See HW4.pdf for rendered LaTeX equations. Github does not render them in markdown files.

Q1

a)

Symmetric

$$K(x_i,x_j) = \sum_{a=1}^m w_a K_a(x_i,x_j)$$

Since $K_1, ..., K_a$ are valid kernel functions:

$$K(x_i,x_j) = \sum_{a=1}^m w_a K_a(x_j,x_i) riangleq K(x_j,x_i)$$

∴ K is symmetric

Positive Semi-Definite

Using Mercer's Theorem, a Kernel is positive semi-definite if:

$$egin{aligned} & c^T \underline{Kc} \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n c_i c_j K(x_i, x_j) \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n c_i c_j \sum_{a=1}^m w_a K_a(x_i, x_j) \geq 0 \ & \sum_{a=1}^m w_a \sum_{i=1}^n \sum_{j=1}^n c_i c_j \langle \underline{\phi}(x_i), \underline{\phi}(x_j)
angle \geq 0 \ & \sum_{a=1}^m w_a \sum_{i=1}^n \sum_{j=1}^n c_i c_j \sum_{k=1}^p \phi_k(x_i) \phi_k(x_j) \geq 0 \ & \sum_{a=1}^m w_a \sum_{k=1}^p \sum_{i=1}^n \sum_{j=1}^n c_i \phi_k(x_i) c_j \phi_k(x_j) \geq 0 \ & \sum_{a=1}^m w_a \sum_{k=1}^p \left(\sum_{i=1}^n c_i \phi_k(x_i)
ight)^2 \geq 0 \end{aligned}$$

Since $w_i \geq 0, \forall w_i$, The inequality holds, \therefore K is positive semi-definite.

The function is symmetric and positive semi-definite, $\therefore K$ is a Kernel.

b)

Symmetric

$$K(x_i,x_j)=K_1(x_i,x_j)K_2(x_i,x_j)$$

Since K_1 and K_2 are valid kernel functions:

$$K(x_i,x_j) = K_1(x_j,x_i)K_2(x_j,x_i) riangleq K(x_j,x_i)$$

 $ext{:} ext{K is symmetric}$

Positive Semi-Definite

Using Mercer's Theorem, a Kernel is positive semi-definite if:

$$egin{aligned} rac{c^T K c}{\sum i=1} \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n c_i c_j K(x_i, x_j) \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n c_i c_j K_1(x_i, x_j) K_2(x_i, x_j) \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n c_i c_j \langle \phi_1(x_i), \phi_1(x_j) \rangle \langle \phi_2(x_i), \phi_2(x_j) \rangle \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^p c_i c_j \phi_{(1,k)}(x_i) \phi_{(1,k)}(x_j) \phi_{(2,k)}(x_i) \phi_{(2,k)}(x_j) \geq 0 \ & \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^p c_i \phi_{(1,k)}(x_i) \phi_{(2,k)}(x_i) c_j \phi_{(1,k)}(x_j) \phi_{(2,k)}(x_j) \geq 0 \ & \sum_{k=1}^p \left(\sum_{i=1}^n c_i \phi_{(1,k)}(x_i) \phi_{(2,k)}(x_i)
ight)^2 \geq 0 \end{aligned}$$

Inequality holds, : K is positive semi-definite.

The function is symmetric and positive semi-definite, $\therefore K$ is a Kernel.

c)

Symmetric

$$K(x,x')=(xx'+1)^{2015}$$
 $K(x,x')=(x'x+1)^{2015} riangleq K(x',x)$ $extrm{ \therefore K is symmetric}$

Positive Semi-Definite

$$K(x,x') = (xx'+1)^{2015} \ K(x,x') = \sum_{i=1}^{2015} \left(rac{2015!}{i!(2015-i)!}(x)^i(x')^i
ight)$$

Using Mercer's Theorem, a Kernel is positive semi-definite if:

$$c^T K c > 0$$

$$egin{aligned} \sum_{i=1}^n \sum_{j=1}^n c_i c_j K(x_i, x_j) &\geq 0 \ \ \sum_{i=1}^n \sum_{j=1}^n c_i c_j \sum_{k=1}^{2015} \left(rac{2015!}{k!(2015-k)!} (x_i)^k (x_j)^k
ight) &\geq 0 \ \ \sum_{k=1}^{2015} rac{2015!}{k!(2015-k)!} \sum_{i=1}^n \sum_{j=1}^n c_i c_j (x_i)^k (x_j)^k &\geq 0 \ \ \ \sum_{k=1}^{2015} rac{2015!}{k!(2015-k)!} \left(\sum_{i=1}^n c_i (x_i)^k
ight)^2 &\geq 0 \end{aligned}$$

Inequality holds, : K is positive semi-definite.

