

$$1. \quad (i) \quad \min_{w_1, w_0 \in \mathbb{R}} \frac{1}{N} \sum_{t=1}^N (r^t - (w_1 x^t + w_0))^2$$

$$\text{error} = e = \sum_{t=1}^N \|r^t - (w_1 x^t + w_0)\|^2$$

$$\frac{\partial e}{\partial w_1} = 0 \Rightarrow \sum_{t=1}^N (r^t - (w_1 x^t + w_0)) x^t = 0 \Rightarrow \left(\sum_{t=1}^N r^t x^t - w_1 \sum_{t=1}^N (x^t)^2 - w_0 \sum_{t=1}^N x^t \right) = 0$$

$$\Rightarrow w_1 \sum_{t=1}^N (x^t)^2 + w_0 \sum_{t=1}^N x^t = \sum_{t=1}^N r^t x^t$$

$$\frac{\partial e}{\partial w_0} = 0 \Rightarrow \sum_{t=1}^N (r^t - (w_1 x^t + w_0)) = 0 \Rightarrow \sum_{t=1}^N r^t - w_1 \sum_{t=1}^N x^t - N w_0 = 0$$

$$\Rightarrow w_1 \sum_{t=1}^N x^t + N w_0 = \sum_{t=1}^N r^t$$

$$\begin{pmatrix} \sum_{t=1}^N (x^t)^2 & \sum_{t=1}^N x^t \\ \sum_{t=1}^N x^t & N \end{pmatrix} \cdot \begin{pmatrix} w_1 \\ w_0 \end{pmatrix} = \begin{pmatrix} \sum_{t=1}^N r^t x^t \\ \sum_{t=1}^N r^t \end{pmatrix}$$

$$\text{Let } A = \begin{pmatrix} \sum_{t=1}^N (x^t)^2 & \sum_{t=1}^N x^t \\ \sum_{t=1}^N x^t & N \end{pmatrix} \quad \det A = \sum_{t=1}^N (x^t)^2 \cdot (N-1) \quad \text{if } N \neq 1, \begin{pmatrix} w_1 \\ w_0 \end{pmatrix} = A^{-1} \begin{pmatrix} \sum_{t=1}^N r^t x^t \\ \sum_{t=1}^N r^t \end{pmatrix}$$

$$\text{if } N=1 \quad \det A = 0 \quad \text{eliminator} \quad A \rightarrow \begin{pmatrix} \sum_{t=1}^N (x^t)^2 & \sum_{t=1}^N x^t \\ 0 & 0 \end{pmatrix}$$

$$\text{Thus } 0 = r, \quad \sum_{t=1}^N (x^t)^2 \cdot w_1 + \sum_{t=1}^N x^t w_0 = 0 \Rightarrow x \cdot w_1 + w_0 = 0$$

$$\begin{pmatrix} w_1 \\ w_0 \end{pmatrix} = \begin{pmatrix} 1 \\ -x \end{pmatrix} \quad \text{where } x \in \mathbb{R}$$

$$(ii) \quad \text{error} = e = \sum_{t=1}^N (r^t - (v_2 (x^t)^2 + v_1 x^t + v_0))^2$$

$$\frac{\partial e}{\partial v_2} = 0 \Rightarrow \sum_{t=1}^N (r^t - (v_2 (x^t)^2 + v_1 x^t + v_0)) \cdot (x^t)^2 = 0$$

$$\Rightarrow v_2 \sum_{t=1}^N (x^t)^4 + v_1 \sum_{t=1}^N (x^t)^3 + v_0 \sum_{t=1}^N (x^t)^2 = \sum_{t=1}^N r^t (x^t)^2$$

$$\frac{\partial e}{\partial v_1} = 0 \Rightarrow \sum_{t=1}^N (r^t - (v_2 (x^t)^2 + v_1 x^t + v_0)) x^t = 0$$

$$\Rightarrow v_2 \sum_{t=1}^N (x^t)^3 + v_1 \sum_{t=1}^N (x^t)^2 + v_0 \sum_{t=1}^N x^t = \sum_{t=1}^N r^t x^t$$

$$\frac{\partial e}{\partial V_0} = 0 \Rightarrow \sum_{i=1}^N (y^i - (V_2 (x^i)^2 + V_1 x^i + V_0)) = 0$$

$$\Rightarrow V_2 \sum_{i=1}^N (x^i)^2 + V_1 \sum_{i=1}^N x^i + N V_0 = \sum_{i=1}^N y^i$$

