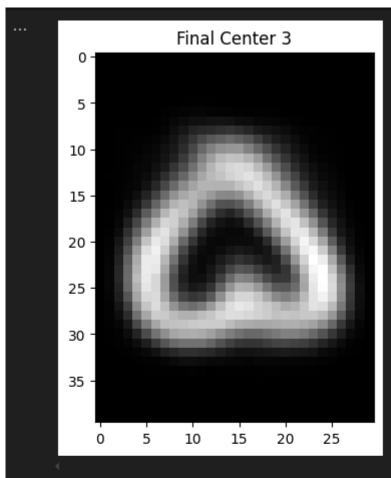


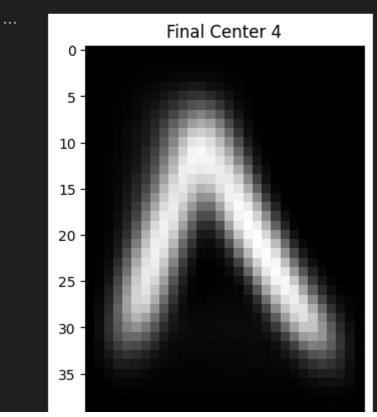


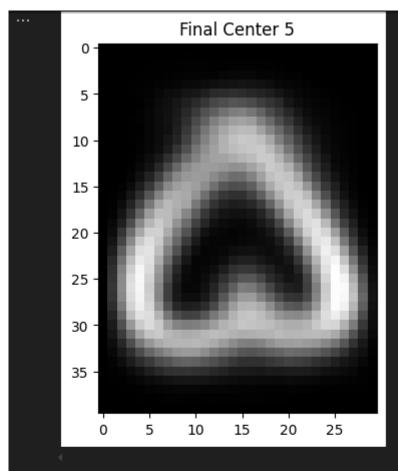


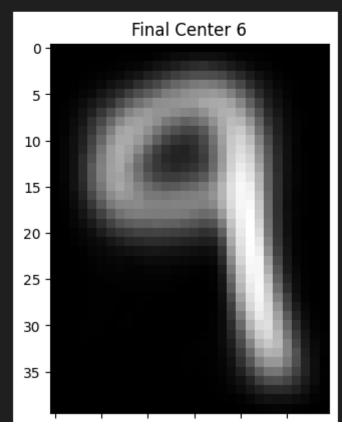


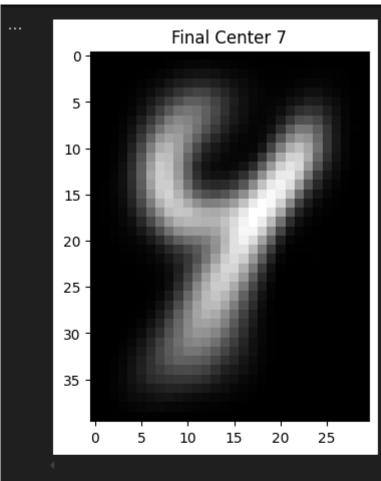


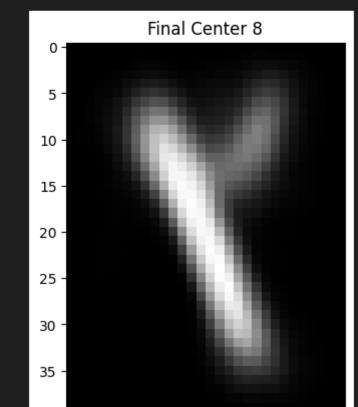


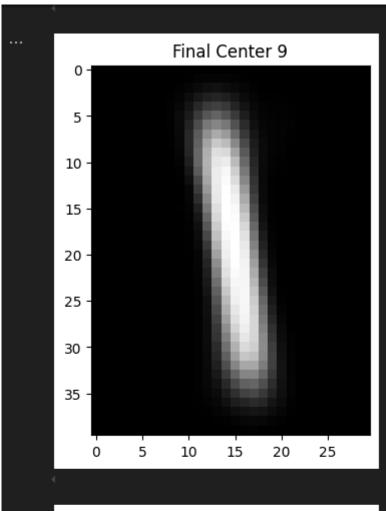


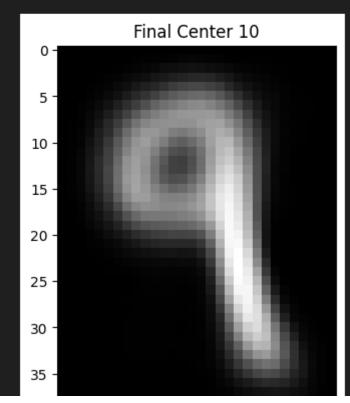


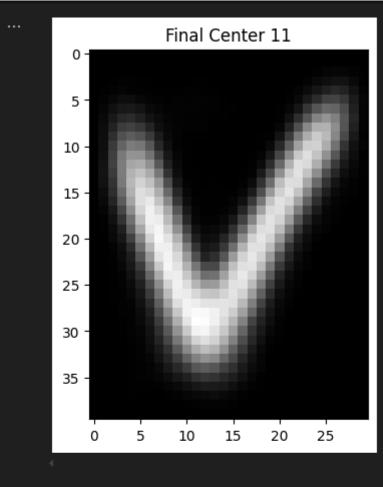


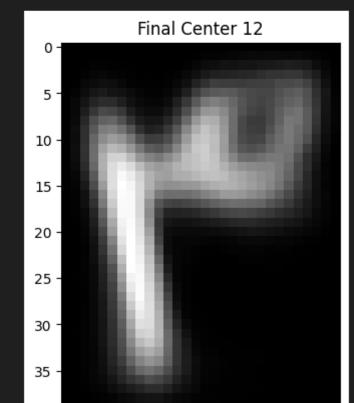


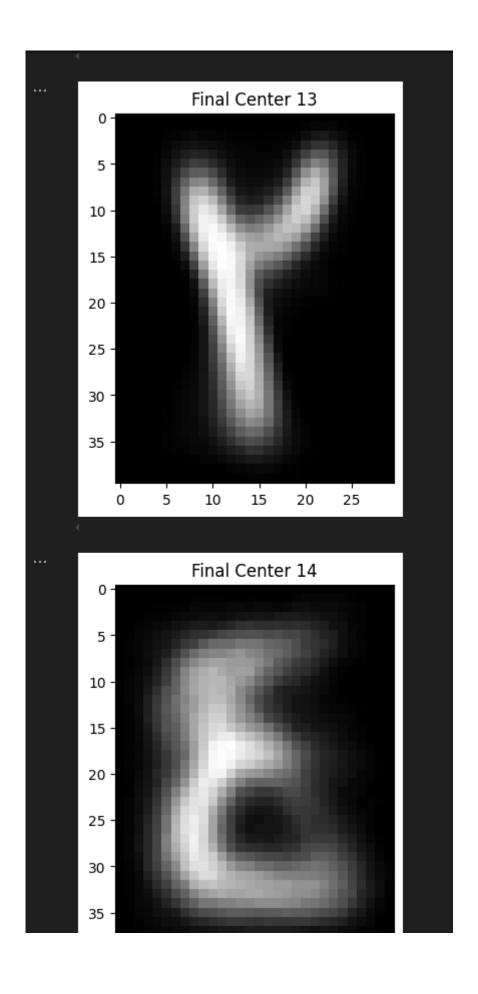


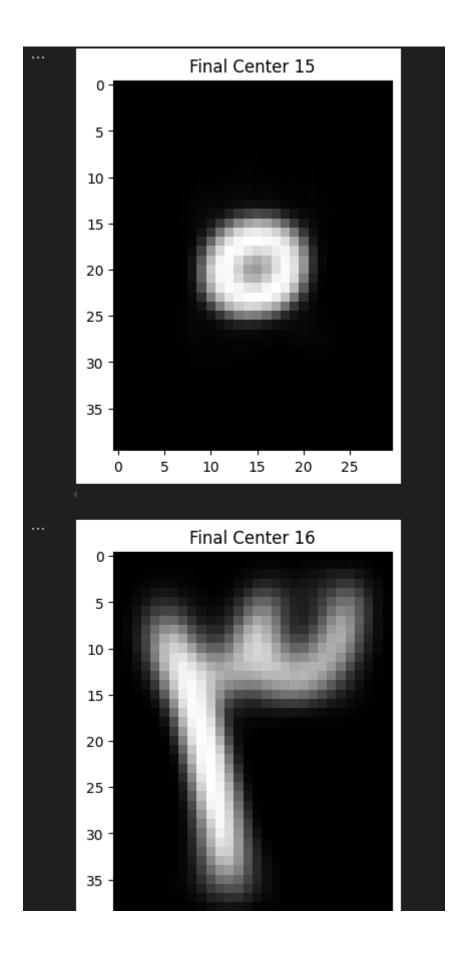


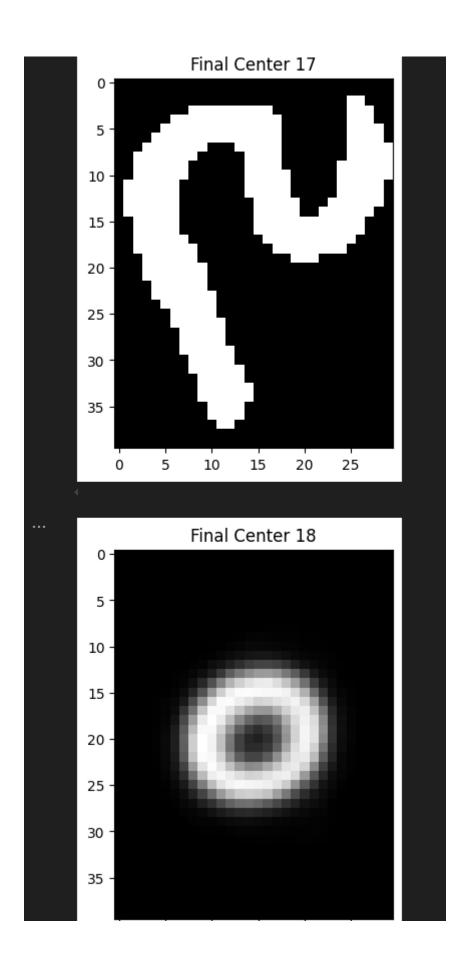


