Estimate ordinal regression's parameters:

In this section, there are four algorithms, SGD, SGD with gamma decay, ASGD with gamma decay and Batch. These can be help us finding out the parameters. In order to choose the best one, will simulate some dataset and estimate its parameters by above method identically and calculate the error, the smaller one will be the best choice.

After generating data and estimate parameters, we use RMSE to compare performance and summarized the overall RMSE as Table.1.1 and the training time as Table.1.2.

Table.1.1 Overall RMSE with each method in different Epoch

Epoch	SGD	SGD with		Batch SGD	Batch SGD	Batch
		gamma	ASGD with		with	ASGD with
		ga	decay gamma decay		gamma	gamma
		decay			decay	decay
1	0.536017	0.04172	0.046478	0.042584	0.23638	0.34591
2	0.247521	0.035421	0.024703	0.025134	0.218038	0.326721
5	0.444694	0.033234	0.017606	0.030741	0.080138	0.169311
10	0.265318	0.030453	0.013888	0.043584	0.057672	0.123812
20	0.306517	0.030858	0.011511	0.047609	0.01691	0.0698
30	0.404615	0.024136	0.010792	0.034093	0.010392	0.058852
50	0.458559	0.030856	0.010022	0.033026	0.0066	0.030975

Table.1.2 Training time with each method in different Epoch

Epoch	SGD	SGD with gamma decay	ASGD with gamma decay	Batch SGD	Batch SGD with gamma	Batch ASGD with gamma
					decay	decay
1	9.333325	8.690249	8.690249	0.141679	0.154905	0.154905
2	19.6473	19.38494	19.38494	0.250332	0.268075	0.268075
5	50.16117	44.437	44.437	0.671879	0.664393	0.664393
10	98.45409	87.10141	87.10141	1.22089	1.373705	1.373705
20	196.0749	170.5268	170.5268	2.482988	2.75993	2.75993
30	284.3657	258.3534	258.3534	3.756515	4.242949	4.242949
50	452.7361	444.0151	444.0151	6.568452	6.713576	6.713576

Estimate regression parameters

Random variable $X_t \in R^N$, $Y_t \in R^1$ and t = 1, 2, ..., then

$$Y_t = X_t^T \theta + \xi_t$$

 $\theta \in R^N$ is unknown parameters and ξ is a random noise.

$$\theta_t = \theta_{t-1} + \gamma_t \varphi(Y_t - \theta_{t-1}^T X_t) X_t$$

$$\overline{\theta_{\mathsf{t}}} = \frac{1}{t} \sum_{\mathsf{i}=0}^{\mathsf{t}} \theta_{\mathsf{i}}$$

By compare SGD's gradient with MLS, we can find out that

$$\varphi(x) = x$$

Now, let N=5, and $X_t \sim N(5,1) \forall t$

$$B = EX_1 X_1^T = \begin{pmatrix} 26 & 25 & 25 & 25 \\ 25 & 26 & 25 & 25 & 25 \\ 25 & 25 & 26 & 25 & 25 \\ 25 & 25 & 25 & 26 & 25 \\ 25 & 25 & 25 & 25 & 26 \end{pmatrix}$$

$$\psi(X) = E_{\psi}(X + \xi_1) = X$$

$$\chi(X) = E_{\psi^2}(X + \xi_1) = X^2 + 1$$

 $\overline{\theta}_{\rm t} \to \theta$ almost surely and $(\overline{\theta}_{\rm t} - \theta) \sqrt{t} \stackrel{D}{\to} N(0, V)$ where

$$V = B^{-1} \frac{\chi(0)}{\psi'^{2}(0)} = B^{-1}$$

Now, we set t=1000, 10000, 100000, 1000000 and 5000000. And simulated $100 \ \overline{\theta_t}$ and transform them by $(\overline{\theta_t}-\theta)\sqrt{t}$. Then we can calculate the covariance matrix from these data to compare with V to verify the theorem.

In this part, we choose RMSE and MAPE/100 with index to help us displaying the difference between simulation and theoretical value. The result was summarized as Table2.

Table.2 simulation data and theoretical value compare

	t	SGD	ASGD	CCD C D	ASGD_G	MLE
				SGD_GammaDeca	ammaDe	
				У	cay	
R M S	1000	186.796	559.524	542.0701	770 2150	0.0646
		1	9	542.9791	778.2158	
	10000	5.96061	155.422	1002 565	2626.76	0.0796
			6	1003.565		
	100000	65.5071	19.6957	140 1507	2619.446	0.0773
		7	8	149.1597		
Е	100000	696.574	1.48226	2.454555	315.4811	
	0	6	8	2.434333	313.4811	
	500000	3817.50	0.651	4.635725	69.50927	
	0	4	0.031	4.033723	09.30927	
		117.039	351.752			0.0674
	1000	2	2	341.2168	489.2872	
		0.57798	97.7190			0.1179
M	10000	9	5	630.7652	1651.574	
A		16.3794	12.3832			0.0310
P	100000	7	8	93.50557	1646.982	
Е	100000	195.113	0.93048			
	0	8	5	0.550106	198.3595	
	500000	499.009				
	0	2	0.3471	0.463195	43.70374	