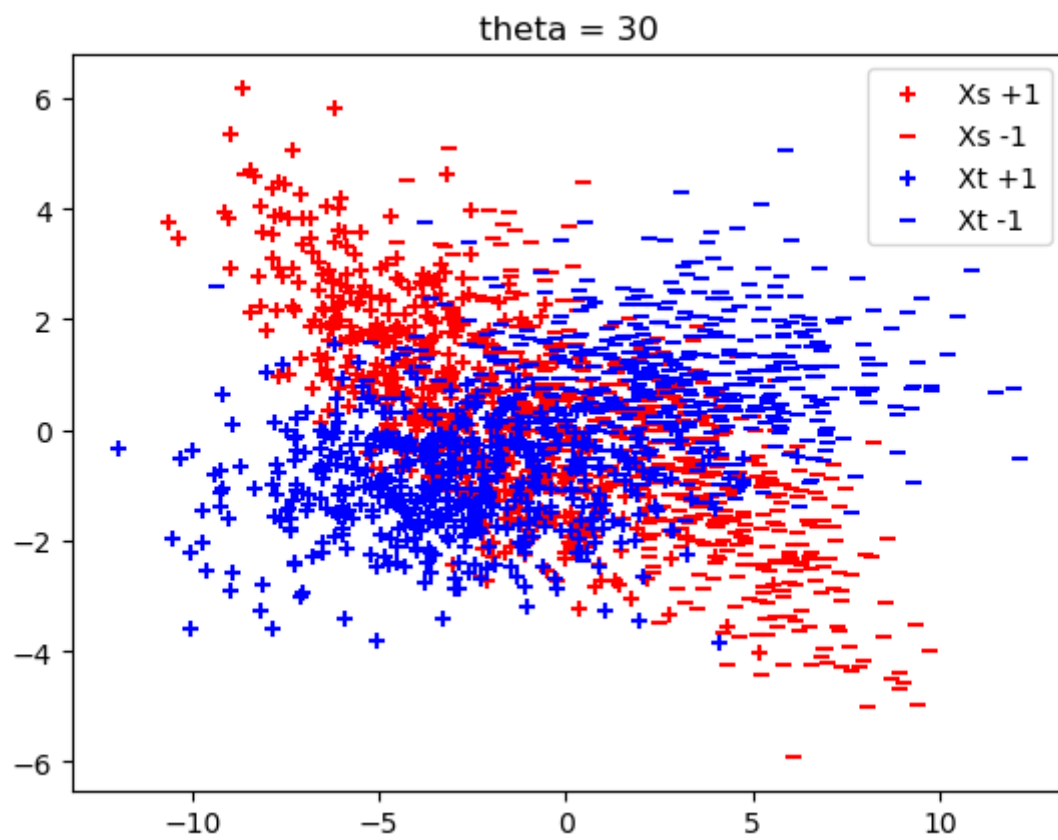
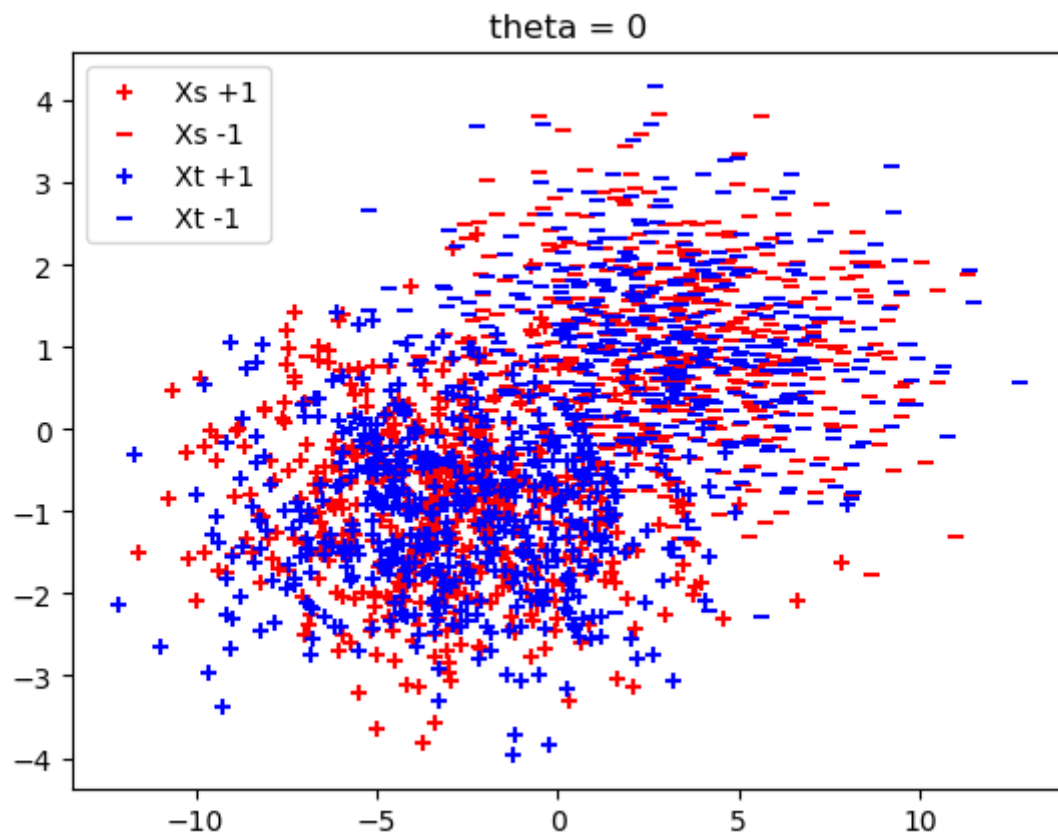
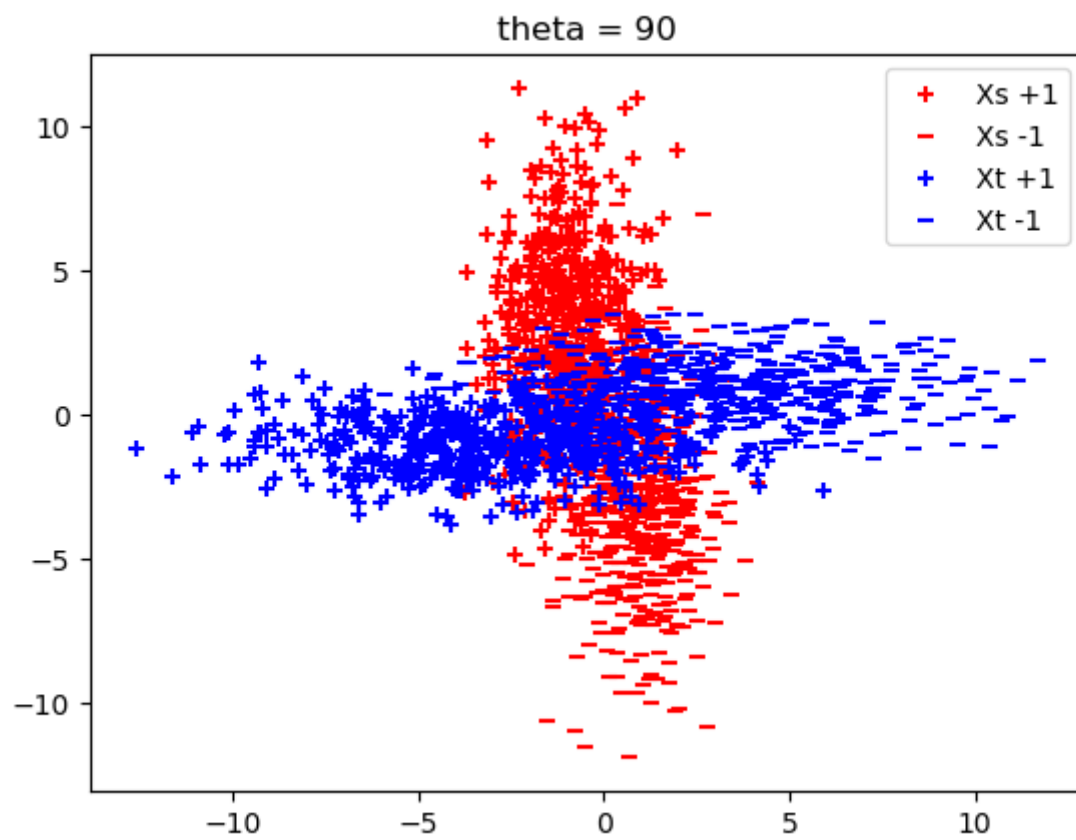
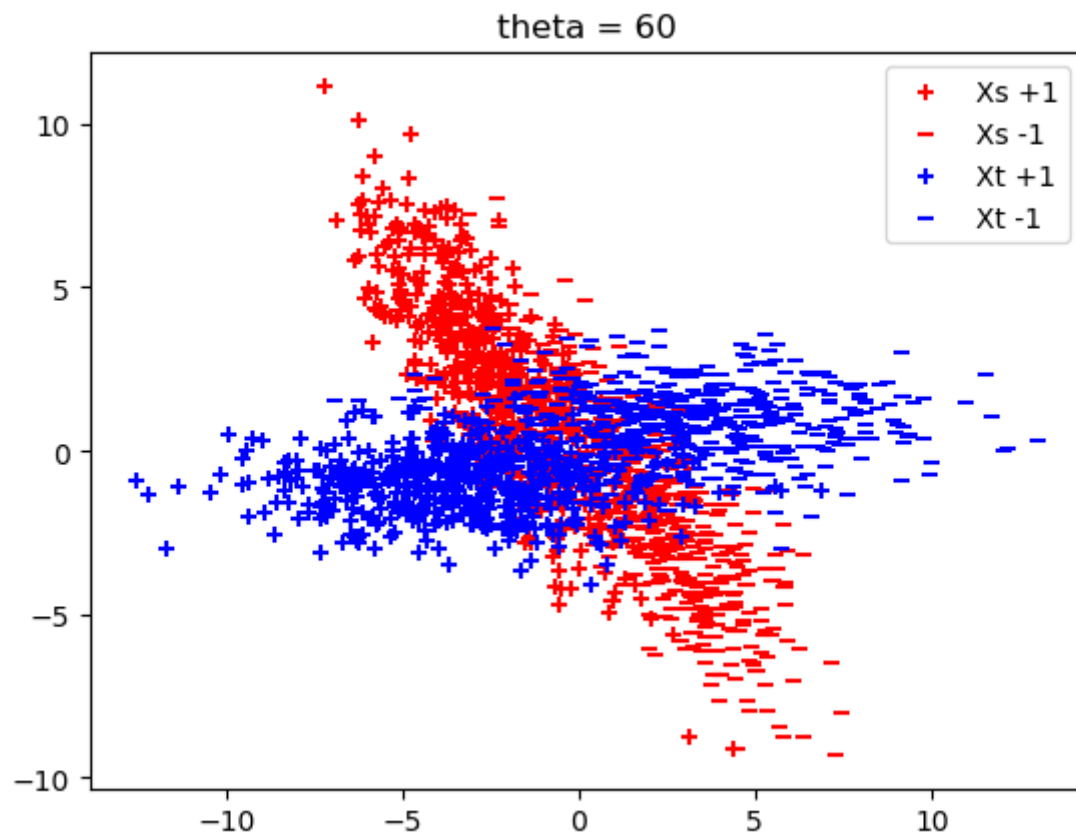


