

2.4.1

a) The performance of flexible learning would be better if n is large and p is small because it would fit the data more accurately and result in a better fit

b) When p is large and n is small the flexible learning model would overfit the data and thus would a poor model than inflexible

c) flexible learning method would fit the data more accurately because of more degrees of freedom and thus would be a better fit

d) Flexible methods would be worse if $\text{var}(\epsilon)$ is large because it would overfit and fit the noise in this case instead of generalizing and being robust to noise

2.4.7

a)	Obs	X_1	X_2	X_3
	1	0	3	0
	2	2	0	0
	3	0	1	3
	4	0	1	2
	5	-1	0	1
	6	1	1	7

Euclidean distn
$\sqrt{3^2} = 3$
$\sqrt{2^2} = 2$
$\sqrt{10} = 3.2$
$\sqrt{5} = 2.2$
$\sqrt{2} = 1.41$
$\sqrt{3} = 1.7$

- b) for $k=1$ Obs # 5, $y = \text{green}$ would be the prediction because it is the closest neighbour to it
- c) if $k=3$, Obs # 5, #6, #2 would be the closest. Obs #2 and #5 are Red and #6 is green. Therefore prediction would be Red better prediction because it appears in majority
- d) If Bayes decision boundary is non-linear, then $k = \text{small}$ would present a more non-linear boundary than $k = \text{large}$. Thus best value for k would be small