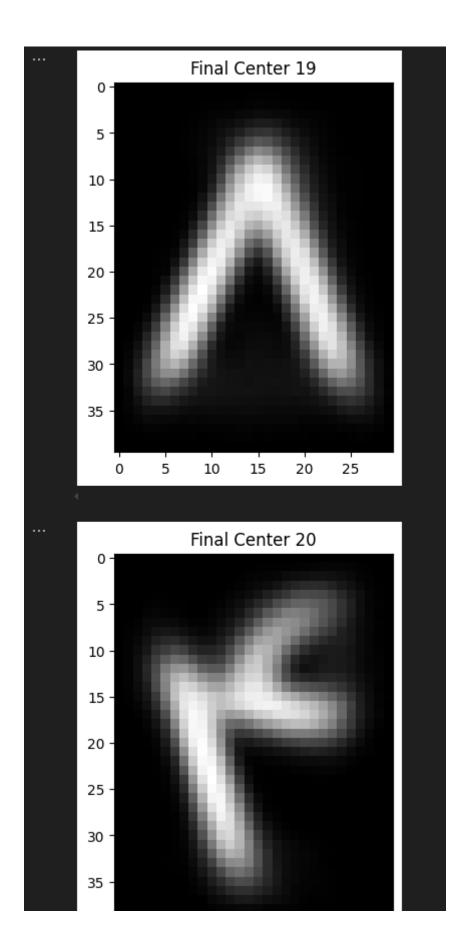


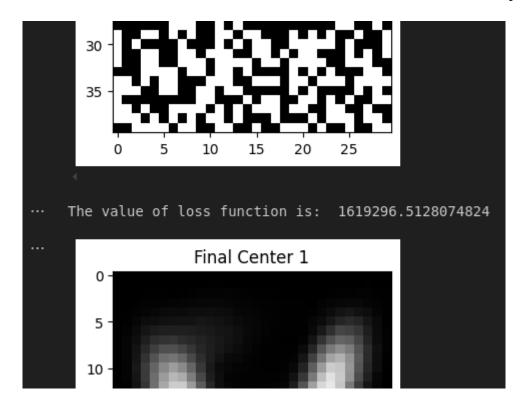


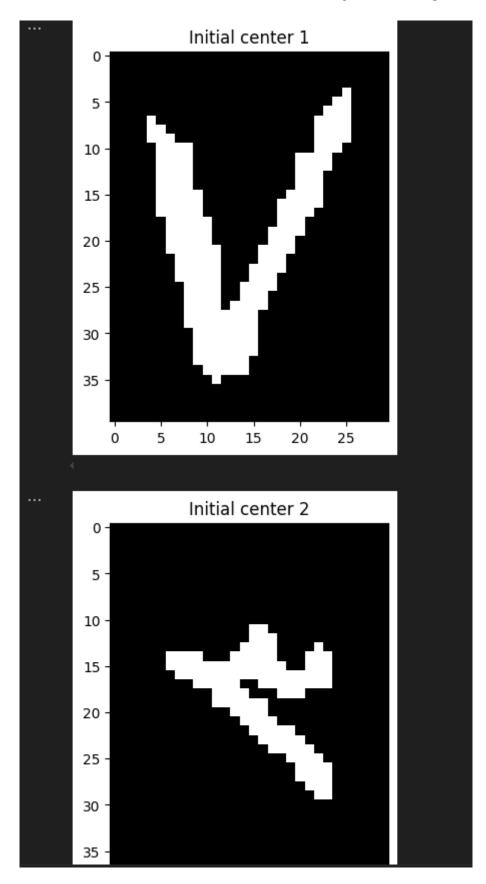


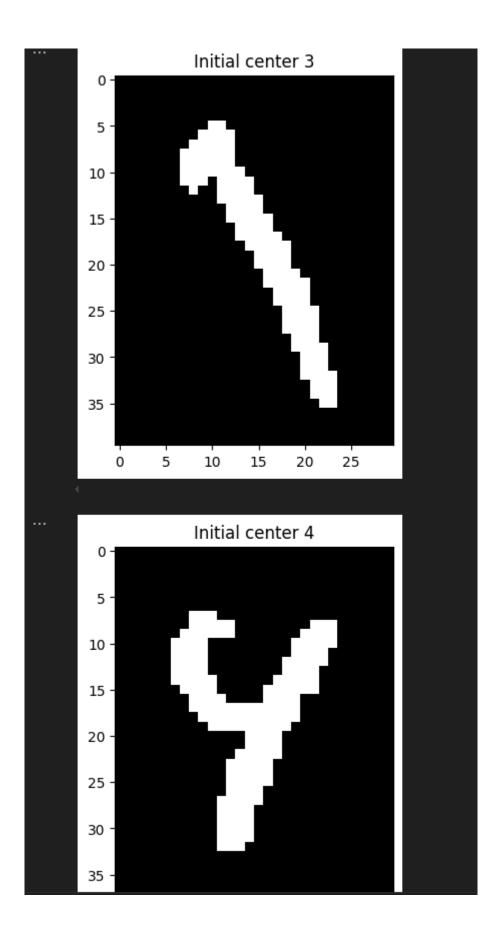


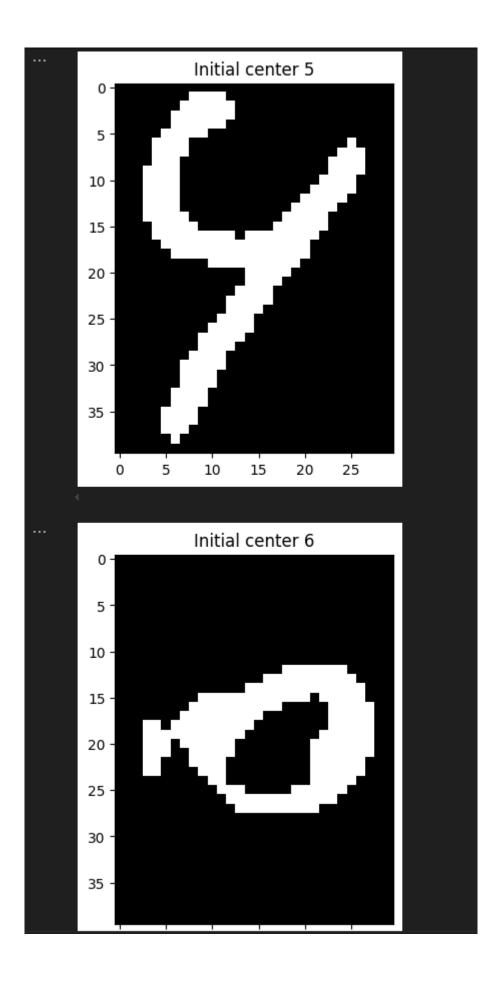


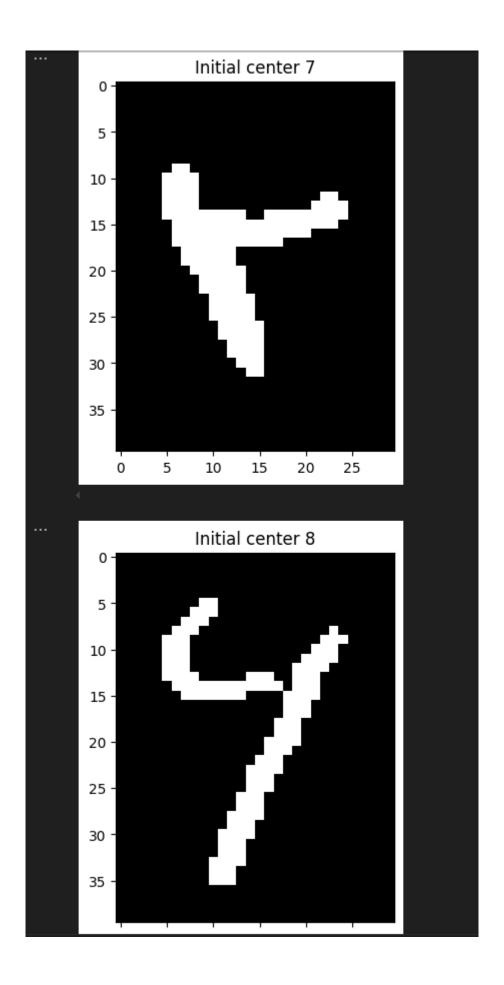


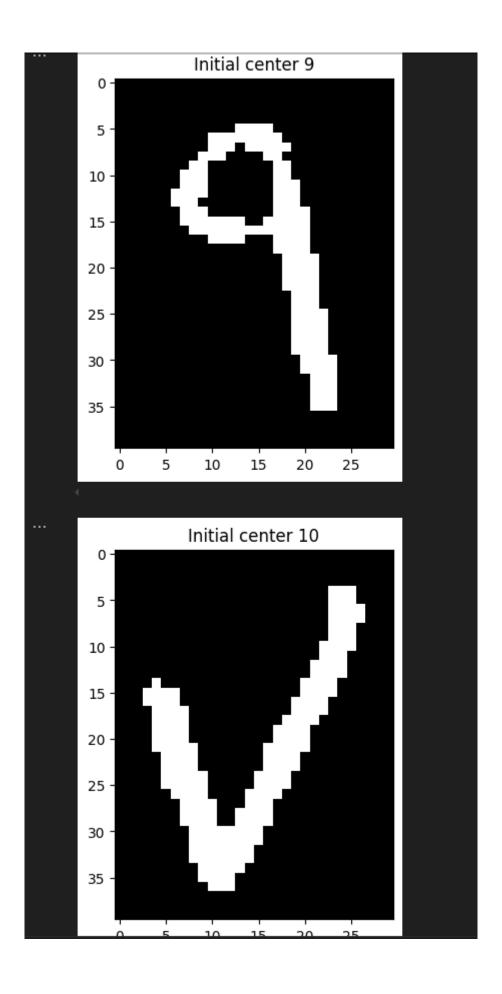


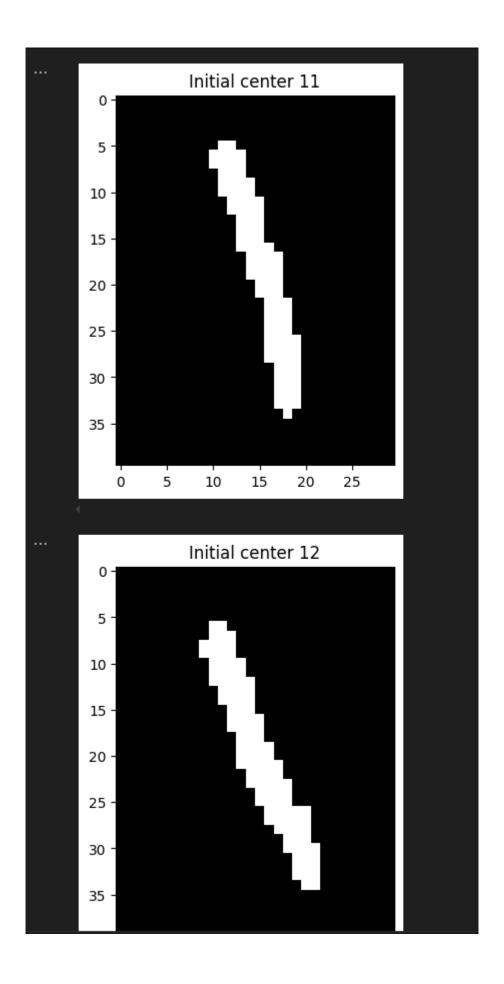


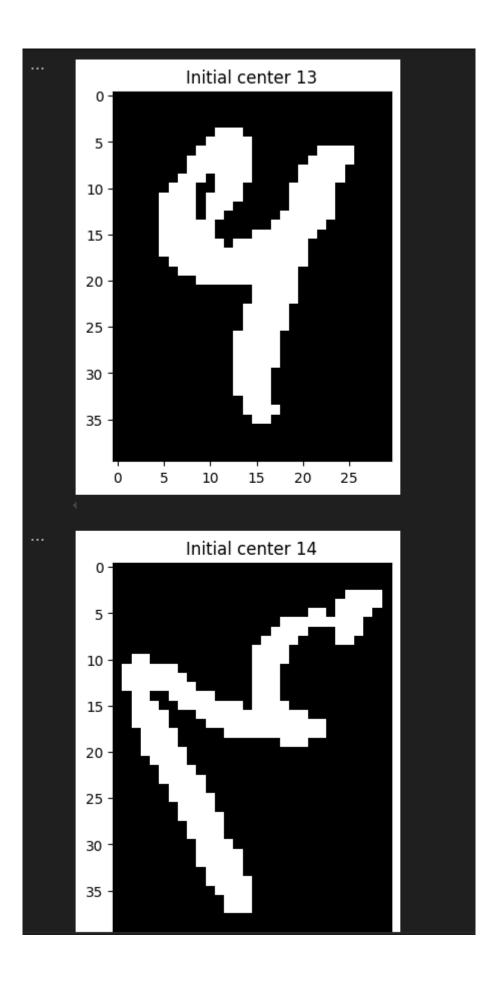


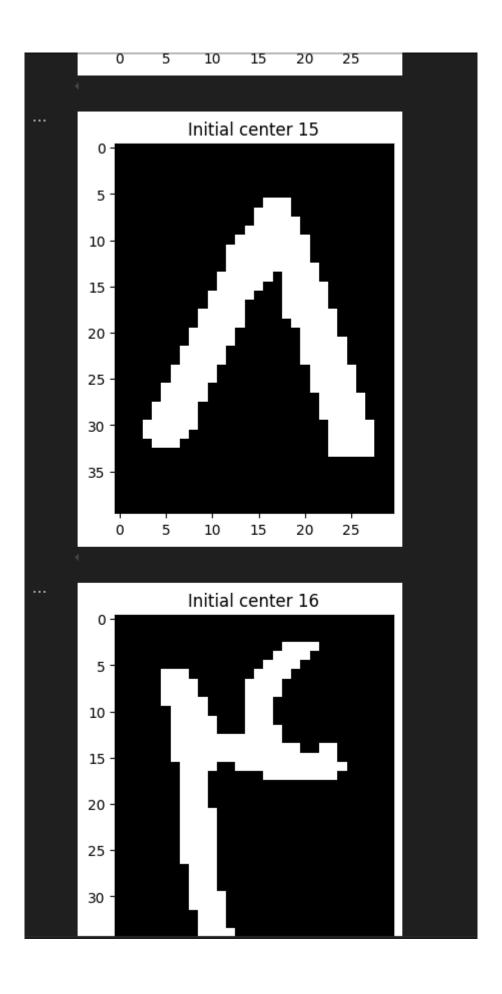


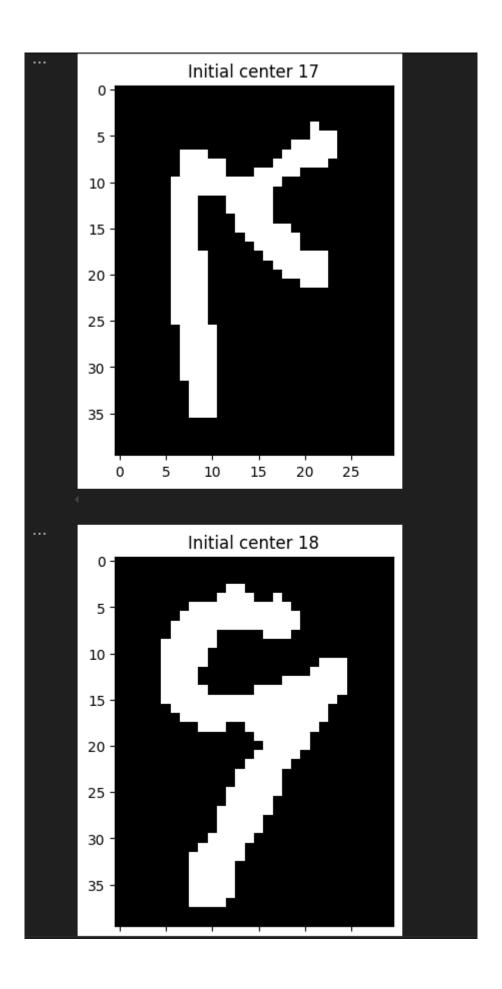


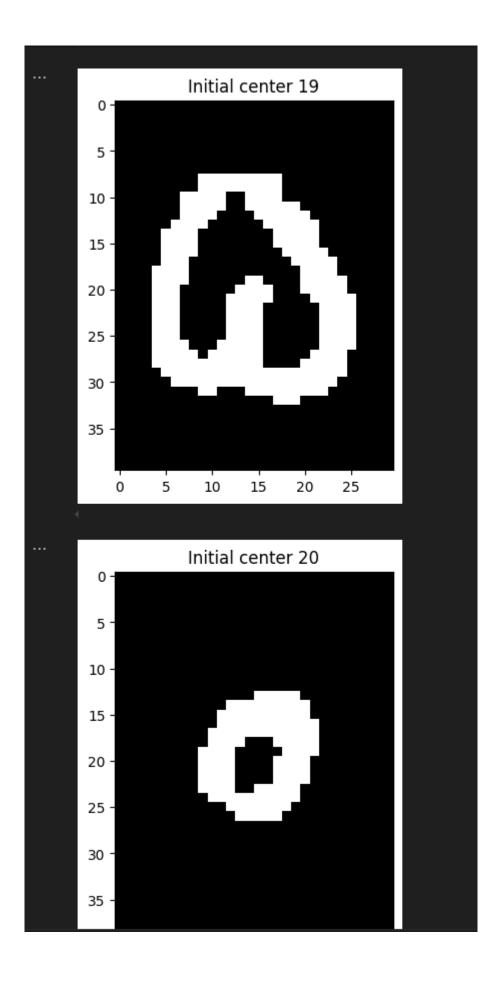


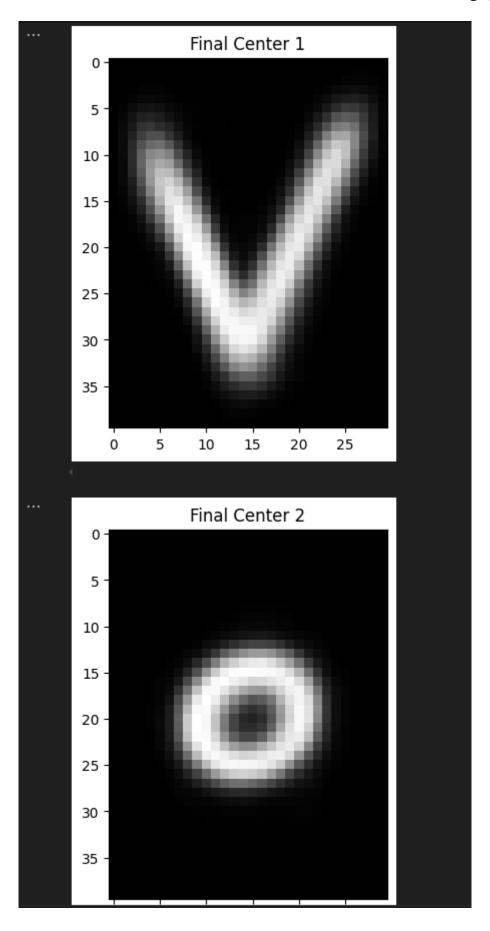


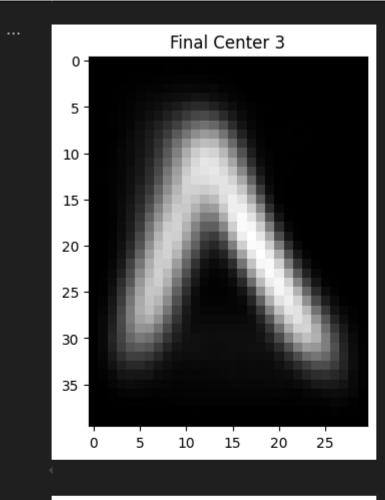


