

CS/ECE/ME532 Activity 9

Estimated Time: 25 minutes for P1, 30 minutes for P2

1. Consider the system of linear equations $\mathbf{X}\mathbf{w} = \mathbf{y}$ where $\mathbf{X} = \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix}$, $\mathbf{w} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$ and $\mathbf{y} = \begin{bmatrix} 2 \\ -4 \end{bmatrix}$.

- a) Sketch the set of all \mathbf{w} that satisfy $\mathbf{X}\mathbf{w} = \mathbf{y}$ in the w_1 - w_2 plane. Is the solution unique? What is the value of the squared error $\min_{\mathbf{w}} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$?
- b) Use your sketch to find the \mathbf{w} of minimum norm that satisfies the system of equations: $\min_{\mathbf{w}} \|\mathbf{w}\|_2^2$ subject to $\mathbf{X}\mathbf{w} = \mathbf{y}$. Is this solution unique? What makes it unique? What is the value of the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ at this solution? What is the value of $\|\mathbf{w}\|_2^2$? *Hint:* The equation $\|\mathbf{w}\|_2^2 = c$ describes a circle in \mathbb{R}^2 with radius \sqrt{c} .
- c) Algebraically find the $\hat{\mathbf{w}}$ that solves the Tikhonov-regularized (or ridge regression) problem $\hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \{\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2\}$ as a function of λ . *Hint:* Recall that

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

- d) Sketch the set solution to the Tikhonov-regularized problem in the w_1 - w_2 plane as a function of λ for $0 < \lambda < \infty$. (Consider the solution for different values of λ in that range.) Find the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ and norm squared of the solution, $\|\mathbf{w}\|_2^2$ for $\lambda = 0$ and $\lambda = 5$. Compare the squared error and norm squared of the solution to those in part b).

2. Let $\mathbf{X} = \begin{bmatrix} 1 & \gamma \\ 1 & -\gamma \\ 1 & -\gamma \\ 1 & \gamma \end{bmatrix}$.

- a) Show that the columns of \mathbf{X} are orthogonal to each other for any γ .
- b) Express $\mathbf{X} = \mathbf{U}\mathbf{\Sigma}$ where \mathbf{U} is a 4-by-2 matrix with orthonormal columns and $\mathbf{\Sigma}$ is a 2-by-2 diagonal matrix (the non-diagonal entries are zero).

- c) Express the solution to the least-squares problem $\min_{\mathbf{w}} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ as a function of \mathbf{U} , $\mathbf{\Sigma}$, and \mathbf{y} .

- d) Let $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Find the weights \mathbf{w} as a function of γ . What happens to $\|\mathbf{w}\|_2^2$ as $\gamma \rightarrow 0$?

- e) The ratio of the largest to the smallest diagonal values in $\mathbf{\Sigma}$ is termed the condition number of \mathbf{X} . Find the condition number if $\gamma = 0.1$ and $\gamma = 10^{-8}$. Also find $\|\mathbf{w}\|_2^2$ for these two values of γ .

- f) A system of linear equations with a large condition number is said to be “ill-conditioned”. One consequence of an ill-conditioned system of equations is solutions with large norms as you found in the previous part of this problem. A second consequence is that the solution is very sensitive to small errors in \mathbf{y} such as may

result from measurement error or numerical error. Suppose $\mathbf{y} = \begin{bmatrix} 1 + \epsilon \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Write

$\mathbf{w} = \mathbf{w}_o + \mathbf{w}_\epsilon$ where \mathbf{w}_o is the solution for arbitrary γ when $\epsilon = 0$ and \mathbf{w}_ϵ is the perturbation in that solution due to some error $\epsilon \neq 0$. How does the norm of the perturbation due to $\epsilon \neq 0$, $\|\mathbf{w}_\epsilon\|_2^2$, depend on the condition number? Find $\|\mathbf{w}_\epsilon\|_2^2$ for $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$.

