Dataset1-9 features

	Mo	odel Selection		Performance		
	Best Param $\log_2 \lambda$	Mean of MSE	STD of MSE	MSE on Train	MSE ON Test	
Least Square	-	-	-	5.275e-28	480.897	
	w	-8.4867963		5861 -2.0105663 3234 -1.3685425 32438] Spars = 0		
LASSO	2.0	1067.508	1492.756	14.107	233.383	
	W	5.01163416		.433000281	423 -0.	
		L1(w)= 18.107	L2(w)= 8.870	Spars = 4		
RIDGE	4.0	-840.630	628.977	18.117	264.804	
	w	[-0.24139412 2.5665958 -0.28155627 -1.71113314 -1.61199141 2.81003838 2.21325862 -3.03423719 -2.75553818 1.62177076]				
		L1(w)= 18.847	L2(w)= 6.669	Spars = 0		

Dataset2-9 features

	Model Selection			Performance			
	Best Param	Mean of MSE	STD of MSE	MSE on Train	MSE ON Test		
	Log ₂ λ						
Least Square	-	-	-	86.336	112.651		
Square	w			.568, -3.870, 0.855, 2.250, .804, 1.254]			
		L1(w)=	L2(w)=	Spars = 0			
		21.654	8.613				
LASSO	0.0	723.813	1269.935	87.635	110.196		
	w	_	1.924361	43 0.43071582 - 18 -6.33525333			
		L1(w)=	L2(w)=	Spars = 1			
		19.533	8.238				
RIDGE	6.0	-121.861	27.571	89.147	111.420		
	w	-0.3226920		0.55844399 - 19 2.05571123 - I			
		L1(w)= 19.546	L2(w)= 7.371	Spars = 0			

Dataset3-9 features

	Model Selection		Performance			
	Best Param	Mean of MSE	STD of MSE	MSE on Train	MSE ON Test	
	Log ₂ λ					
Least Square	-	-	-	98.213	109.124	
Square	w	[1.715, 1.904, 0.412, -3.172, 0.253, 4.872, -0.252, -8.712, 0.805, 0.891]				
		L1(w)= 22.994	L2(w)= 10.864	Spars = 0		
LASSO	-2.0	674.273	1094.949	98.45	109.072	
	w	_		.37895217 -2.91361 3 -0. 0.87016587]	253 0.	
		L1(w)=	L2(w)=	Spars = 3		
		20.265	9.977			
RIDGE	2.0	-102.137	2.223	98.222	108.986	
	W	[1.71554948	1.90414688 0	.41169323 -3.16239	173 0.24355474	
		4.83535943 -0.21617519 -8.46183322 0.55511326 0.8910176]				
		L1(w)=	L2(w)=	Spars = 0		
		22.396	10.628			

Q2) a) ii)

- 1) Test MSE is best obtained with L1 (LASSO) regularizer. For Dataset 1&2, Test MSE was better for Ridge regularizer compared to no regularizer.
- 2) Yes, each regularizer lowers the corresponding norm of w. As training data increases, the difference in corresponding norms starts decreasing (with Ntr → 1000+, it's almost insignificant). This is because, with few training samples, our weight vector takes higher values for Linear regression with no regularizer classifier. As the training samples increases, our Linear regression model with no regularizer performs well and takes lower weight values.
- 3) We obtain sparsity with only LASSO regularizer. As training data increases, our best fit Lambda value (λ) decreases. In general, Sparsity is seen more with lower sized training dataset.

Q2) b) i)

Dataset4-2 features

	N	Model Selection		Performance		
	Best Para m Log ₂	Mean of MSE	STD of MSE	MS	SE on Train	MSE ON Test
Least Squa re	-	-	-	95	.38019904643618	163.48761227397 387
	W	[6.77265711 -2.49	928513 7.238016	12]		
		L1(w)=	L2(w)=		Spars=0	
		16.50352453166	10.22115792212	29		
		5292	628			
LASS	2.0	256.2514695270	249.1697091372	24	98.929344200270	134.48864379047
0		9184	862		42	166
	w	[5.53704387 0. 4.0	00141645]	L		
		L1(w)= 9.538460320163 804	L2(w)= 6.831558271694 6	104	Spars=1	
RIDG	4.0	-196.337	50.48813742733	L63	102.92593770240	127.9073429213
E			1		387	3521
	w	[4.61680832 1.551	139276 2.1674841	[2]		
		L1(w)= 8.335685196859 789	L2(w)= 5.331015470702	211!	Spars=0	

Dataset5-2 features

	N	Model Selection	Pe		Performance	
	Best Para m Log ₂	Mean of MSE	STD of MSE	M	SE on Train	MSE ON Test
Least Squa re	-	-	-	87	7.12261767437649	114.70433167932 684
	W	[4.05510307 2.748	84213 -0.297840	02		
		L1(w)=	L2(w)=		Spars=0	
		7.101785217832	4.90802431192	7	·	
		998	913			
LASS	2.0	139.2991234405	75.9825791058	3	88.723533705402	105.57083378205
0		7894	324		13	994
	w	[3.7532937 2.4795	6376 0.]			
		L1(w)= 6.232857452937 665	L2(w)= 4.49838304224 42	30	Spars=1	
RIDG	6.0	-115.603	14.6318487946	99	88.86631773419	106.0017283071
E			655		617	7882
	w	[3.79650626 2.419	 54902 -0.18839922]]	1
		L1(w)= 6.404454500561 9415	L2(w)= 4.50590407379 1	831	Spars=0	