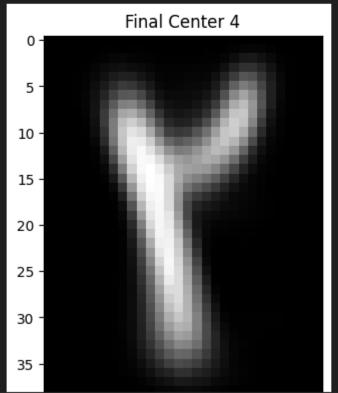


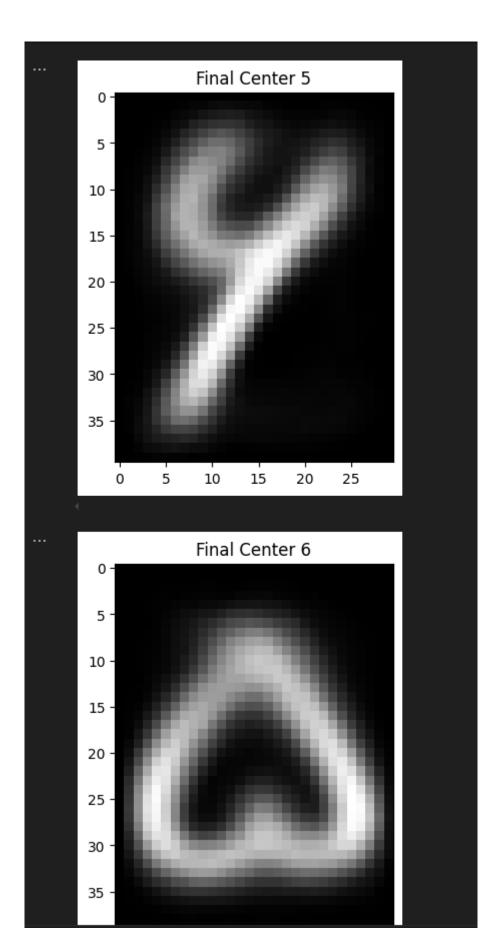


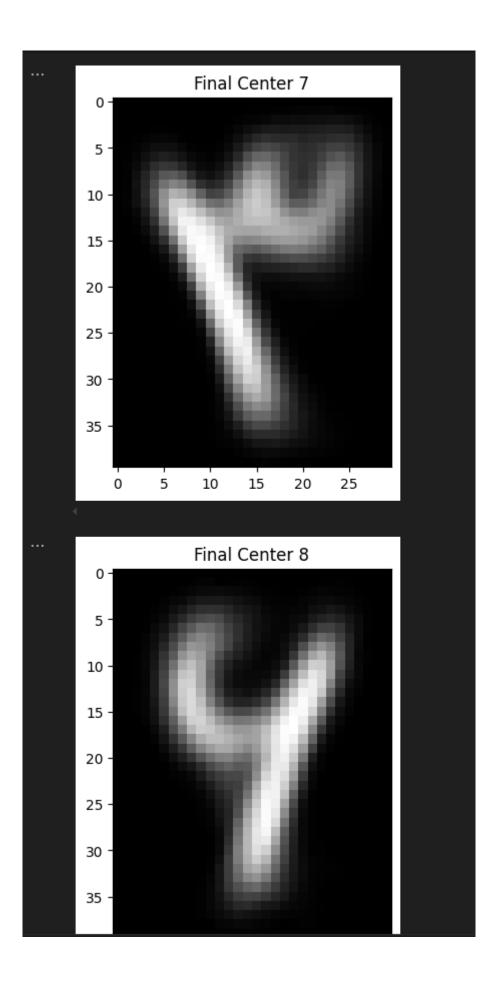


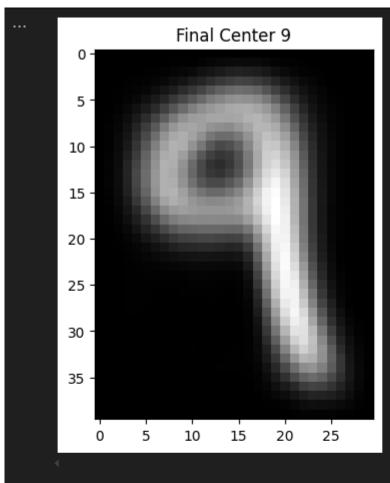


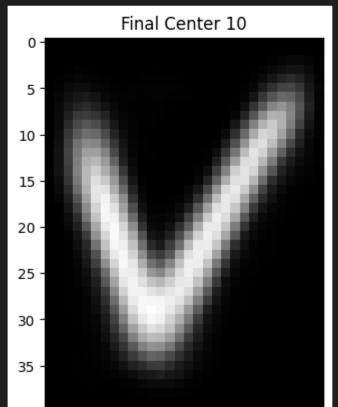


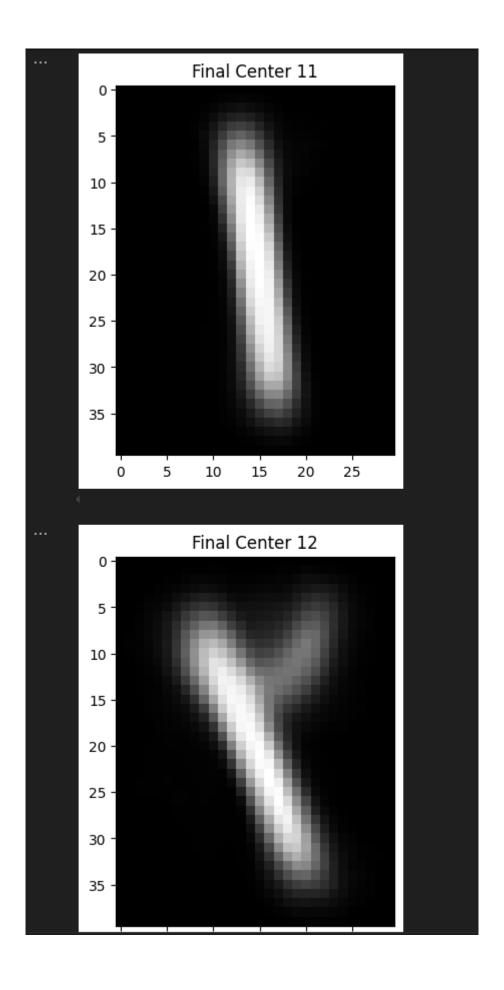


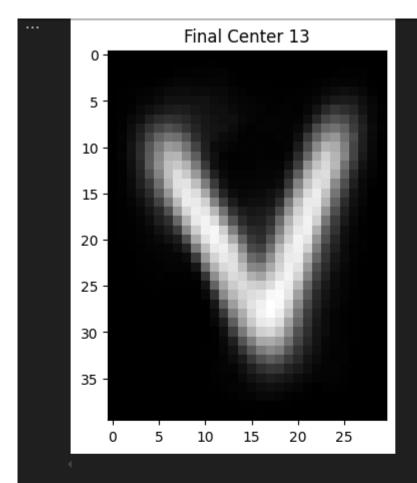


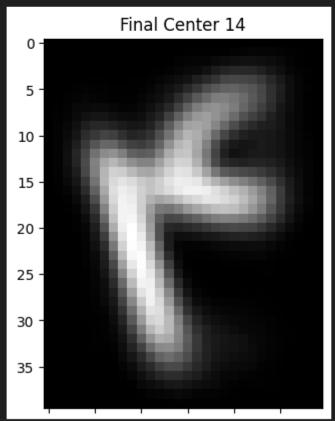


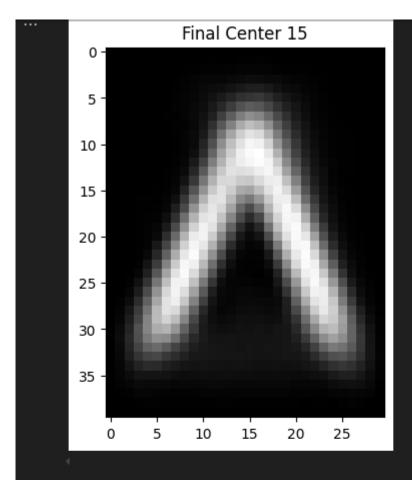


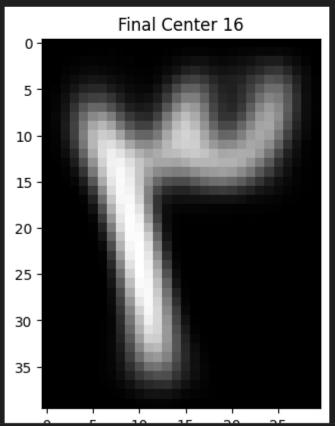


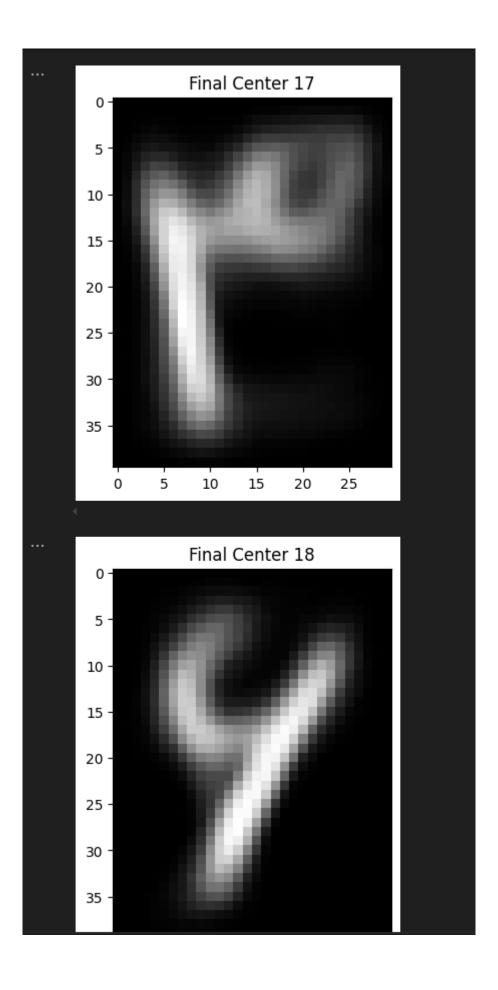


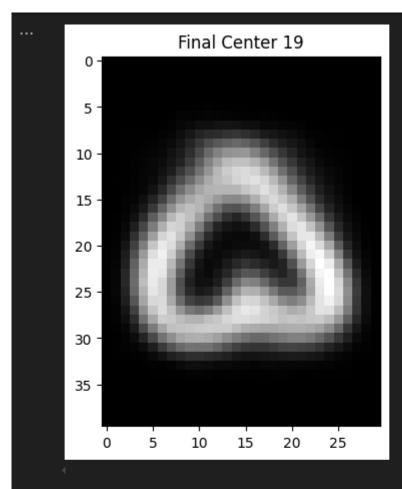


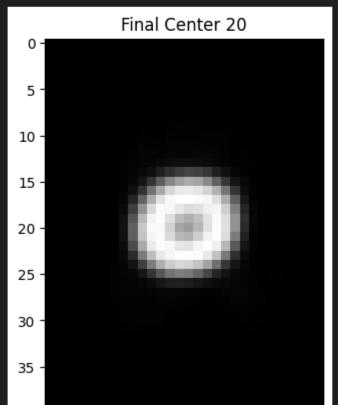