- g) Now apply ridge regression, i.e., Tikhonov regularization. Solve for \mathbf{w}_o and \mathbf{w}_ϵ as a function of λ . Find $\|\mathbf{w}_o\|_2^2$ and $\|\mathbf{w}_\epsilon\|_2^2$ for $\lambda = 0.1$, $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$. Comment on the impact of regularization.

$$\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2$$

532 Activity 9 - DEVIN BRESSER

1. Consider the system of linear equations $\mathbf{X}\mathbf{w} = \mathbf{y}$ where $\mathbf{X} = \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix}$, $\mathbf{w} =$

$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \text{ and } \mathbf{y} = \begin{bmatrix} 2 \\ -4 \end{bmatrix}.$$

- Sketch the set of all \mathbf{w} that satisfy $\mathbf{X}\mathbf{w} = \mathbf{y}$ in the w_1 - w_2 plane. Is the solution unique? What is the value of the squared error $\min_{\mathbf{w}} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$?
- Use your sketch to find the \mathbf{w} of minimum norm that satisfies the system of equations: $\min_{\mathbf{w}} \|\mathbf{w}\|_2^2$ subject to $\mathbf{X}\mathbf{w} = \mathbf{y}$. Is this solution unique? What makes it unique? What is the value of the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ at this solution? What is the value of $\|\mathbf{w}\|_2^2$? *Hint:* The equation $\|\mathbf{w}\|_2^2 = c$ describes a circle in \mathbb{R}^2 with radius \sqrt{c} .
- Algebraically find the $\hat{\mathbf{w}}$ that solves the Tikhonov-regularized (or ridge regression) problem $\hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \{\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2\}$ as a function of λ . *Hint:* Recall that

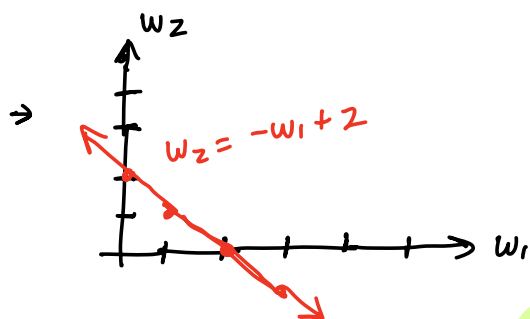
$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

- Sketch the set solution to the Tikhonov-regularized problem in the w_1 - w_2 plane as a function of λ for $0 < \lambda < \infty$. (Consider the solution for different values of λ in that range.) Find the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ and norm squared of the solution, $\|\mathbf{w}\|_2^2$ for $\lambda = 0$ and $\lambda = 5$. Compare the squared error and norm squared of the solution to those in part b).

$$a.) \quad \underline{\mathbf{X}} \underline{\mathbf{w}} = \underline{\mathbf{y}} = \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} 2 \\ -4 \end{bmatrix}$$

$$\rightarrow w_1 + w_2 = 2 \rightarrow w_2 = 2 - w_1$$

$$-2w_1 - 2w_2 = -4 \rightarrow 2w_2 = 4 - 2w_1 \rightarrow w_2 = 2 - w_1$$

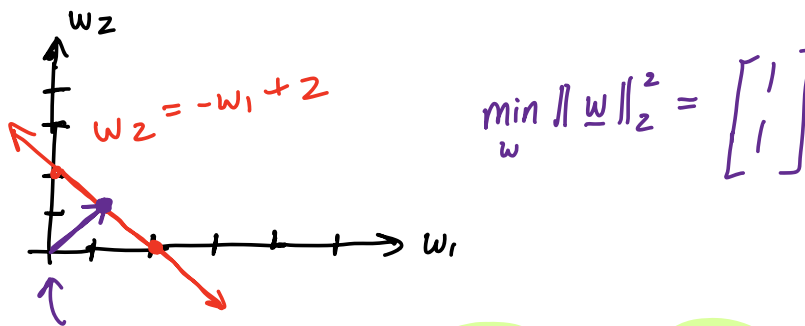


solution is nonunique, all $\begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$ so that $w_2 = 2 - w_1$ satisfy $\underline{\mathbf{X}} \underline{\mathbf{w}} = \underline{\mathbf{y}}$

$$\min \|\underline{\mathbf{X}} \underline{\mathbf{w}} - \underline{\mathbf{y}}\|_2^2 = 0 \quad \text{for each solution}$$

- b) Use your sketch to find the \mathbf{w} of minimum norm that satisfies the system of equations: $\min_{\mathbf{w}} \|\mathbf{w}\|_2^2$ subject to $\mathbf{X}\mathbf{w} = \mathbf{y}$. Is this solution unique? What makes it unique? What is the value of the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ at this solution? What is the value of $\|\mathbf{w}\|^2$? *Hint:* The equation $\|\mathbf{w}\|_2^2 = c$ describes a circle in \mathbb{R}^2 with radius \sqrt{c} .
- c) Algebraically find the $\hat{\mathbf{w}}$ that solves the Tikhonov-regularized (or ridge regression) problem $\hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \{\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2\}$ as a function of λ . *Hint:* Recall that

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$



$\underline{\mathbf{w}} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is a unique solution for $\min_{\mathbf{w}} \|\underline{\mathbf{w}}\|_2^2$.