Dataset6-2 features

	N	Model Selection	Pe		Performance	
	Best Para m Log ₂ λ	Mean of MSE	STD of MSE	MS	SE on Train	MSE ON Test
Least Squa re	-	-	-	10	1.35833777888638	101.44570933707 959
	w	[1.21077089 2.302	240071 0.2315254	47]		
		L1(w)= 3.744697069153 53	L2(w)= 2.61163152696 835	79	Spars=0	
LASS O	-10.0	137.0553420605 17	63.83629912070 816	08	101.35833791468 285	101.44595275106 839
	w	[1.2107991 2.3023	36633 0.23141756	6]		
		L1(w)= 3.744582986648 264	L2(w)= 2.611604730399 63	586	Spars=0	
RIDG E	6.0	-112.144	4.024677329122 95	253	101.47660067151 448	101.3026147405 2405
	W	[1.22012634 2.218	80724 0.24090883	3]	- L	1
		L1(w)= 3.679107562730 1373	L2(w)= 2.542949176508	362	Spars=0	

Dataset7-2 features

	N	Model Selection		F	Performance	
	Best Para m	Mean of MSE	STD of MSE	MS	SE on Train	MSE ON Test
	\log_2 λ					
Least Squa	-	-	-	25	.417551693358597	116.51141337592
re						615
	w	[1.6193184 4.358	346137 -2.05316003]			
		L1(w)=	L2(w)=		Spars=0	
		8.030939797861	5.08270043370	7		
		016	281			
LASS	0.0	80.56417433998	88.3922127145	7	26.522508540648	108.51285152596
0		845	387		026	137
	w	[1.49398093 3.84	541442 -1.46163	77		
		L1(w)=	L2(w)=		Spars=0	
		6.801033058092	4.37670833943	42		
		814	2			
RIDG	2.0	-70.890	13.9113914290	17	26.617548498116	109.2976131253
E			938		584	2439
	w	[1.59580358 3.78	3666035 -1.4946	5134	4]	•
		L1(w)=	L2(w)=		Spars=0	
		6.877115268463	4.37256998866	517		
		562	7			

Dataset8-2 features

	N	Model Selection	Pe		Performance	
	Best Para m Log ₂	Mean of MSE	STD of MSE	M	SE on Train	MSE ON Test
Least Squa re	-	-	-	95	5.15432277075584	109.24257017878 496
	W	[3.58068323 1.918	63829 0.6043447	73]		
		L1(w)=	L2(w)=		Spars=0	
		6.103666255666	4.10703029596	9	·	
		34	647			
LASS	0.0	155.9448030451	88.2502422075	9	95.201467966002	109.46763164804
0		9337	911		13	094
	W	[3.51599517 1.882	23308 0.593483]		
		L1(w)= 5.991711258559 851	L2(w)= 4.03202746914 41	01	Spars=0	
RIDG	6.0	-115.911	10.6981996182	03	95.90349733909	110.5119829694
E			887		382	2115
	w	[3.20005109 1.411	74703 0.97725369]			1
		L1(w)= 5.589051809257 596	L2(w)= 3.63158111896 24	508	Spars=0	

Dataset9-2 features

	N	Model Selection	Performance			
	Best Para m Log ₂	Mean of MSE	STD of MSE	MS	SE on Train	MSE ON Test
Least Squa re	-	-	-	83	.3240152571577	111.41165530589 785
	w	[4.11404128 3.04	009919 -0.51630)424	1]	
		L1(w)= 7.670444709675 4755	L2(w)= 5.14141116684 956		Spars=0	
LASS O	0.0	123.0456877361 9322	57.1598361765 494		83.90836860287 891	109.50398425369 923
	W	[4.10151715 2.504	147238 0.]			
		L1(w)= 6.605989527505 749	_		Spars=1	
RIDG E	4.0	-90.854	3.34154427448 117	63	83.44263614295 154	110.3600950184 0029
	W	[4.10848705 2.79				
		L1(w)= 7.190815012245 55	L2(w)= 4.97928384123 2	L2(w)= 4.97928384123978		

Q2) b) ii) --- plots are in the pdf file containing code part. Please refer over there.

Q2) b) iii)

1) The plots have 2 figures each, MSE AND regularizer constraints drawn over a range of w1 and w2 for a particular value of w0(for the best lambda). We can see that MSE with no regularizer intercepts the constraints contour at greater values of w1 and w2. Also, LASSO (I1 regularizer)

- gives us sparsity which can be observed by the intersection of MSE (lowest) with constraints contour at either w1=0 or w2=0.
- 2) With lower training data size, the regularizer has a larger effect on MSE. Regularized MSE takes a lower value compared with unregularized MSE.
- and Ntr=30 but it does give sparsity for Ntr=100 which is not the case in set of datasets 4,5,6 where we got sparsity for Ntr=10(dataset 4) and Ntr=30(dataset 5) but did not get sparsity for Ntr=100(dataset 6). In general, Sparsity is observed in lower sized training datasets as its easier for model to overfit training data with lesser samples and obtain poor generalization. This was opposite with Datasets 7,8,9 (Special Case).