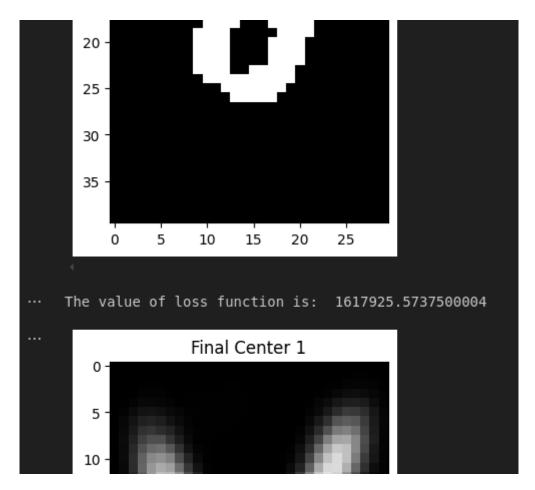


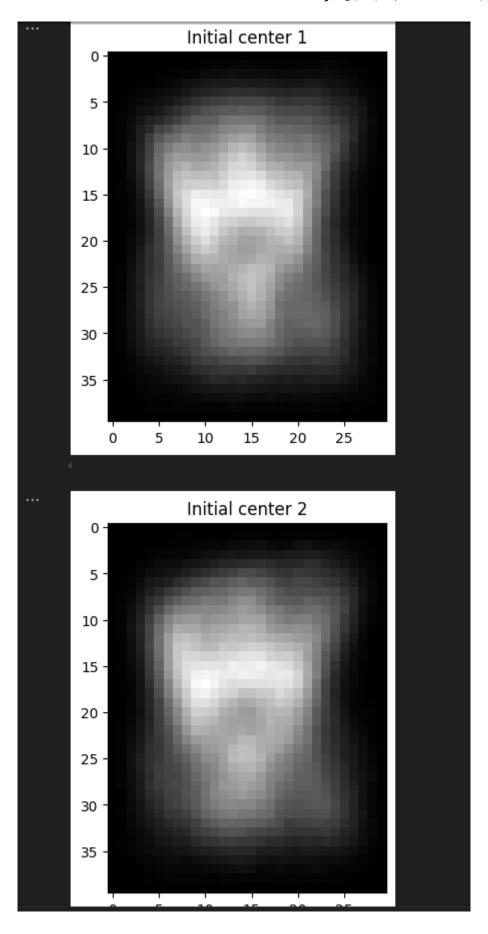


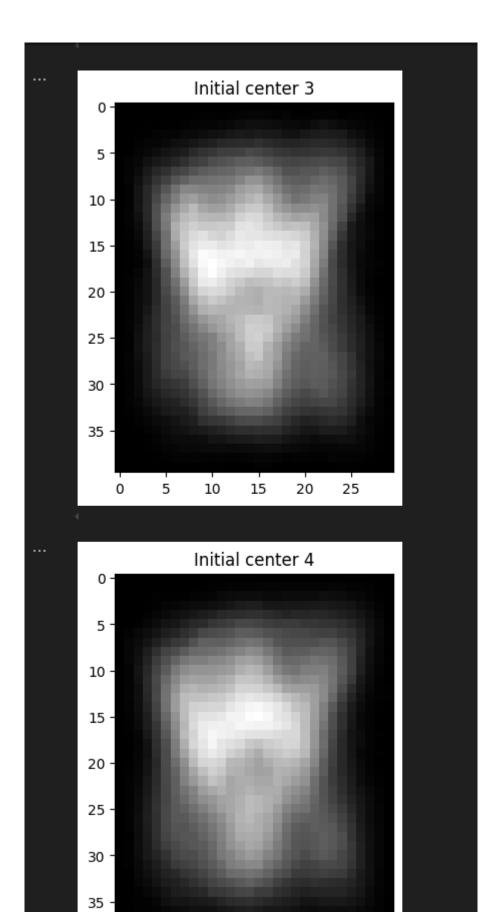


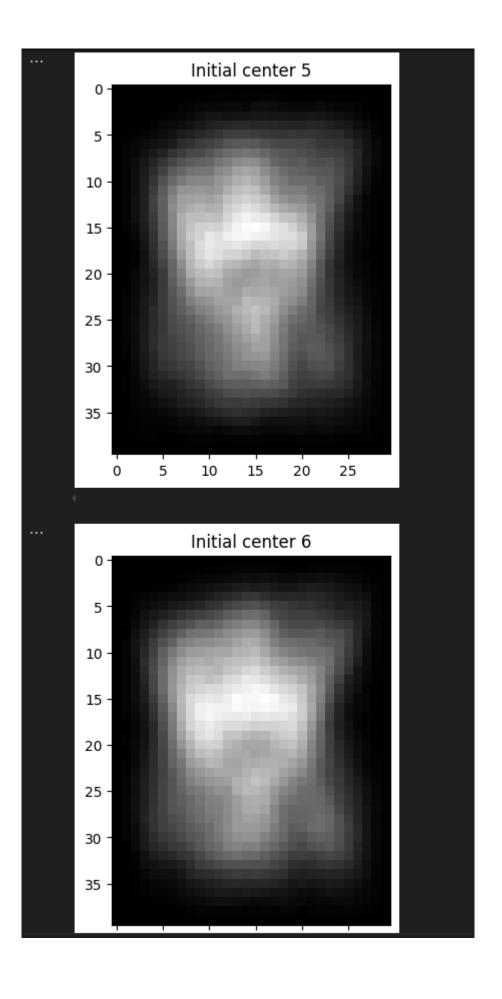


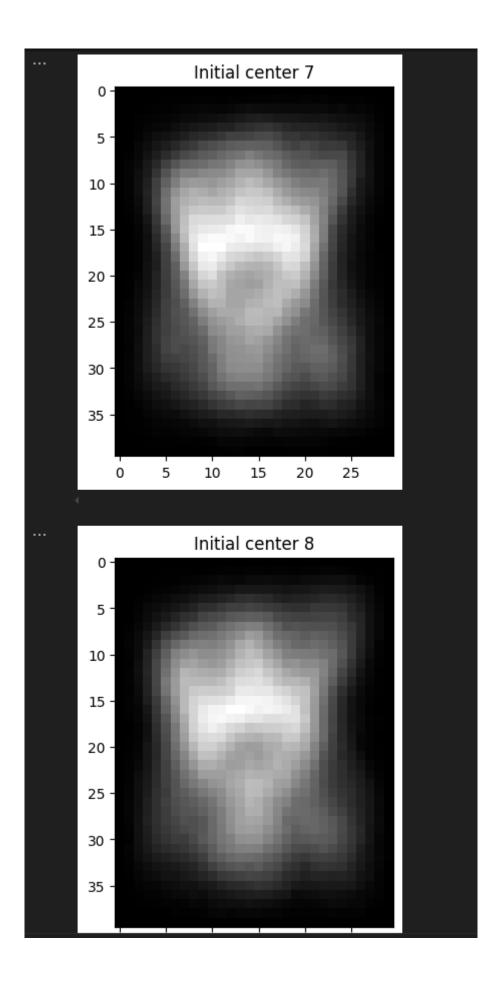


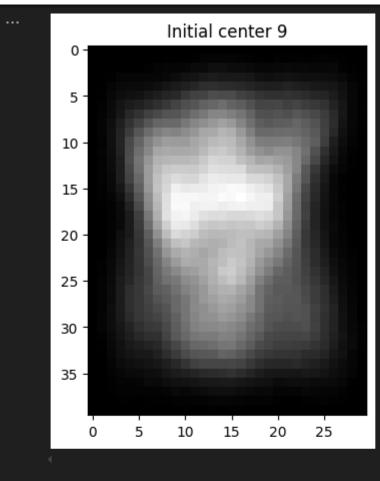


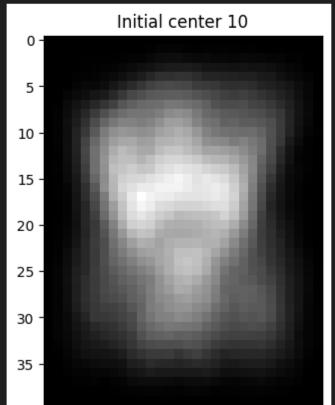


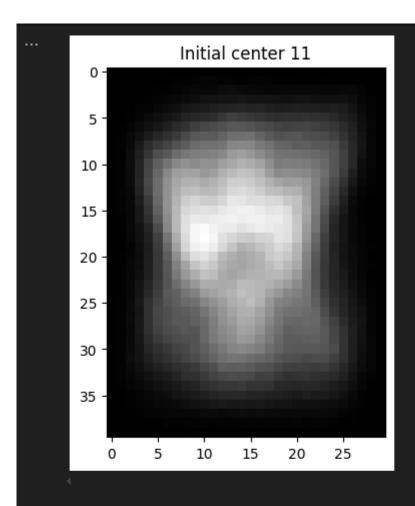


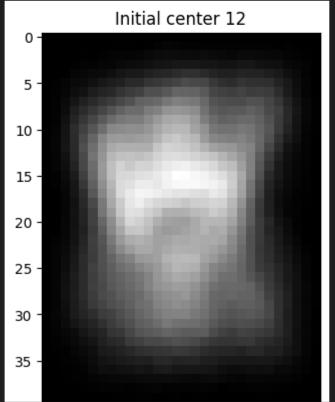


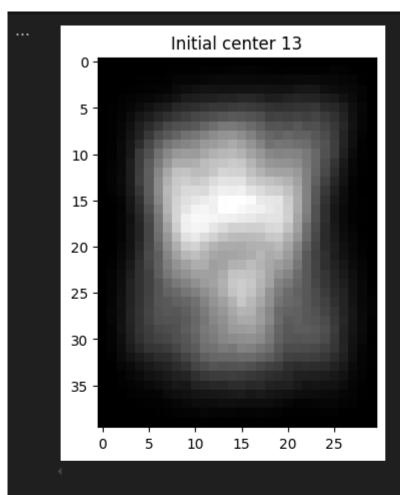


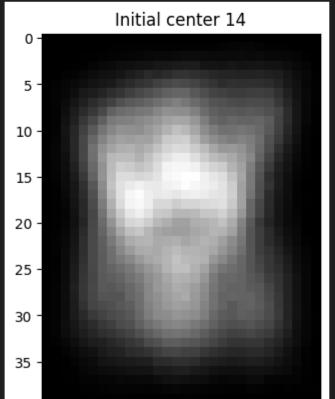


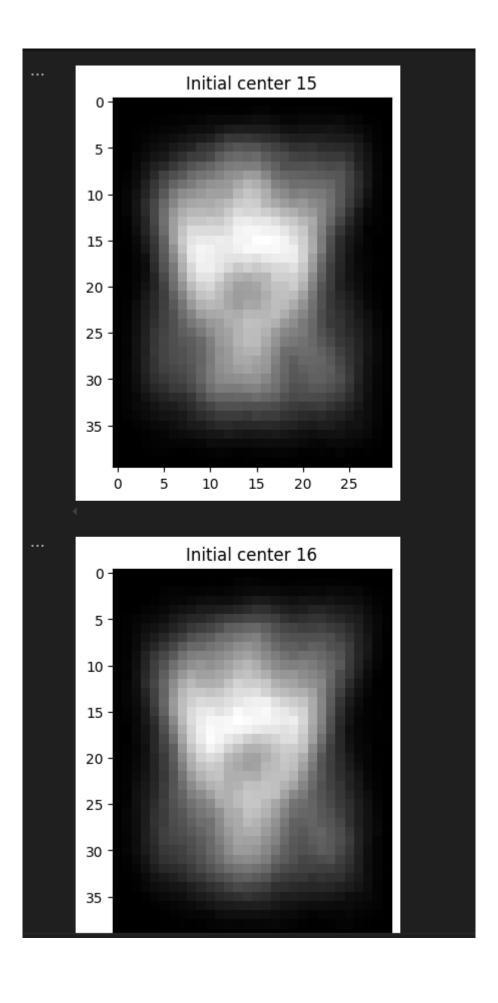


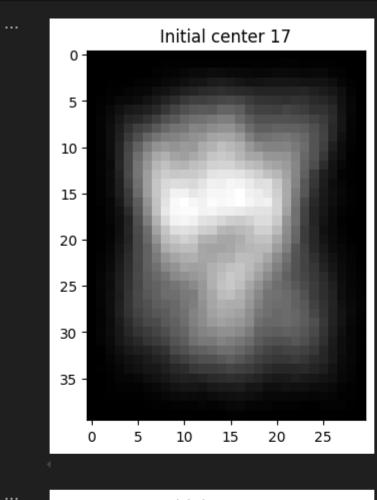