It is the single point along the solution line that is aimed at the origin orthogonal to the line (intersects min(ℓ_2 norm) circle)

The value of $\|\mathbf{X}\underline{\mathbf{w}} - \mathbf{y}\|_2^2 = 0$ at this solution, because it is an exact solution.

$$\|\underline{\mathbf{w}}\|_2^2 = \sqrt{(1^2) + (1^2)}^2 = 2$$

- c) Algebraically find the $\hat{\mathbf{w}}$ that solves the Tikhonov-regularized (or ridge regression) problem $\hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \{ \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2 \}$ as a function of λ . *Hint: Recall that*

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

- d) Sketch the set solution to the Tikhonov-regularized problem in the w_1 - w_2 plane as a function of λ for $0 < \lambda < \infty$. (Consider the solution for different values of λ in that range.) Find the squared error $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ and norm squared of the solution, $\|\mathbf{w}\|_2^2$ for $\lambda = 0$ and $\lambda = 5$. Compare the squared error and norm squared of the solution to those in part b).

$$c) \quad \hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \left\{ \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2 \right\} \quad (\text{check})$$

$$\rightarrow \text{fact: } \|\underline{a}\|_2^2 + \|\underline{b}\|_2^2 = \underline{a}^T \underline{a} + \underline{b}^T \underline{b} = \left\| \begin{bmatrix} \underline{a} \\ \underline{b} \end{bmatrix} \right\|_2^2$$

$$\rightarrow \left\| \begin{bmatrix} \mathbf{X}\mathbf{w} - \mathbf{y} \\ \lambda^{1/2} \mathbf{w} \end{bmatrix} \right\|_2^2 \rightarrow \min_{\mathbf{w}} \left\| \begin{bmatrix} \mathbf{X} \\ \lambda^{1/2} \mathbf{I} \end{bmatrix} \underset{\tilde{\mathbf{X}}}{\mathbf{w}} - \underset{\tilde{\mathbf{y}}}{\begin{bmatrix} \mathbf{y} \\ 0 \end{bmatrix}} \right\|_2^2$$

$$\rightarrow \min_{\mathbf{w}} \|\tilde{\mathbf{X}}\mathbf{w} - \tilde{\mathbf{y}}\| = (\tilde{\mathbf{X}}^T \tilde{\mathbf{X}})^{-1} \tilde{\mathbf{X}}^T \tilde{\mathbf{y}} =$$

$$= \left(\begin{bmatrix} \mathbf{X}^T & \lambda^{1/2} \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ \lambda^{1/2} \mathbf{I} \end{bmatrix} \right)^{-1} \begin{bmatrix} \mathbf{X}^T & \lambda^{1/2} \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{y} \\ 0 \end{bmatrix}$$

$$\mathbf{w}_{\min} = \left(\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I} \right)^{-1} \mathbf{X}^T \mathbf{y}$$

$$\mathbf{w}_{\min} = \left(\begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix} + \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} 2 \\ -4 \end{bmatrix}$$

$$= \begin{pmatrix} 5+\lambda & 5 \\ 5 & 5+\lambda \end{pmatrix}^{-1} \begin{bmatrix} 10 \\ 10 \end{bmatrix}$$

$$= \frac{1}{(5+\lambda)^2 - 25} \begin{bmatrix} 5+\lambda & -5 \\ -5 & 5+\lambda \end{bmatrix} \begin{bmatrix} 10 \\ 10 \end{bmatrix}$$

$$\cancel{10\lambda} + 10\lambda + \lambda^2 \cancel{625}$$

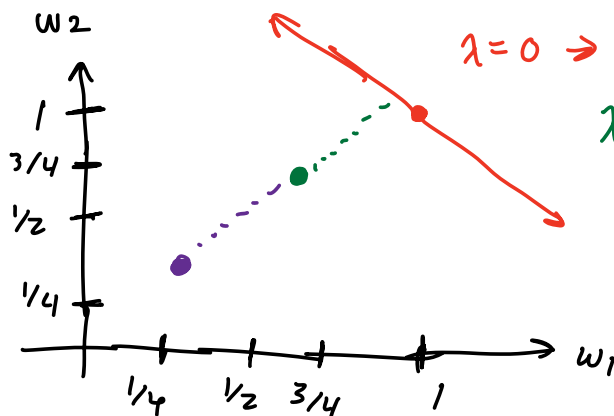
$$= \frac{1}{10\lambda + \lambda^2} \begin{bmatrix} 5+\lambda & -5 \\ -5 & 5+\lambda \end{bmatrix} \begin{bmatrix} 10 \\ 10 \end{bmatrix}$$