In summary,

$$\text{Let } A = \begin{pmatrix} \sum_{i=1}^N (x^i)^4 & \sum_{i=1}^N (x^i)^3 & \sum_{i=1}^N (x^i)^2 \\ \sum_{i=1}^N (x^i)^3 & \sum_{i=1}^N (x^i)^2 & \sum_{i=1}^N x^i \\ \sum_{i=1}^N (x^i)^2 & \sum_{i=1}^N x^i & N \end{pmatrix}, \quad A \cdot \begin{pmatrix} V_2 \\ V_1 \\ V_0 \end{pmatrix} = \begin{pmatrix} \sum_{i=1}^N y^i (x^i)^2 \\ \sum_{i=1}^N y^i x^i \\ \sum_{i=1}^N y^i \end{pmatrix}$$

Acquisition

$$\begin{pmatrix} \sum_{i=1}^N (x^i)^4 & \sum_{i=1}^N (x^i)^3 & \sum_{i=1}^N (x^i)^2 \\ 0 & 0 & 0 \\ 0 & 0 & \sum_{i=1}^N x^i \cdot (N-1) \end{pmatrix} \quad \text{and } A=0$$

$$\Rightarrow V_2 \sum_{i=1}^N (x^i)^4 + V_1 \sum_{i=1}^N (x^i)^3 + V_0 \sum_{i=1}^N (x^i)^2 = \sum_{i=1}^N y^i (x^i)^2$$

$$0 = \sum_{i=1}^N y^i x^i$$

$$V_0 \sum_{i=1}^N x^i (N-1) = \sum_{i=1}^N y^i$$

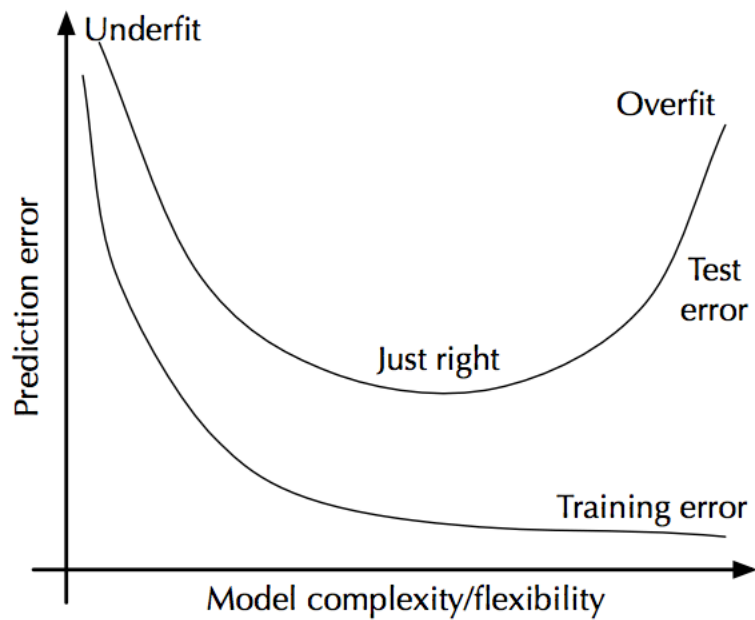
$$\Rightarrow V_2 \sum_{i=1}^N (x^i)^4 + V_1 \sum_{i=1}^N (x^i)^3 = 0$$

$$\Rightarrow \begin{pmatrix} V_2 \\ V_1 \end{pmatrix} = \begin{pmatrix} 1 \\ -\sum_{i=1}^N (x^i)^3 \end{pmatrix}$$

(iii) Yes. As the order of the polynomial model increases, the error in the training set decreases, since it better fits the training data.