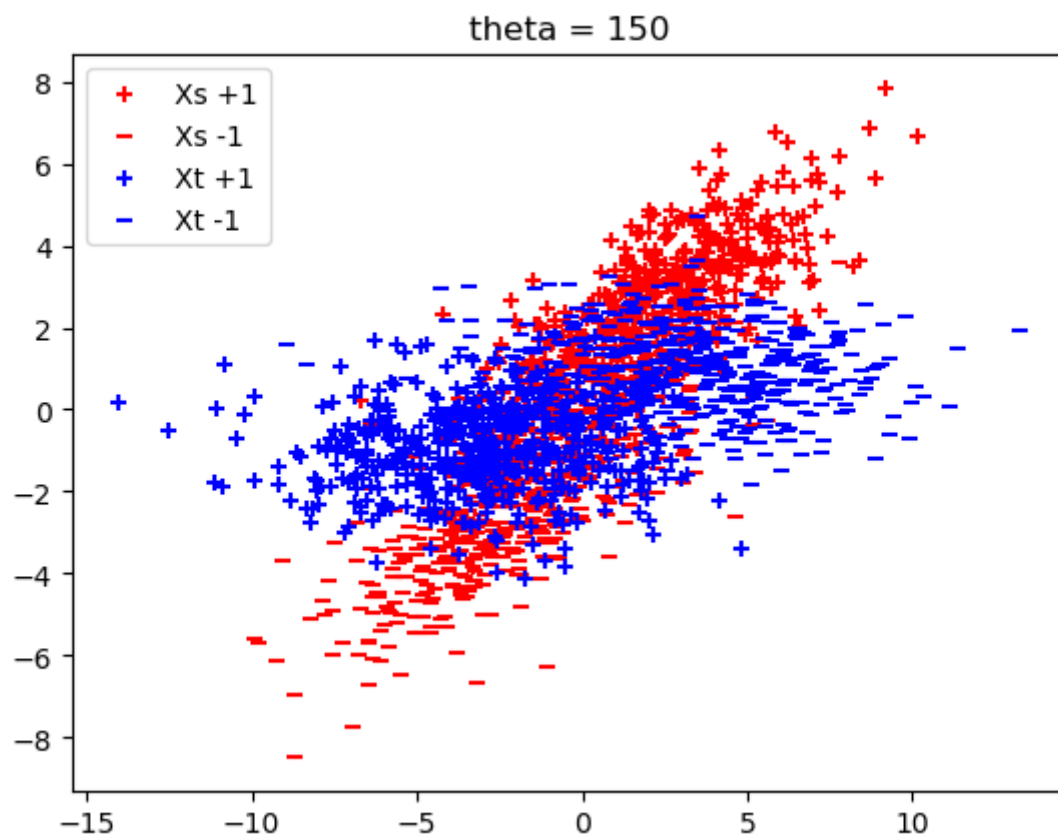
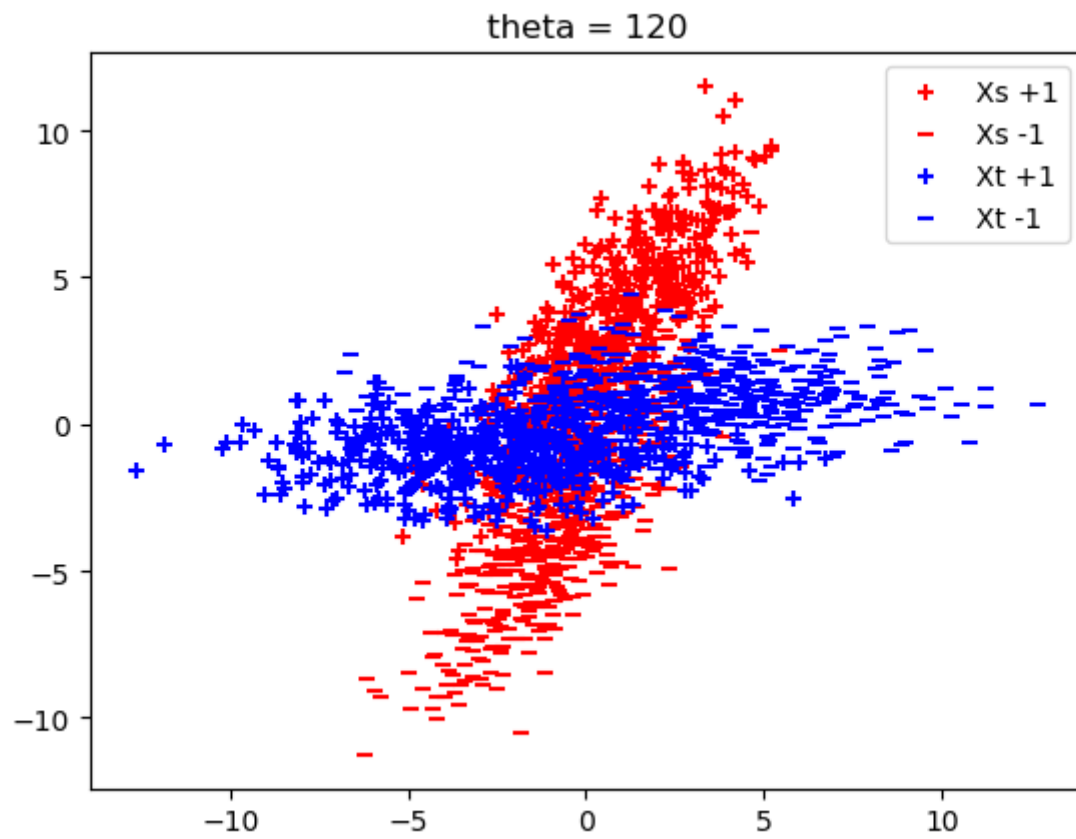
# 1