$$\frac{10}{10+\lambda} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$= \frac{1}{10\lambda + \lambda^2} \begin{bmatrix} 10\lambda \\ 10\lambda \end{bmatrix}$$

$$= \frac{10\lambda}{10\lambda + \lambda^2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{\lambda(10)}{\lambda(10+\lambda)} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{10}{10+\lambda} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

d) Sketch the set solution to the Tikhonov-regularized problem in the w_1 - w_2 plane as a function of λ for $0 < \lambda < \infty$. (Consider the solution for different values of λ in that range.) Find the squared error $\|X\mathbf{w} - \mathbf{y}\|_2^2$ and norm squared of the solution, $\|\mathbf{w}\|_2^2$ for $\lambda = 0$ and $\lambda = 5$. Compare the squared error and norm squared of the solution to those in part b).



$$\lambda = 0 \Rightarrow \underline{w} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\lambda = 5 \Rightarrow \underline{w} = \begin{bmatrix} 2/3 \\ 2/3 \end{bmatrix}$$

$$\lambda = 20 \Rightarrow \underline{w} = \begin{bmatrix} 1/3 \\ 1/3 \end{bmatrix}$$

$$\underline{\lambda} = 0:$$

$$\|X\underline{w} - \underline{y}\|_2^2 = 0$$

$$\|\underline{w}\|_2^2 = 2$$

$$\underline{\lambda} = 5:$$

$$\|X\underline{w} - \underline{y}\|_2^2 = 2.222$$

$$\|\underline{w}\|_2^2 = 0.888$$

2. Let $\mathbf{X} = \begin{bmatrix} 1 & \gamma \\ 1 & -\gamma \\ 1 & -\gamma \\ 1 & \gamma \end{bmatrix}$.

a) Show that the columns of \mathbf{X} are orthogonal to each other for any γ .

b) Express $\mathbf{X} = \mathbf{U}\mathbf{\Sigma}$ where \mathbf{U} is a 4-by-2 matrix with orthonormal columns and $\mathbf{\Sigma}$ is a 2-by-2 diagonal matrix (the non-diagonal entries are zero).

a.) $\begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix} \beta \begin{bmatrix} \gamma \\ -\gamma \\ -\gamma \\ \gamma \end{bmatrix} = \beta\gamma - \beta\gamma - \beta\gamma + \beta\gamma = 0$
for all β
So $\underline{x}_1, \underline{x}_2$ are orthogonal

b.) $\underline{X} = \underline{U} \underline{\Sigma}$

$$\begin{bmatrix} 1/2 & 1/2 \\ 1/2 & -1/2 \\ 1/2 & -1/2 \\ 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 2\gamma \end{bmatrix} = \begin{bmatrix} 1 & \gamma \\ 1 & -\gamma \\ 1 & -\gamma \\ 1 & \gamma \end{bmatrix}$$

$\nwarrow \underline{U}$
 $\nwarrow \underline{\Sigma}$
 $\nwarrow \underline{X}$

c) Express the solution to the least-squares problem $\min_w \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ as a function of \mathbf{U} , $\mathbf{\Sigma}$, and \mathbf{y} .

$$\begin{aligned} \min_w \|\underline{X} \underline{w} - \underline{y}\|_2^2 &= \min_w \|\underline{U} \underline{\Sigma} \underline{w} - \underline{y}\|_2^2 \\ &= (\underline{X}^T \underline{X})^{-1} \underline{X}^T \underline{y} = ((\underline{U} \underline{\Sigma})^T \underline{X})^{-1} (\underline{U} \underline{\Sigma})^T \underline{y} \\ &= (\underline{\Sigma}^T \underline{U}^T \underline{U} \underline{\Sigma})^{-1} \underline{\Sigma}^T \underline{U}^T \underline{y} \\ &= (\underline{\Sigma}^T \underline{\Sigma})^{-1} \underline{\Sigma}^T \underline{U}^T \underline{y} \end{aligned}$$

c) Express the solution to the least-squares problem $\min_{\mathbf{w}} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ as a function of \mathbf{U} , Σ , and \mathbf{y} .