The function is symmetric and positive semi-definite, $\therefore K$ is a Kernel.

d)

Symmetric

$$K(x,x')=\exp\left(-rac{(x-x')^2}{2}
ight)$$

Since x - x' is being squared, it is also equal to:

$$K(x,x') = \exp\left(-rac{(x'-x)^2}{2}
ight) riangleq K(x',x)$$

 \therefore K is symmetric

Positive Semi-Definite

$$K(x,x') = \exp\left(-rac{(x-x')^2}{2}
ight)$$
 $K(x,x') = \exp\left(-rac{x^2-2xx'+(x')^2}{2}
ight)$
 $K(x,x') = \exp\left(-rac{x^2}{2}+xx'-rac{(x')^2}{2}
ight)$
 $K(x,x') = rac{\exp\left(xx'
ight)}{\exp\left(rac{x^2}{2}
ight)\exp\left(rac{(x')^2}{2}
ight)}$
 $K(x,x') = rac{\sum_{k=0}^{\infty}rac{(xx')^k}{k!}}{\exp\left(rac{x^2}{2}
ight)\exp\left(rac{(x')^2}{2}
ight)}$
 $K(x,x') = \sum_{k=0}^{\infty}rac{(x)^k(x')^k}{k!\exp\left(rac{x^2}{2}
ight)\exp\left(rac{(x')^2}{2}
ight)}$

Using Mercer's Theorem, a Kernel is positive semi-definite if:

$$c^T K c > 0$$

$$egin{aligned} \sum_{i=1}^n \sum_{j=1}^n c_i c_j K(x_i, x_j) &\geq 0 \ \ \sum_{i=1}^n \sum_{j=1}^n c_i c_j \sum_{k=0}^\infty rac{(x_i)^k (x_j)^k}{k! \exp\left(rac{x_i^2}{2}
ight) \exp\left(rac{x_j^2}{2}
ight)} &\geq 0 \ \ \sum_{k=0}^\infty rac{1}{k!} \sum_{i=1}^n \sum_{j=1}^n c_i c_j rac{(x_i)^k}{\exp\left(rac{x_i^2}{2}
ight)} rac{(x_j)^k}{\exp\left(rac{x_j^2}{2}
ight)} &\geq 0 \ \ \ \sum_{k=0}^\infty rac{1}{k!} \left(\sum_{i=1}^n c_i rac{(x_i)^k}{\exp\left(rac{x_i^2}{2}
ight)}
ight)^2 &\geq 0 \end{aligned}$$

Inequality holds, : K is positive semi-definite.

The function is symmetric and positive semi-definite, $\therefore K$ is a Kernel.

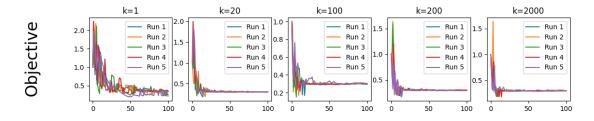
Q2

Summary and Results

 $max_iter = 100$

lambda = 1e5 (A high regularization coefficient was used to stabilize the plots quickly)

Objective vs. Iteration



Iteration

```
Avg runtime for 5 runs with minibatch size of 1: 1.39 sec
Std runtime for 5 runs with minibatch size of 1: 0.03 sec
Avg runtime for 5 runs with minibatch size of 20: 1.48 sec
Std runtime for 5 runs with minibatch size of 20: 0.03 sec
Avg runtime for 5 runs with minibatch size of 100: 1.58 sec
Std runtime for 5 runs with minibatch size of 100: 0.04 sec
Avg runtime for 5 runs with minibatch size of 200: 1.75 sec
Std runtime for 5 runs with minibatch size of 200: 0.03 sec
Avg runtime for 5 runs with minibatch size of 2000: 3.97 sec
Std runtime for 5 runs with minibatch size of 2000: 0.03 sec
```

Code

Code can be found on Github.

Pegasos algorithm:

https://github.com/DeBestTrap/Intro-to-Machine-Learning/blob/main/HW4/pegasos.py

Code to run and plot pegasos algorithm (for instructions to run code see README.md):

https://github.com/DeBestTrap/Intro-to-Machine-Learning/blob/main/HW4/mysgdsvm.py