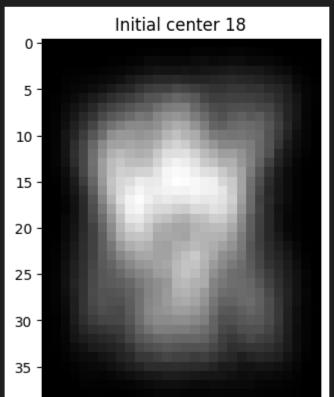


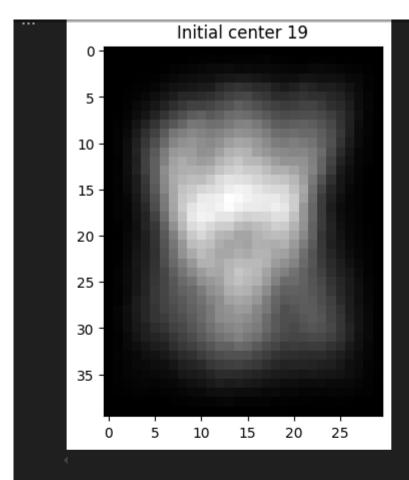


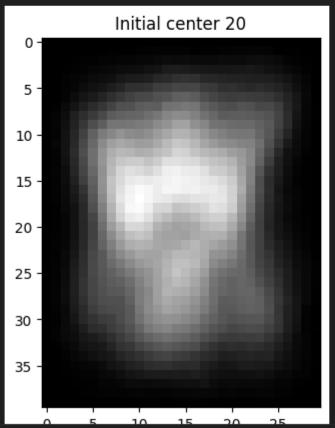


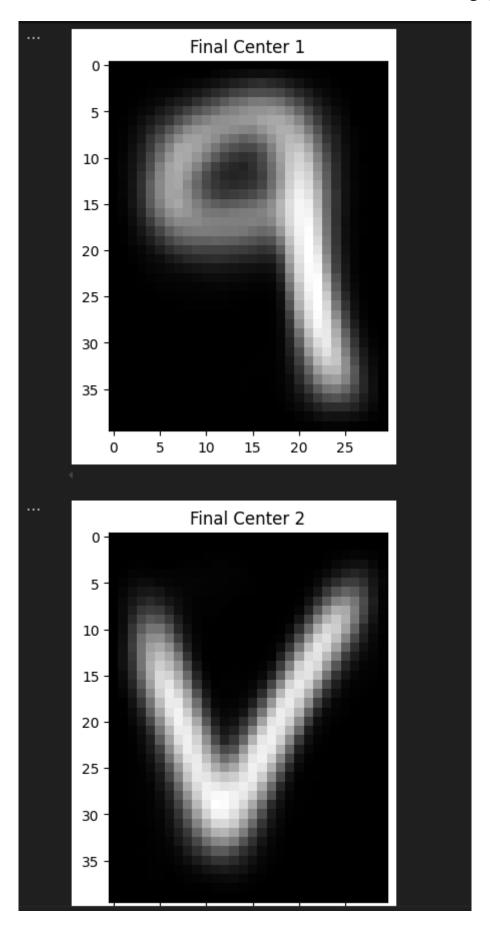


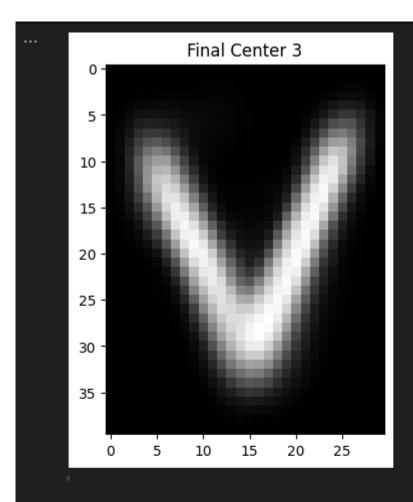


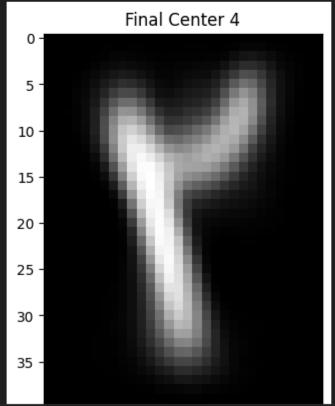


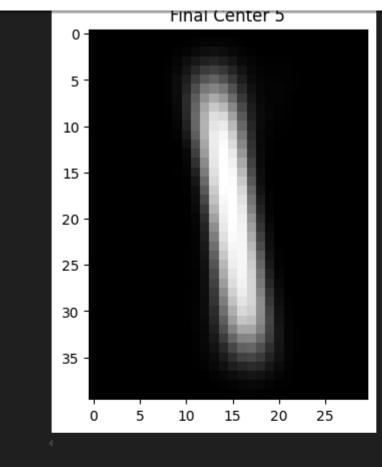


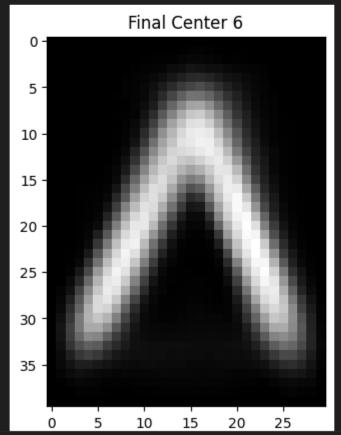


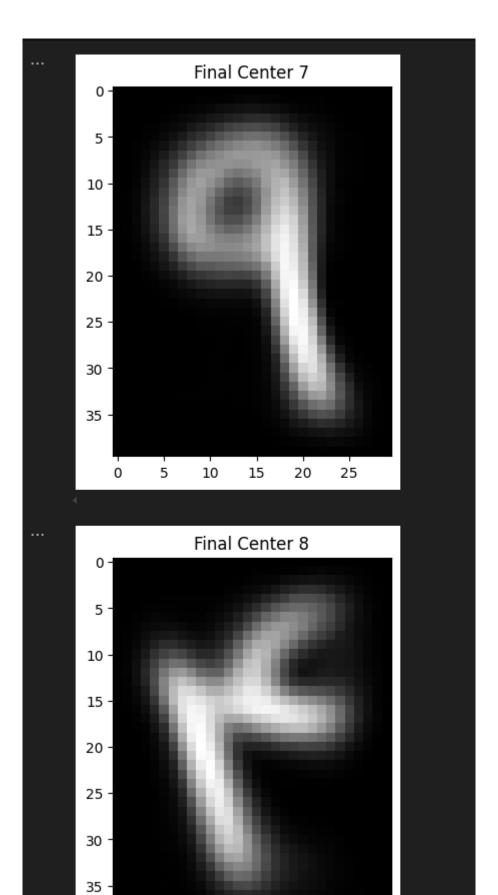


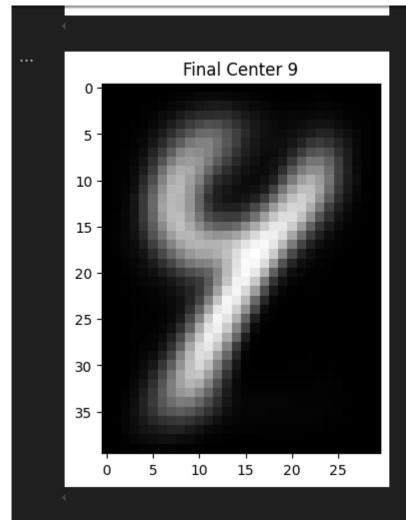


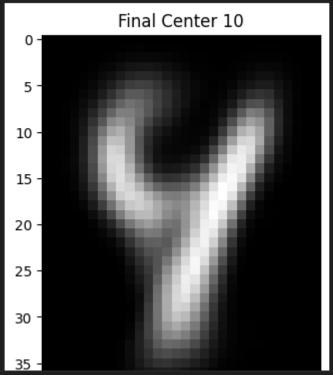


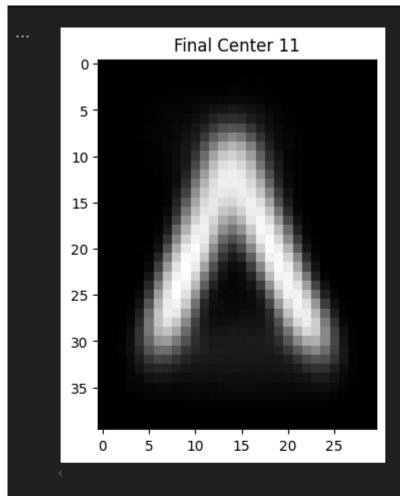


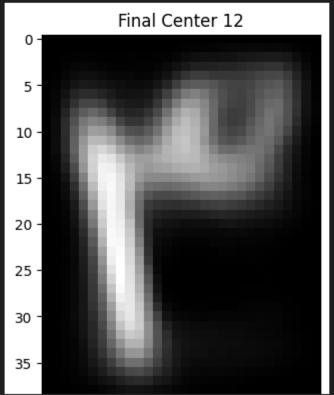