d) Let $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Find the weights \mathbf{w} as a function of γ . What happens to $\|\mathbf{w}\|_2^2$ as $\gamma \rightarrow 0$?

$$d.) \underline{\mathbf{w}}_{\min} = (\underline{\mathbf{E}}^T \underline{\mathbf{E}})^{-1} \underline{\mathbf{E}} \underline{\mathbf{U}}^T \mathbf{y} \quad \rightarrow (\underline{\mathbf{E}}^T \underline{\mathbf{E}})^{-1} = \begin{bmatrix} 1/4 & 0 \\ 0 & 1/\gamma^2 \end{bmatrix}$$

$$\rightarrow \underbrace{(\underline{\mathbf{E}}^T \underline{\mathbf{E}})^{-1} \underline{\mathbf{E}} \underline{\mathbf{U}}^T}_{2 \times 4} = \begin{bmatrix} 0.25 & 0.25 & 0.25 & 0.25 \\ 0.25/\gamma & -0.25/\gamma & -0.25/\gamma & 0.25/\gamma \end{bmatrix}$$

$$\Rightarrow \quad \quad \cdot \mathbf{y} = \begin{bmatrix} 0.5 \\ 0.5/\gamma \end{bmatrix}$$

$$\rightarrow \underline{\mathbf{w}}_{\min} = \begin{bmatrix} 0.5 \\ 0.5/\gamma \end{bmatrix}$$

$$\|\underline{\mathbf{w}}_{\min}\|_2^2 = \sqrt{(0.5)^2 + (0.5/\gamma)^2}^2 = 0.25 + 0.25/\gamma^2$$

$$\text{As } \gamma \rightarrow 0, \|\underline{\mathbf{w}}_{\min}\|_2^2 \rightarrow \infty$$

e) The ratio of the largest to the smallest diagonal values in Σ is termed the condition number of \mathbf{X} . Find the condition number if $\gamma = 0.1$ and $\gamma = 10^{-8}$. Also find $\|\mathbf{w}\|_2^2$ for these two values of γ .

f) A system of linear equations with a large condition number is said to be "ill-conditioned". One consequence of an ill-conditioned system of equations is solutions with large norms as you found in the previous part of this problem. A second consequence is that the solution is very sensitive to small errors in \mathbf{y} such as may

result from measurement error or numerical error. Suppose $\mathbf{y} = \begin{bmatrix} 1 + \epsilon \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Write

$\mathbf{w} = \mathbf{w}_o + \mathbf{w}_\epsilon$ where \mathbf{w}_o is the solution for arbitrary γ when $\epsilon = 0$ and \mathbf{w}_ϵ is the perturbation in that solution due to some error $\epsilon \neq 0$. How does the norm of the perturbation due to $\epsilon \neq 0$, $\|\mathbf{w}_\epsilon\|_2^2$, depend on the condition number? Find $\|\mathbf{w}_\epsilon\|_2^2$ for $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$.

g) Now apply ridge regression, i.e., Tikhonov regularization. Solve for \mathbf{w}_o and \mathbf{w}_ϵ as a function of λ . Find $\|\mathbf{w}_o\|_2^2$ and $\|\mathbf{w}_\epsilon\|_2^2$ for $\lambda = 0.1$, $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$. Comment on the impact of regularization.

$$e.) \quad \underline{\Sigma} = \begin{bmatrix} 2 & 0 \\ 0 & 2\gamma \end{bmatrix}$$

$$\underline{\Sigma}_{\gamma=0.1} = \begin{bmatrix} 2 & 0 \\ 0 & 0.2 \end{bmatrix} \rightarrow \text{condition \# } 2/0.2 = 10$$

$$\|\underline{w}_{\min}\|_2^2_{\gamma=0.1} = 0.25 + 0.25/0.1^2 = 25.25$$

$$\underline{\Sigma}_{\gamma=10^{-8}} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \cdot 10^{-8} \end{bmatrix} \rightarrow \text{condition \# } 2/2 \cdot 10^{-8} = 1 \cdot 10^8$$

$$\|\underline{w}_{\min}\|_2^2_{\gamma=0.1} = 0.25 + 0.25/(10^{-8})^2 = 2.5 \cdot 10^{15}$$

- f) A system of linear equations with a large condition number is said to be "ill-conditioned". One consequence of an ill-conditioned system of equations is solutions with large norms as you found in the previous part of this problem. A second consequence is that the solution is very sensitive to small errors in \mathbf{y} such as may

result from measurement error or numerical error. Suppose $\mathbf{y} = \begin{bmatrix} 1+\epsilon \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Write

$\mathbf{w} = \mathbf{w}_0 + \mathbf{w}_\epsilon$ where \mathbf{w}_0 is the solution for arbitrary γ when $\epsilon = 0$ and \mathbf{w}_ϵ is the perturbation in that solution due to some error $\epsilon \neq 0$. How does the norm of the perturbation due to $\epsilon \neq 0$, $\|\mathbf{w}_\epsilon\|_2^2$, depend on the condition number? Find $\|\mathbf{w}_\epsilon\|_2^2$ for $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$.