(iv) No. It's hard to tell which model has lower error since on the one hand the higher order model could better fit the data but it may overfit for some test sets. So the overall performance is usually indeterminate.

The relationship between model complexity and prediction error can be described in the graph below:



http://www.stats.ox.ac.uk/~sejdinov/teaching/sdmml15/materials/HT15_lecture12-nup.pdf

$$2. (i) \text{tr}(A) = 1 + 2 + 9 + 64 + 625 = 701$$

$$\text{tr}(A^T) = \text{tr}(A) = 701$$

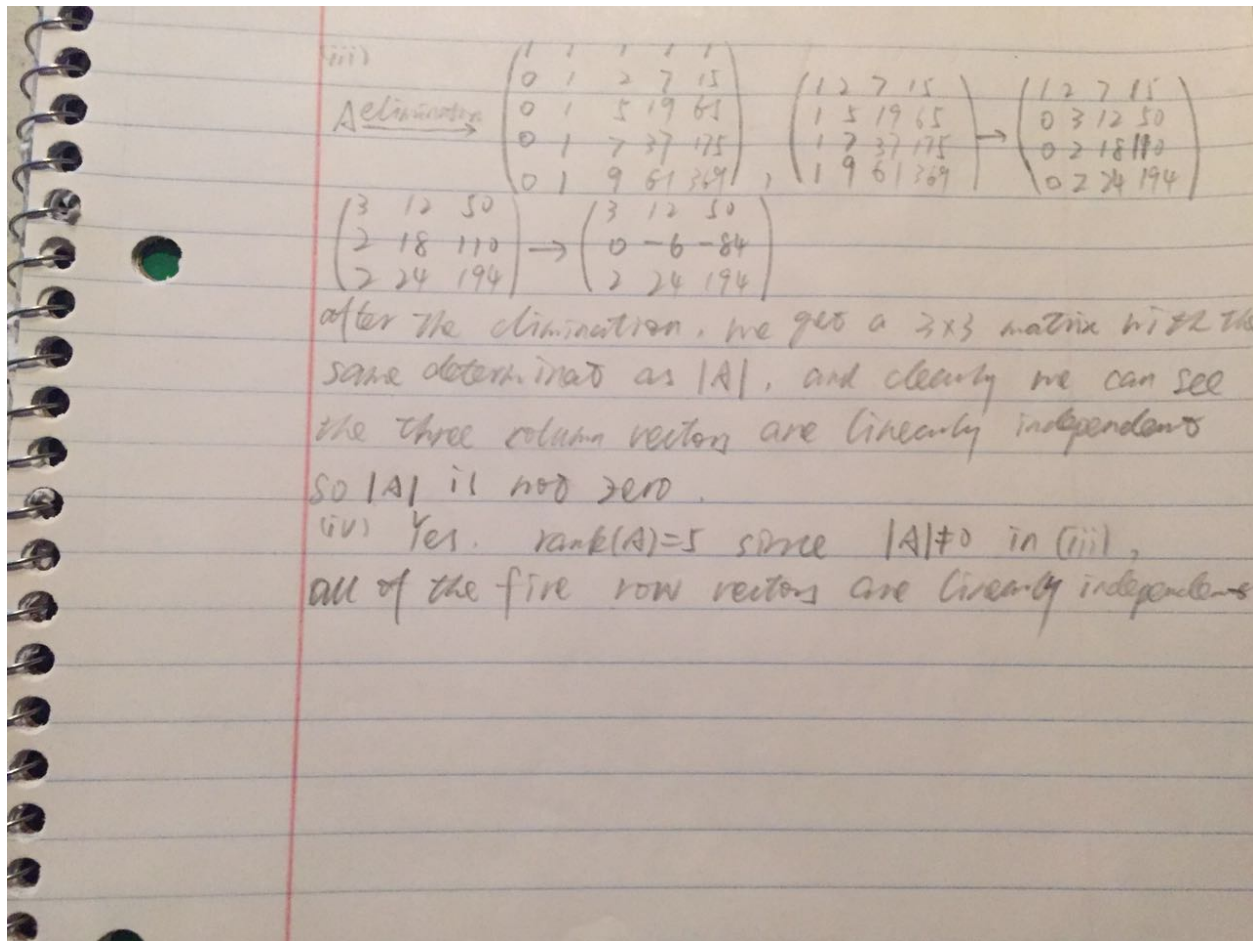
$$\text{tr}(A^T A) = \sum_{i=0}^4 \sum_{j=0}^4 a_{ij}^2 = \left(\sum_{i=0}^4 1^2 \right) + \left(\sum_{i=0}^4 2^2 \right) + \left(\sum_{i=0}^4 3^2 \right) + \left(\sum_{i=0}^4 4^2 \right) + \left(\sum_{i=0}^4 5^2 \right) = 48433$$