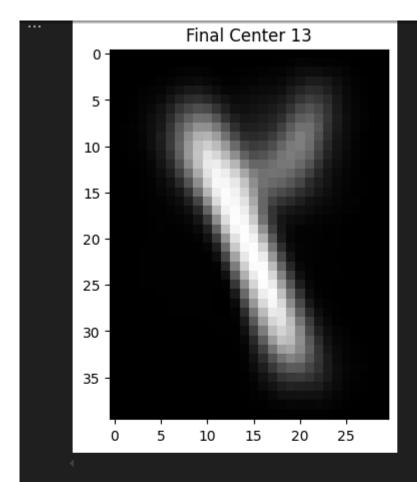


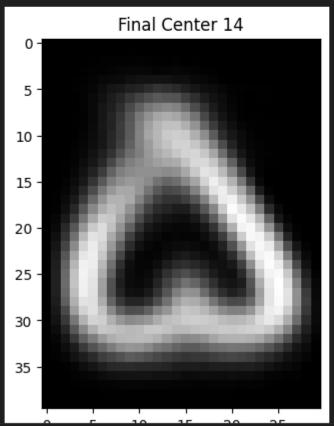


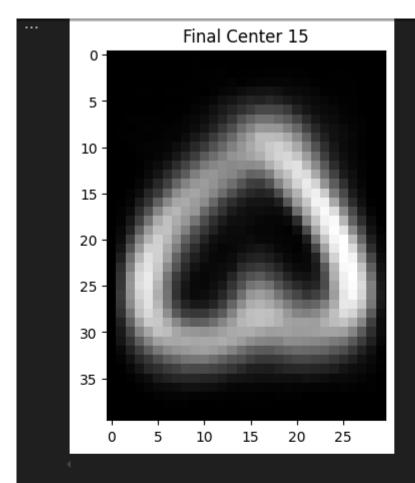


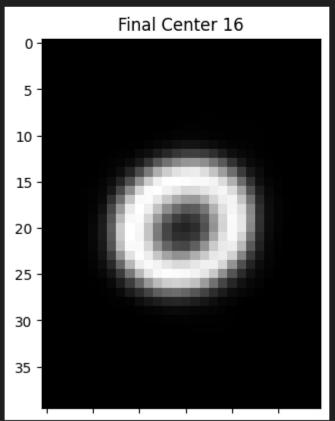


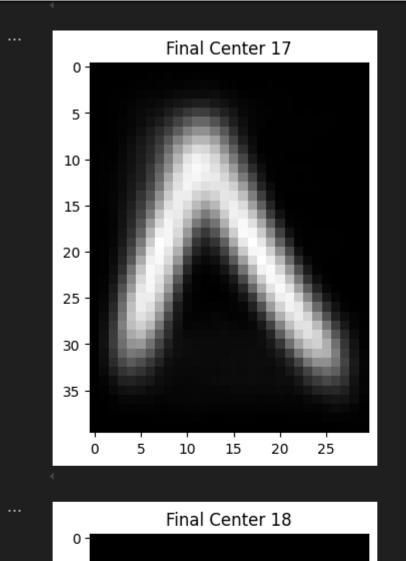


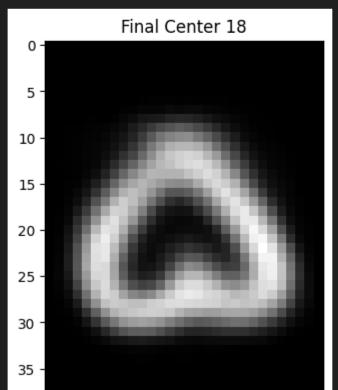


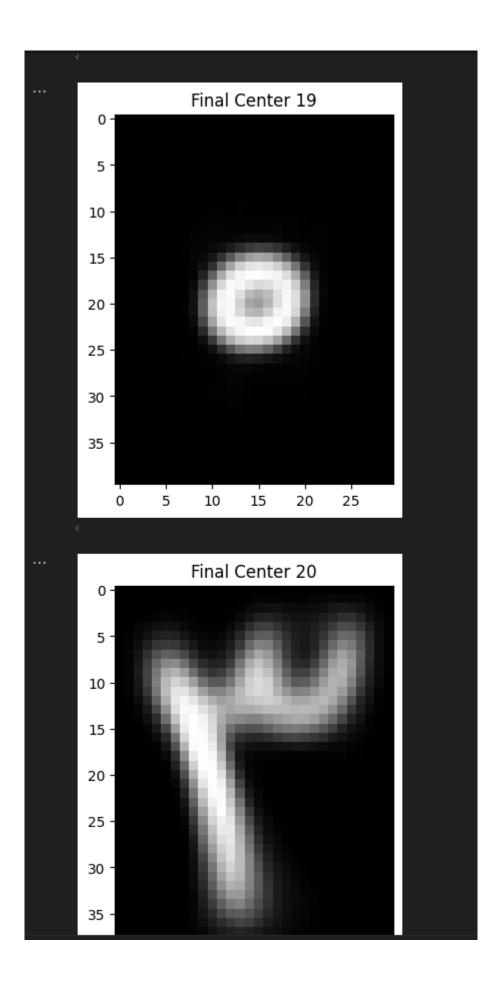


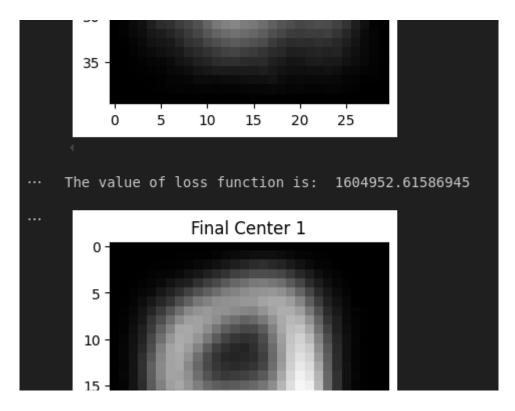












تحليل

از این دو قسمت نتیجه میگیریم که انتخاب مناسب شاخص اولیه میتواند منجر به به دست آوردن شاخص های نهایی مناسب تر و دقیق تر و خوشه بندی بهتر شود ولی تفاوت زیادی در میزان تابع هزینه ایجاد نمی کند اگر تعداد خوشه ها را زیاد کنیم میزان