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$$\mathbf{y} = \begin{bmatrix} 1+\epsilon \\ 0 \\ 0 \\ 1 \end{bmatrix} \xrightarrow{\text{calculator}} \underline{\mathbf{w}} = \begin{bmatrix} 0.5 + 0.25\epsilon \\ \frac{0.5+0.25\epsilon}{\gamma} \end{bmatrix} = \begin{bmatrix} 0.5 \\ 0.5/\gamma \end{bmatrix} + \begin{bmatrix} 0.25\epsilon \\ 0.25\epsilon/\gamma \end{bmatrix}$$

$\underline{\mathbf{w}}_0 \quad \underline{\mathbf{w}}_\epsilon$

$$\|\underline{\mathbf{w}}_\epsilon\|_2^2 = \sqrt{(0.25\epsilon)^2 + (0.25\epsilon/\gamma)^2}^2 = 0.125\epsilon^2 + 0.125\epsilon^2/\gamma^2$$

As γ increases $\gg 2$, condition # decreases & $\|\underline{\mathbf{w}}_\epsilon\|_2^2 \rightarrow 0.125\epsilon^2$
 As γ decreases $\ll 2$, condition # increases & $\|\underline{\mathbf{w}}_\epsilon\|_2^2 \rightarrow \infty$

$$\|\underline{\mathbf{w}}_\epsilon\|_2^2 [\epsilon=0.01, \gamma=0.1] = 0.125(0.01)^2 + 0.125(0.01)^2/(0.1)^2$$

$$= 0.001263$$

$$\|\underline{\mathbf{w}}_\epsilon\|_2^2 [\epsilon=0.01, \gamma=10^{-8}] = \dots = 1.25 \cdot 10^{11}$$

g) Now apply ridge regression, i.e., Tikhonov regularization. Solve for \underline{w}_o and \underline{w}_e as a function of λ . Find $\|\underline{w}_o\|_2^2$ and $\|\underline{w}_e\|_2^2$ for $\lambda = 0.1$, $\epsilon = 0.01$ and $\gamma = 0.1$ and $\gamma = 10^{-8}$. Comment on the impact of regularization.

$$\underline{w}_{\min} = \begin{bmatrix} 0.5 + 0.25\epsilon \\ \frac{0.5 + 0.25\epsilon}{\gamma} \end{bmatrix} \quad \text{when} \quad \underline{y} = \begin{bmatrix} 1 + \epsilon \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\min_{\underline{w}} \quad \|\underline{X}\underline{w} - \underline{y}\|_2^2 + \lambda \|\underline{w}\|_2^2$$

$$\Rightarrow \underline{w}_{\min} = (\underline{X}^T \underline{X} + \lambda \underline{I})^{-1} \underline{X}^T \underline{y} \quad \text{known result.}$$

$$\Rightarrow \left. \begin{array}{l} \lambda = 0.1 \\ \epsilon = 0.01 \\ \gamma = 0.1 \end{array} \right\} \xrightarrow{\text{calculator}} \underline{w}_o = \begin{bmatrix} 0.4902 \\ 1.436 \end{bmatrix}$$

$$\underline{w}_e = \begin{bmatrix} 0.487 \\ 1.428 \end{bmatrix}, \quad \|\underline{w}_e\|_2^2 = 2.27$$

$$\left. \begin{array}{l} \lambda = 0.1 \\ \epsilon = 0.01 \\ \gamma = 10^{-8} \end{array} \right\} \xrightarrow{\text{calculator}} \underline{w}_{\min} = \begin{bmatrix} 0.4902 \\ 2.01 \cdot 10^{-7} \end{bmatrix}$$

$$\underline{w}_e = \begin{bmatrix} 0.487 \\ 2 \cdot 10^{-7} \end{bmatrix}, \quad \|\underline{w}_e\|_2^2 = 0.237$$

Adding

the λ term reduces \underline{w}_e , the perturbation due to errors ϵ .