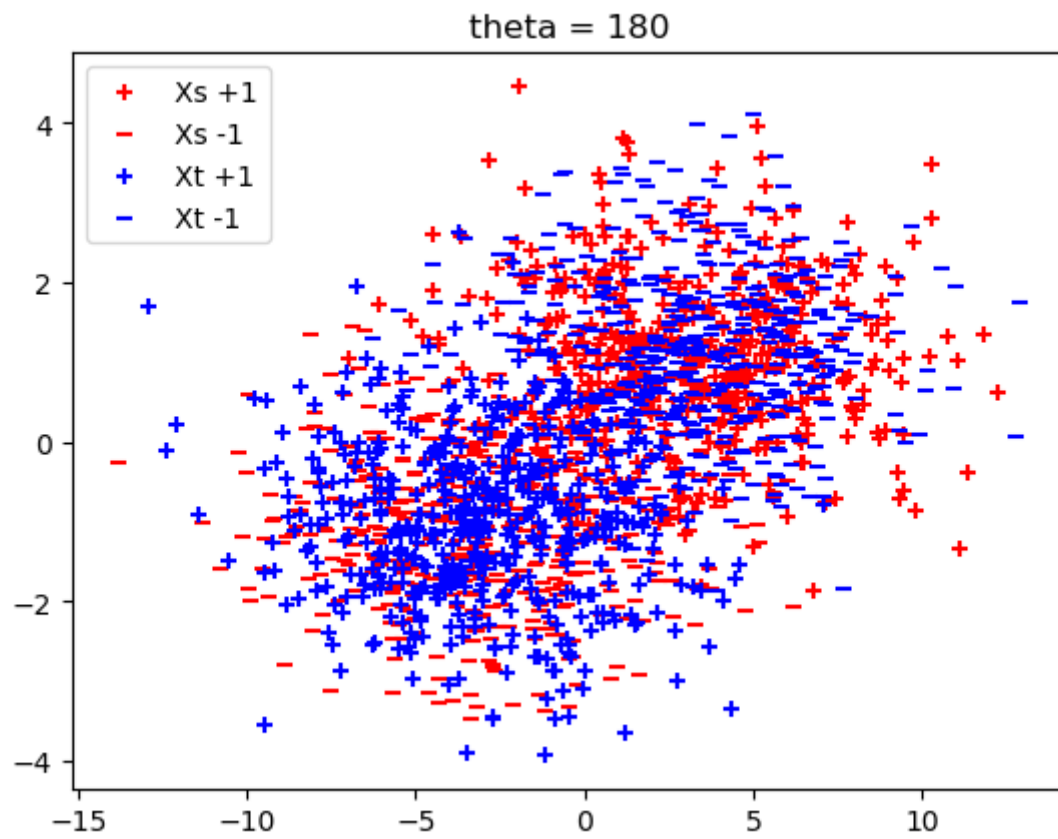
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(a)