$$\text{tr}(A A^T) = \text{tr}(A^T A) = 48433$$

(ii)

$\det A = \text{product of all of } A\text{'s eigenvalues.}$

A can be projected in a way that only its eigenvectors stretch without rotating. And each eigenvector is stretched to an extent that can be measured by multiplying its eigenvalue and all the other vectors can be expressed as a linear combination of all the eigenvectors and thus the volume is only changed to a new one which is the certain amount times the original volume. And the change rate of volume can be calculated by multiplying all the eigenvalues as they measure the change rate of each eigenvector and all the other vector can be expressed as a linear combination of these eigenvectors.



Q3

(i)

LinearSVC with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.24	0.2	0.14	0.7	0.14	0.08	0.34	0.3	0.1	0.58	0.282	0.197474048928

LinearSVC with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.18	0.52	0.04	0.42	0.36	0.32	0.64	0.16	0.04	0.04	0.272	0.202820117345

LinearSVC with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.094972 0670391	0.072625 698324	0.156424 581006	0.100558 659218	0.072625 698324	0.050279 3296089	0.033519 5530726	0.072625 698324	0.122905 027933	0.011173 1843575	0.078770 9497207	0.04032033 64822

SVC with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.66	0.28	0.84	0.84	0.88	0.9	0.46	0.46	0.8	0.0	0.612	0.287638662214

SVC with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.18	0.26	0.04	0.6	0.46	0.66	0.2	0.16	0.04	0.02	0.262	0.221350400948

SVC with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.603351 955307	0.491620 111732	0.625698 324022	0.581005 586592	0.530726 256983	0.586592 178771	0.603351 955307	0.519553 072626	0.446927 374302	0.0055865 9217877	0.499441 340782	0.17316 993993

LogisticRegression with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.12	0.16	0.12	0.12	0.18	0.06	0.28	0.28	0.04	0.16	0.152	0.076

LogisticRegression with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.08	0.08	0.04	0.16	0.14	0.12	0.12	0.14	0.04	0.04	0.096	0.0436348484585

LogisticRegression with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.094972 0670391	0.050279 3296089	0.106145 251397	0.094972 0670391	0.055865 9217877	0.033519 5530726	0.016759 7765363	0.050279 3296089	0.128491 620112	0.011173 1843575	0.064245 8100559	0.037745 6892326

(ii)

LinearSVC with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.528	0.504	0.504	0.496	0.536	0.496	0.528	0.528	0.552	0.496	0.5168	0.0189989473393

LinearSVC with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.296	0.256	0.264	0.272	0.712	0.256	0.648	0.232	0.264	0.464	0.3664	0.168979998816

LinearSVC with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.926339 285714	0.924107 142857	0.899553 571429	0.901785 714286	0.901785 714286	0.883928 571429	0.9174 10714 286	0.89062 5	0.901785 714286	0.904017 857143	0.905133 928571	0.0129867 751855

