## Part 1

1. Split: 1, Training index: [ 1 2 3 4 5 7 8 9 10 11 14 15 16 17 18 19],

Validation index: [ 0 6 12 13]

Split: 2, Training index: [ 0 2 3 4 5 6 8 11 12 13 14 15 16 17 18 19],

Validation index: [1 7 9 10]

Split: 3, Training index: [ 0 1 2 4 6 7 8 9 10 11 12 13 14 15 17 18],

Validation index: [3 5 16 19]

Split: 4, Training index: [ 0 1 2 3 4 5 6 7 9 10 12 13 16 17 18 19],

Validation index: [ 8 11 14 15]

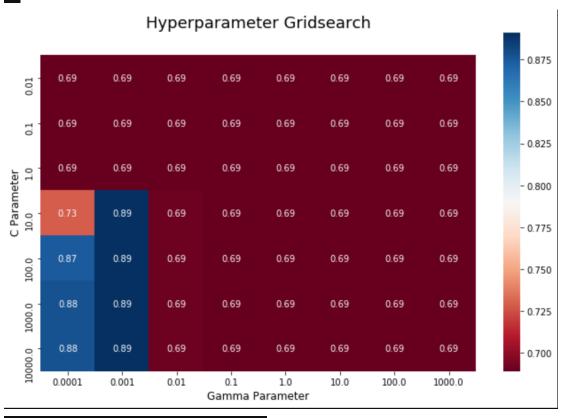
Split: 5, Training index: [ 0 1 3 5 6 7 8 9 10 11 12 13 14 15 16 19],

Validation index: [2 4 17 18]

註:這每次會不一樣,因為是 shuffle 過的

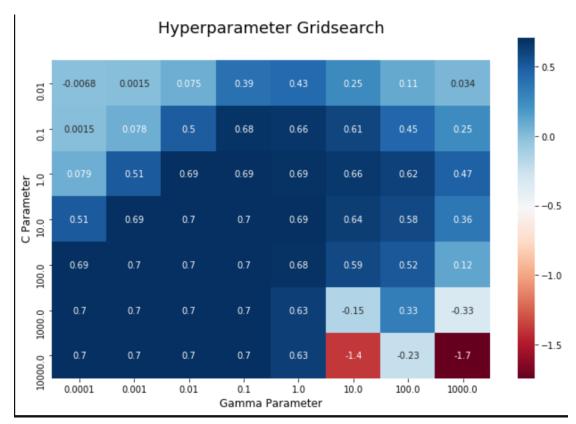
## 2. {'C': 10, 'gamma': 0.001}

3.



4. Accuracy score: 0.8958333333333334

5. {'C': 1000, 'gamma': 0.001}



註: 這裡使用的 score 不是 MSE,是 SVR 裡定義的 score, score = 1-(u/v)

# u = ((y\_true - y\_pred) \*\* 2).sum()

# v = ((y\_true - y\_true.mean()) \*\* 2).sum().

# The best possible score is 1.0(when u is 0) and it can be negative (because the model can be arbitrarily worse).

Square error of Linear regression: 3.4339477869026385

Square error of SVM regresssion model: 3.6002184676788693

分析:這裡不管跑幾次都發現是 Linear regression 的 MSE 比較小

## Part II

$$k(x, x') = ck, (x, x')$$

$$= c(\phi(x)^{T}\phi(x'))$$

$$= (Jc\phi(x))^{T}(Jc\phi(x'))$$

$$k(x, x') = f(x) k, (x, x') f(x')$$

$$= f(x) (\phi(x)^{T}\phi(x')) f(x')$$

$$= (f(x)\phi(x))^{T}((f(x')\phi(x)))$$