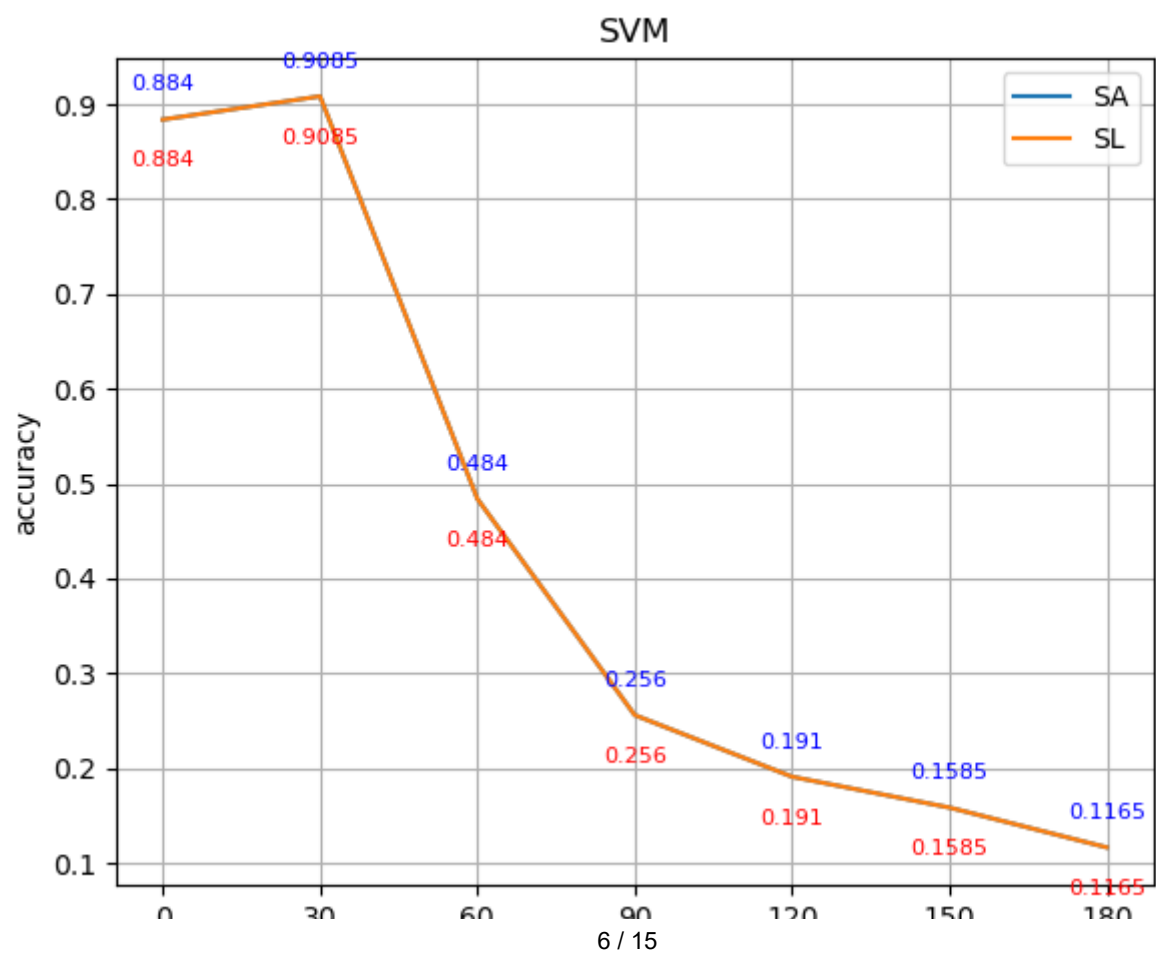
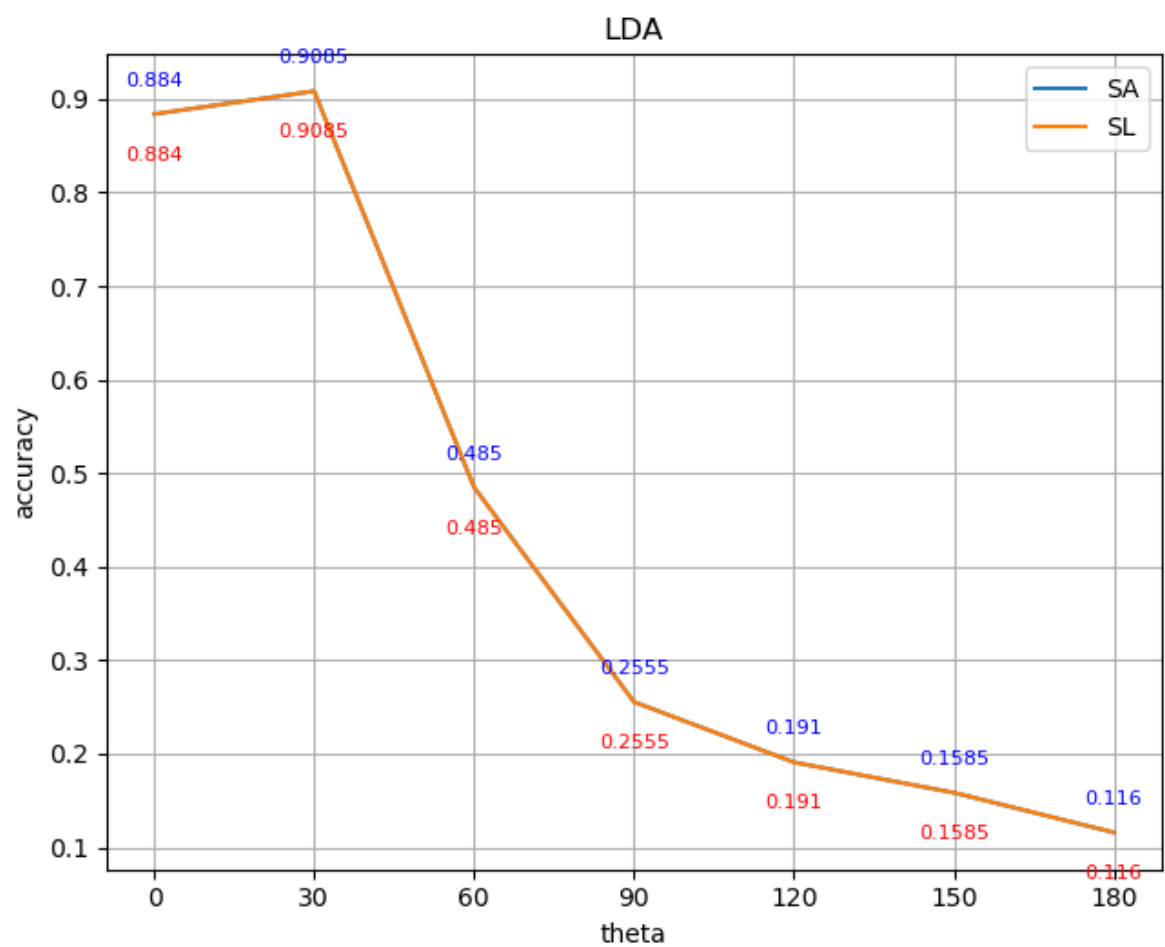