SVC with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.536	0.552	0.512	0.6	0.528	0.44	0.568	0.536	0.48	0.432	0.5184	0.0510748470384

SVC with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.28	0.304	0.2	0.224	0.272	0.288	0.248	0.232	0.296	0.288	0.2632	0.0332649966181

SVC with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.928571 428571	0.901785 714286	0.924107 142857	0.908482 142857	0.90 625	0.90 625	0.90 625	0.912946 428571	0.919642 857143	0.915178 571429	0.912946 428571	0.0083519 1380976

LogisticRegression with Boston50

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.488	0.432	0.48	0.52	0.512	0.488	0.44	0.464	0.504	0.44	0.4768	0.0299759903923

LogisticRegression with Boston75

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.248	0.256	0.248	0.272	0.28	0.28	0.288	0.192	0.208	0.296	0.2568	0.0324863048068

LogisticRegression with Digits

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.912946 428571	0.881696 428571	0.917410 714286	0.895089 285714	0.892857 142857	0.919642 857143	0.908482 142857	0.926339 285714	0.881696 428571	0.8950 892857 14	0.90312 5	0.015145 7102265

Q4

LinearSVC with ~X1

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.156424 581006	0.033519 5530726	0.156424 581006	0.122905 027933	0.134078 212291	0.089385 4748603	0.039106 1452514	0.055865 9217877	0.134078 212291	0.111731 843575	0.103351 955307	0.044077 4688214

LinearSVC with ~X2

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.05027 9329608 9	0.0	0.0670391 061453	0.0335195 530726	0.0223463 687151	0.0223463 687151	0.0055865 9217877	0.0111731 843575	0.0391061 452514	0.0	0.025139 6648045	0.0212 36399 7863

LinearSVC with ~X3

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.111731 843575	0.039106 1452514	0.067039 1061453	0.100558 659218	0.061452 5139665	0.067039 1061453	0.044692 7374302	0.055865 9217877	0.100558 659218	0.016759 7765363	0.066480 4469274	0.028644 5670241

SVC with ~X1

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.905027 932961	0.916201 117318	0.905027 932961	0.899441 340782	0.899441 340782	0.91061 452514	0.91061 452514	0.905027 932961	0.905027 932961	0.0055865 9217877	0.816201 117318	0.270248 473415

SVC with ~X2

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.905027 932961	0.916201 117318	0.905027 932961	0.748603 351955	0.899441 340782	0.91061 452514	0.91061 452514	0.905027 932961	0.905027 932961	0.0055865 9217877	0.801117 318436	0.269389 622485

SVC with ~X3

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.905027 932961	0.916201 117318	0.905027 932961	0.899441 340782	0.899441 340782	0.91061 452514	0.91061 452514	0.905027 932961	0.905027 932961	0.0055865 9217877	0.816201 117318	0.270248 473415

LogisticRegression with ~X1

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.128491 620112	0.033519 5530726	0.145251 396648	0.078212 2905028	0.111731 843575	0.128491 620112	0.055865 9217877	0.078212 2905028	0.117318 435754	0.039106 1452514	0.091620 1117318	0.037988 8268156

LogisticRegression with ~X2

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.0502793 296089	0.0	0.0558659 217877	0.0335195 530726	0.0279329 608939	0.0167597 765363	0.0055865 9217877	0.0111731 843575	0.0391061 452514	0.0	0.0240223 463687	0.0193605 848616

LogisticRegression with ~X3

K=0	K=1	K=2	K=3	K=4	K=5	K=6	K=7	K=8	K=9	Mean	Std
0.128491 620112	0.039106 1452514	0.106145 251397	0.139664 804469	0.072625 698324	0.111731 843575	0.083798 8826816	0.083798 8826816	0.106145 251397	0.011173 1843575	0.088268 1564246	0.037542 5645059