تابع هزینه کم می شود که طبیعی است در این مورد برای بعضی از اعداد چند خوشه در نظر گرفته شود

در کذ برای خواندن داده ها از توابعی کتابخانه پانداس و برای کار با آرایه ها از کتابخانه نامپای هم چنین از الگوریتم خوشه بندی کتابخانه سایکیت لرن استفاده شده است

Data value	a.I	a.II	a.III	a.IV	b.I	b.II	b.III
0	0	0	0	0	0	0	0
1	1	1	1	1	1	1	1
2	2	2	2	2	2	2	2
3	3	3	3	3	3	2	2
4	4	3	4	4	4	3	4
4	4	8	8	4	4	8	3
5	5	5	5	5	5	5	5
5	5	5	5	5	5	5	5
6	2	2	3	6	6	6	6
6	9	9	9	9	9	9	9
7	7	7	7	7	7	7	7
8	8	8	8	8	8	8	8
9	9	9	9	9	9	9	9

در کل به نظر میرسد استفاده از شاخص های اولیه که به صورت رندوم ساخته شدند دقت خوبی دارد و دقت آن در بعضی موارد نسبت به استفاده از میانگین و انتخاب رندوم از بین داده ها بهتر است هم چنین در کل در مواردی که تعداد خوشه ها ۲۰ تاست دقت کمی بیشتر است

pandas: read_csv() for reading data from .txt file

matplotlib.pyplot: imshow(), title(), show() for displaying an image

numpy: np.random.randint() , np.random.choice for generating or selecting random items reshape() , np.array_split() for working with arrays

 $sklearn.cluster:\ Kmeans\ ,\ Kmeans.fit()\ ,\ Kmeans.predict()\ ,\ Kmeans.inertia_\ ,Kmeans.cluster_centers_\ for\ clustering$