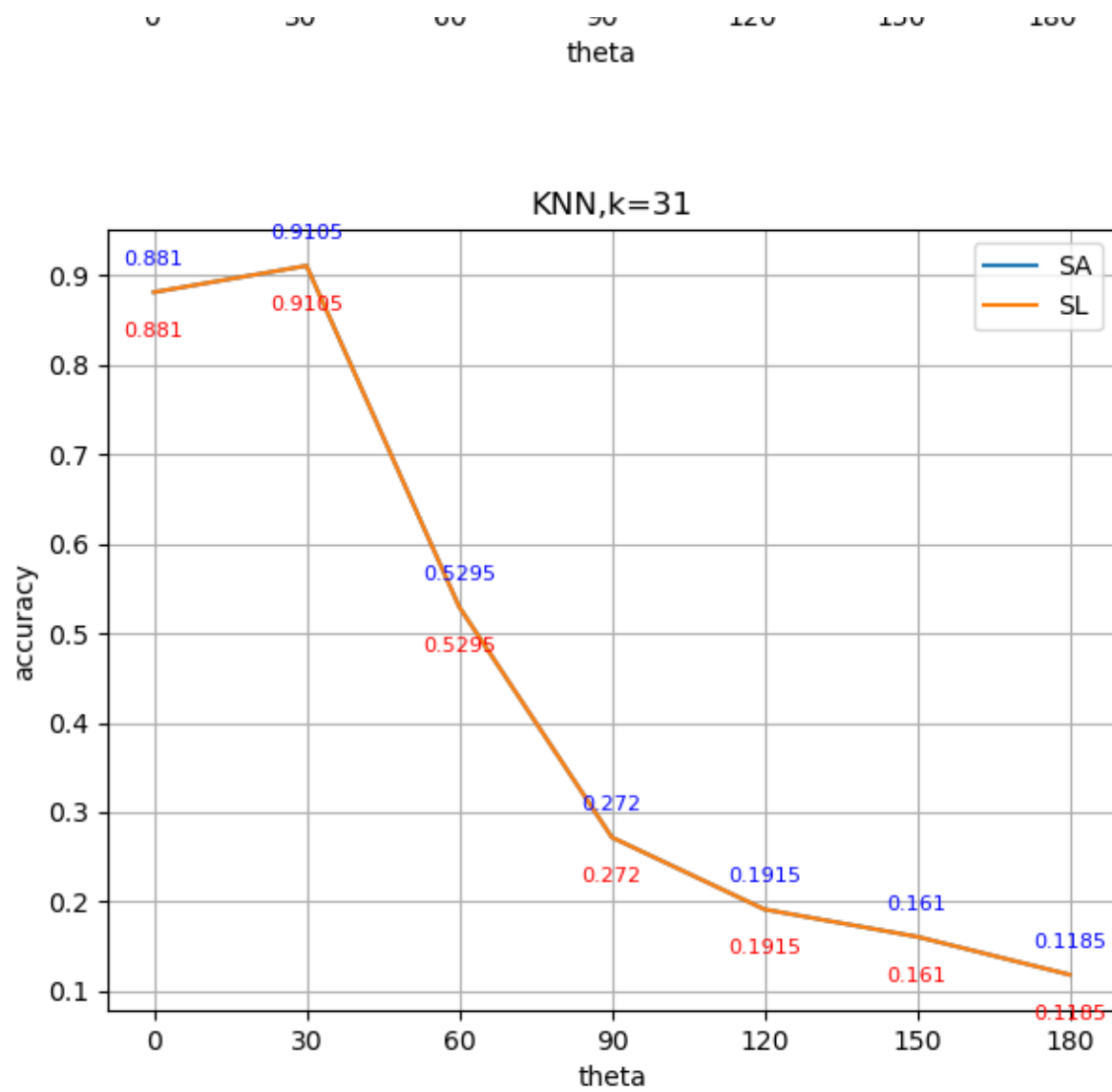






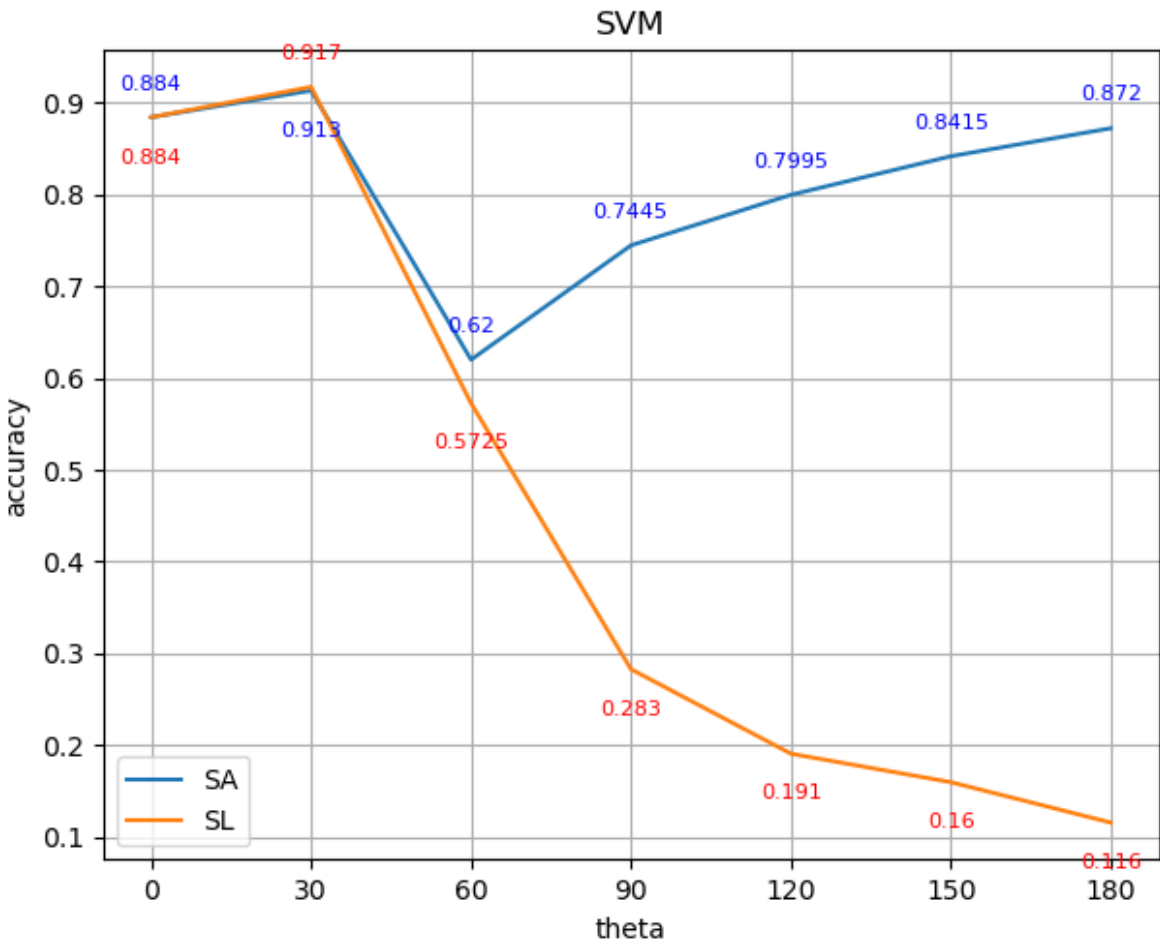
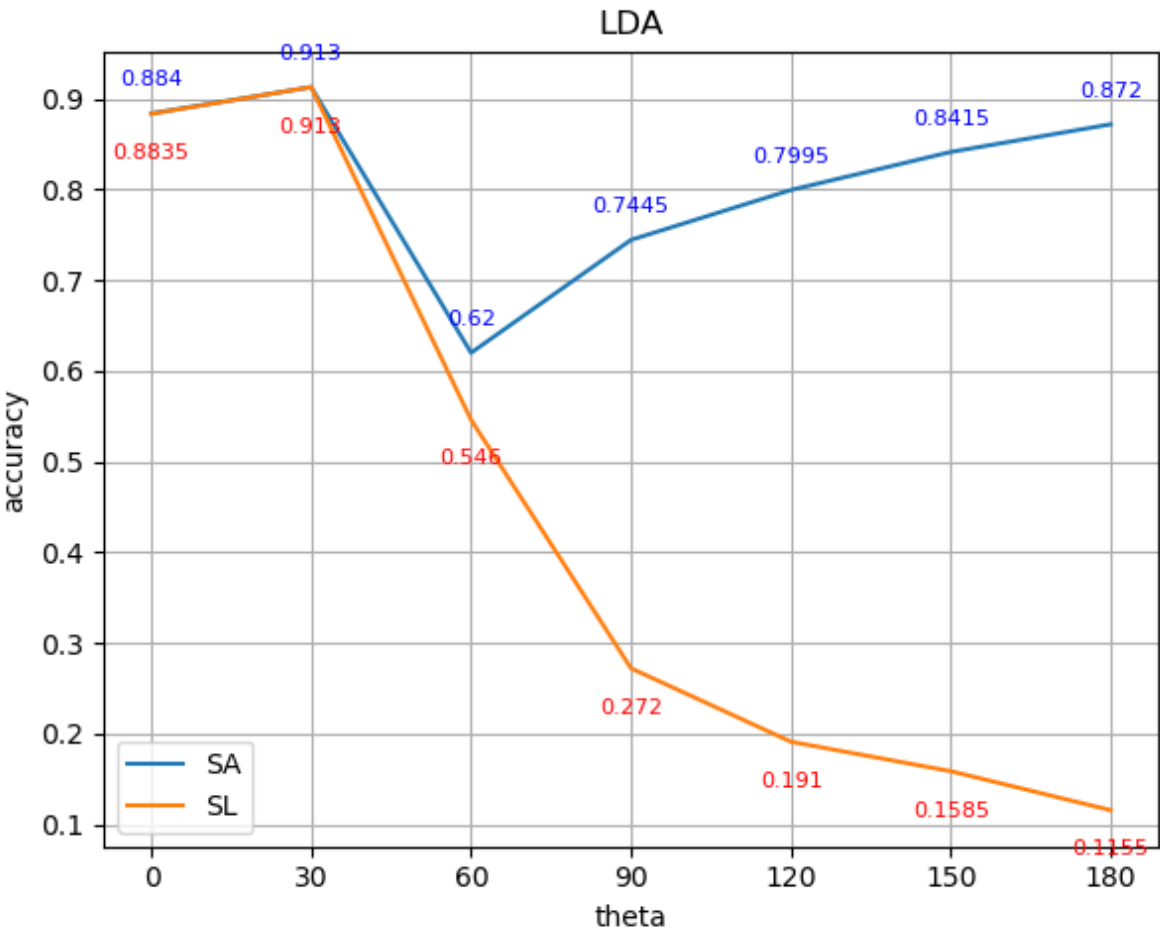
(b)



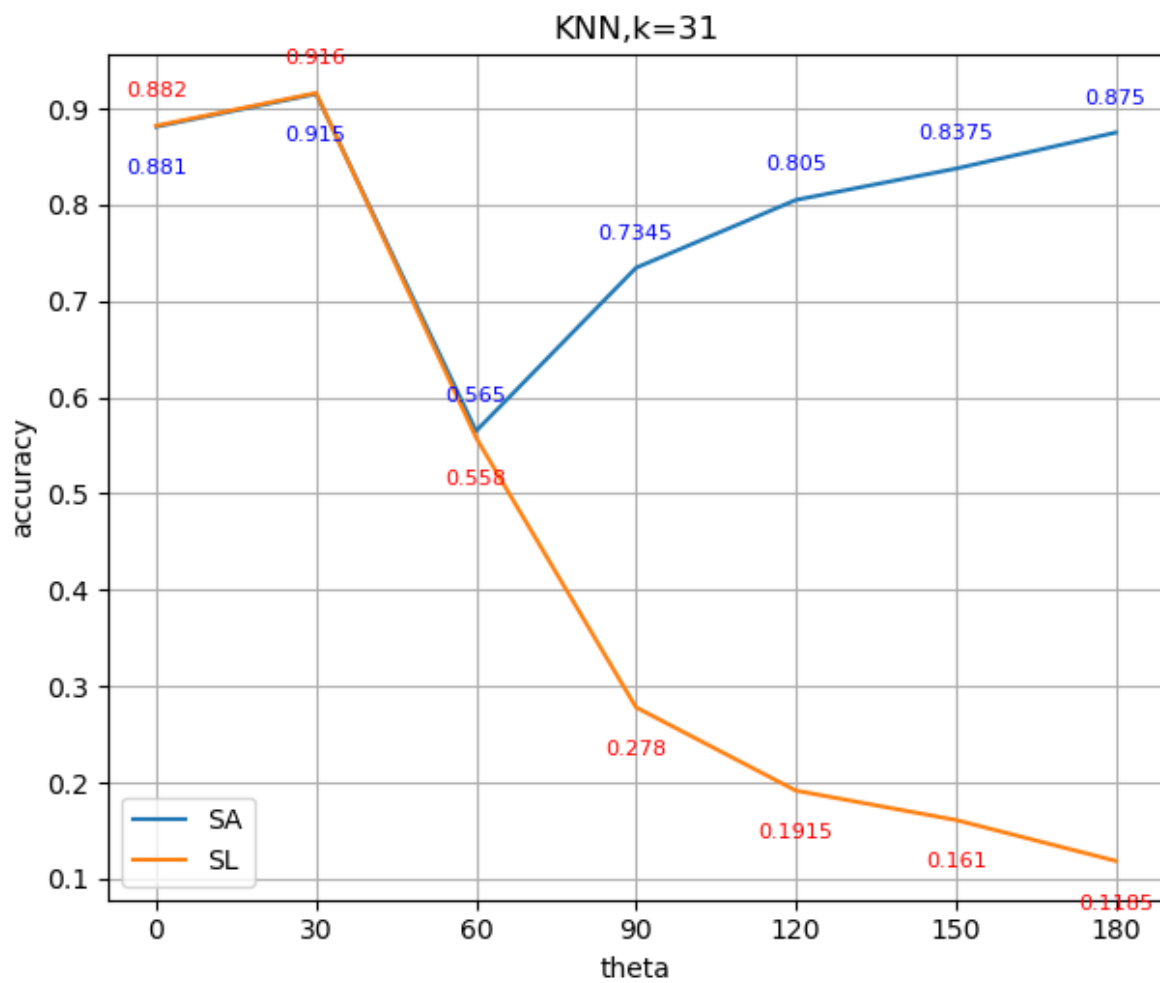


They perform the same.

(c)



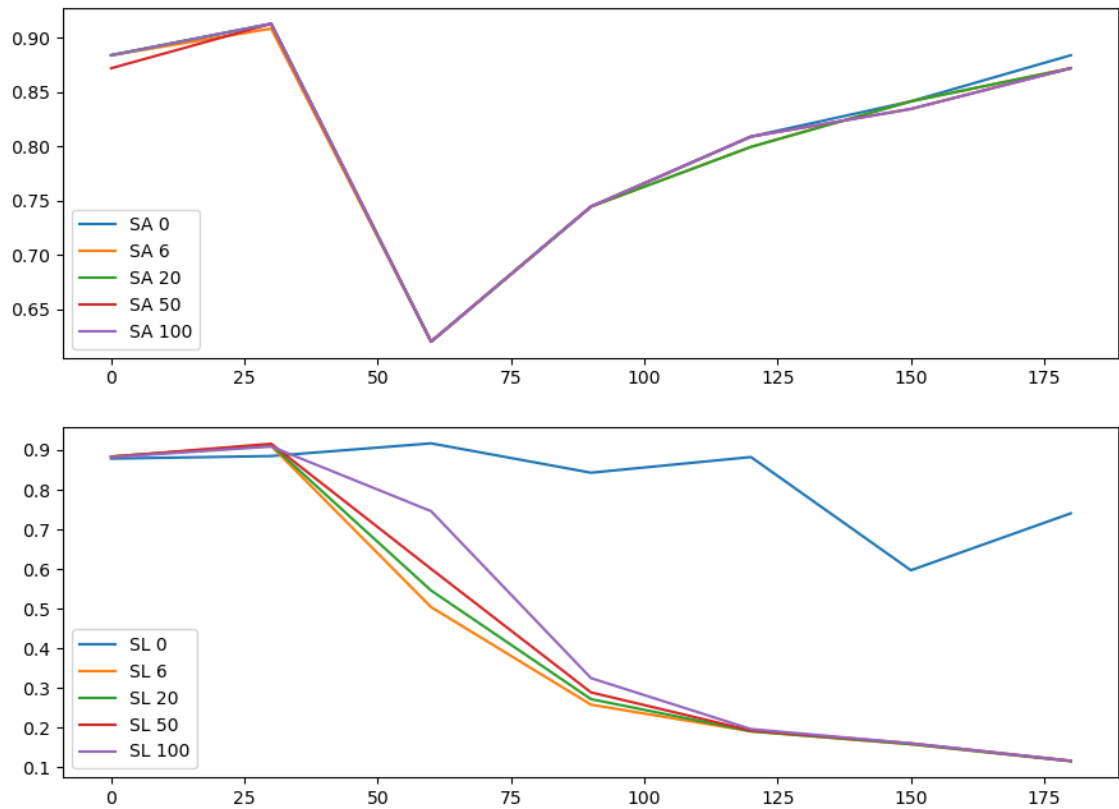




SA performs better.

Sign flipping helps improving performance of SA.

(d)



N\_TL has positive effects on SA with theta is large. Maybe because target domain data can remove the ambiguity of sign of eigenvectors.

N\_TL has negative effects on SL. Maybe because target domain data has no relations to source domain data and mislead classifier.

(d)

sign flipping	standardization	SA	SL
True	False	0.7685	0.6105
True	True	0.913	0.913
False	False	0.561	0.56
False	True	0.9085	0.9085

Maybe standardization makes classifier easier to find optimal points and less affected by those relative large numbers.

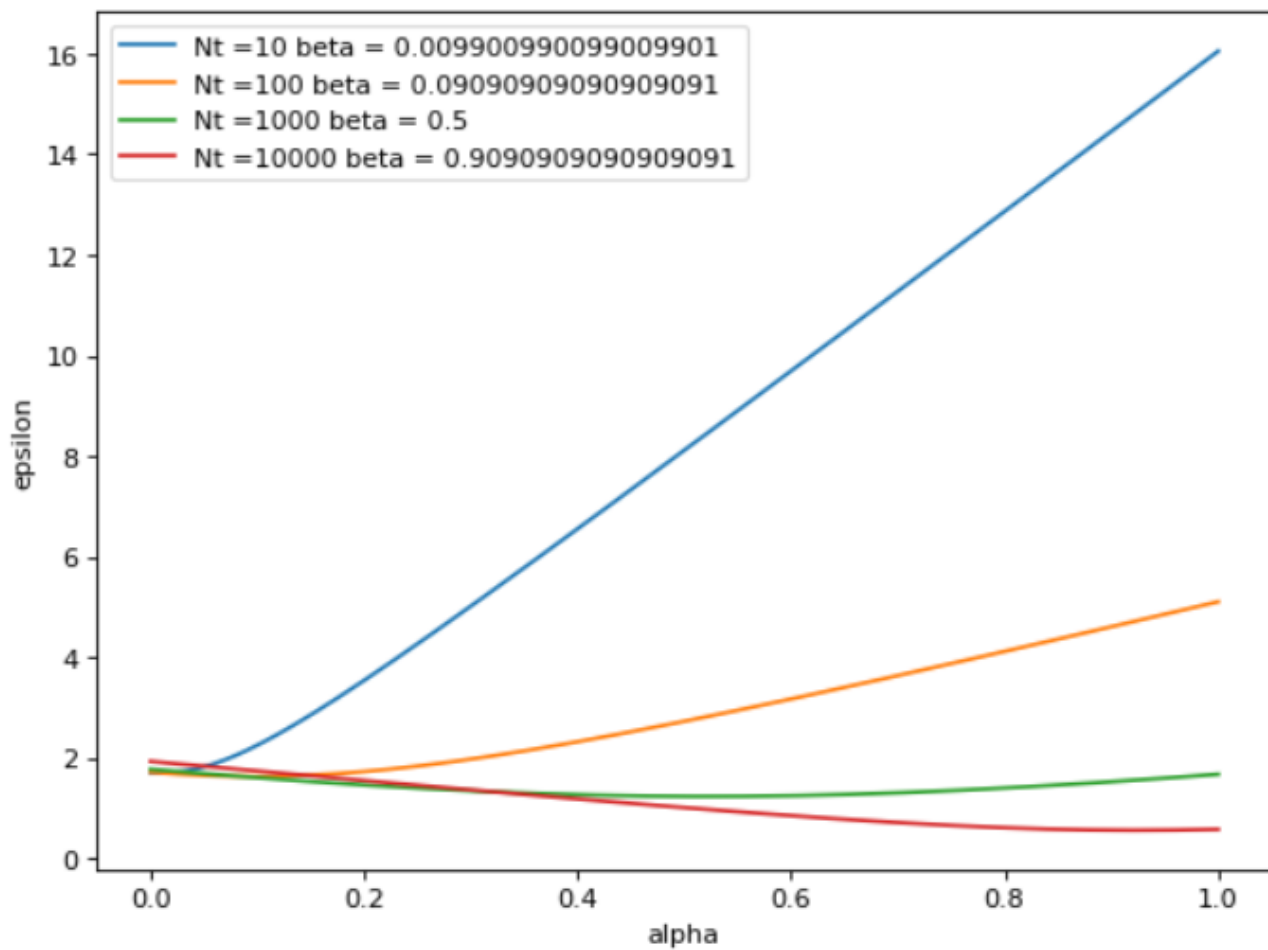
2

(a)

<b>N_T</b>	<b>N_S</b>	<b>alpha</b>	<b>epsilon</b>
1	100	0.1	5.86
1	100	0.5	21.62
1	100	0.9	38.64
---			
10	1000	0.1	2.25
10	1000	0.5	8.12
10	1000	0.9	14.46
---			
100	10000	0.1	0.86
100	10000	0.5	2.94
100	10000	0.9	5.19
---			
1000	100000	0.1	0.36
1000	100000	0.5	1.06
1000	100000	0.9	1.82

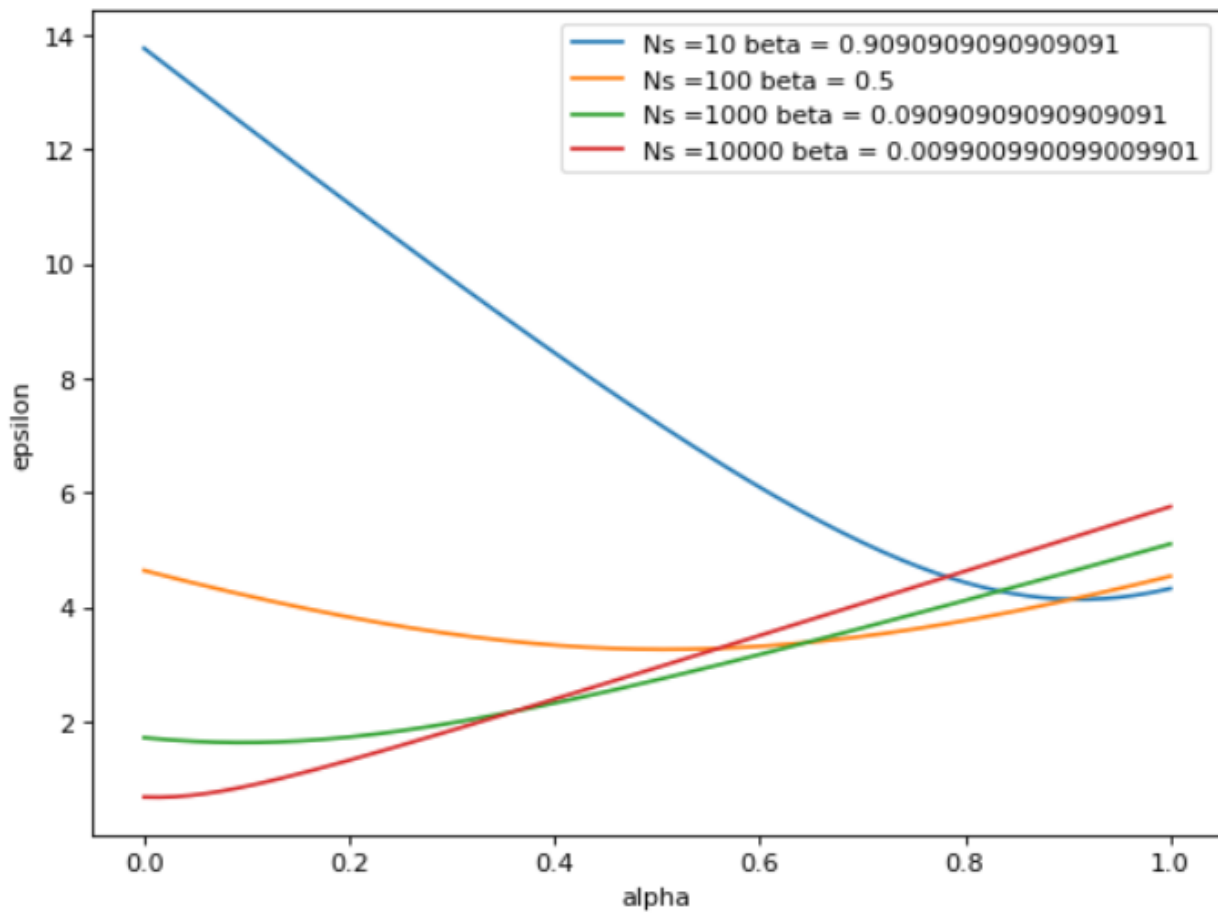
No, it seems that epsilon does not converge.

(b)



As alpha increases, the epsilon goes down first and then goes up. As  $N_T$  goes larger, the optimal value of alpha becomes larger.

(c)

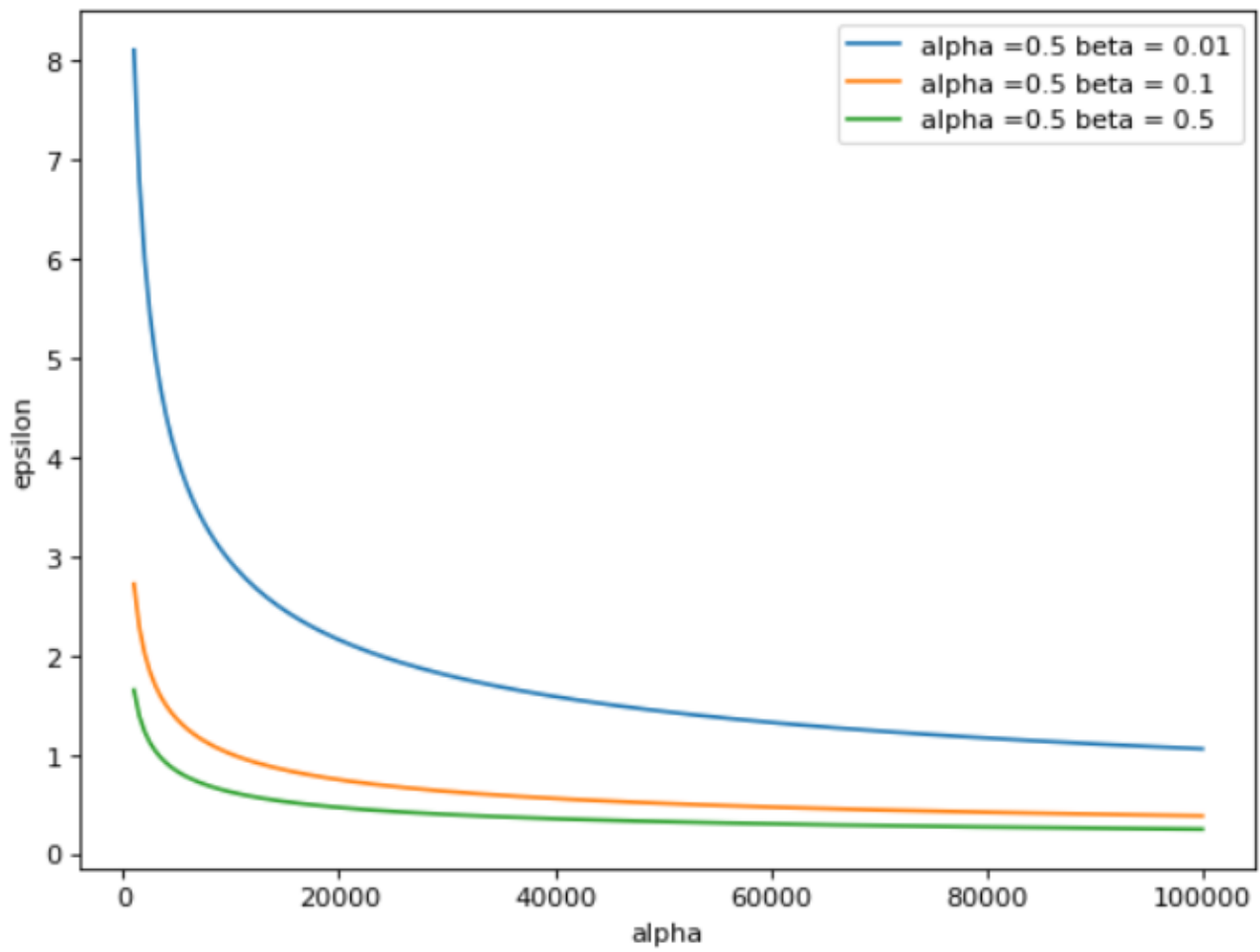


As alpha increases, the epsilon goes down first and then goes up. As N\_S goes larger, the optimal value of alpha becomes smaller.

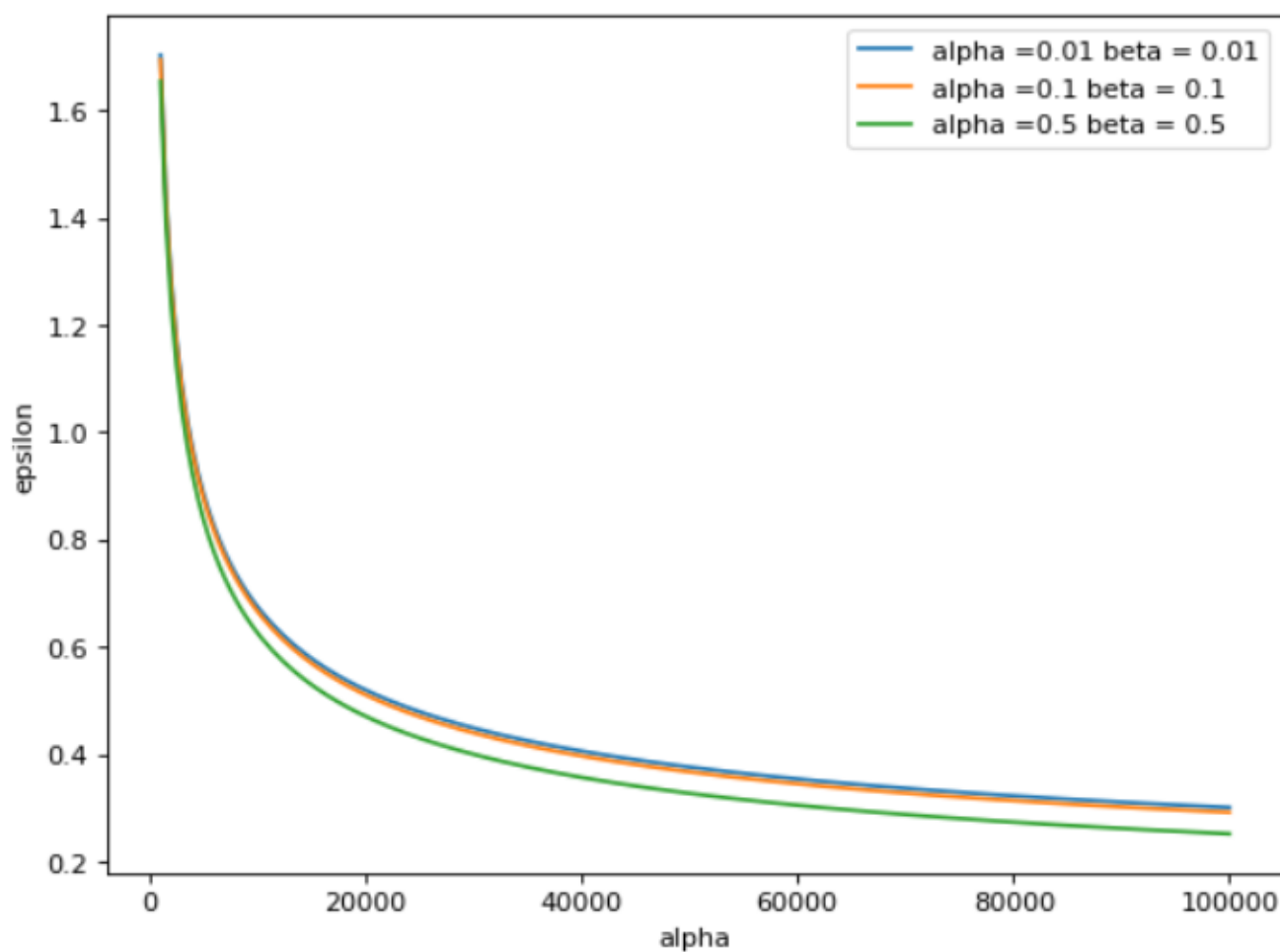
(d)

(iii)

$\alpha = 0.5$



$\alpha = \beta$



$\alpha = 0.5$  can give a better result with less data points. With enough data points,  $\alpha = \beta$  performs better.

(e)

No, it just depends on  $\alpha$ ,  $\beta$  and  $N$ .