

Pipeline

February 24, 2019

1 Kernel methods for biological sequence classification

MVA 2019 - Kernel methods for machine learning

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```
In [1]: %load_ext autoreload
        %autoreload 2
```

```
In [2]: import numpy as np
```

```
In [3]: import backend
        from kernels import LinearKernel, GaussianKernel, GramCSVKernel, FeatureCSVKernel, Mul-
```

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Kernel computed on the fly on imported data

```
In [4]: kernels1a = LinearKernel().load("mat100", indices=[0, 1, 2])
        kernels1b = GaussianKernel(0.1).load("mat100", indices=[0, 1, 2])
```

Linear kernel computed on the fly on imported features

```
In [5]: kernels2a = FeatureCSVKernel("SpectralKernel6", {}).load("spectr6", indices=[0, 1, 2])
        kernels2b = FeatureCSVKernel("SpectralKernel3", {}).load("spectr3", indices=[0, 1, 2])
        kernels2c = FeatureCSVKernel("TranslatedKernel", {}).load("trans6", indices=[0, 1, 2])
```

Precomputed kernel defined from stored Gram matrices

```
In [6]: kernels3 = GramCSVKernel("SubstringKernel", {}).load("substring4_0.7", indices=[0])
```

Parameter tuning

```
In [7]: lambdas = np.logspace(-3, 2, 50)
```

```

In [8]: three_kernels = []
        three_lambdas = []

        use_multiple_kernel = False

        for d in [0, 1, 2]:

            kernels_to_combine_or_compare = [
                kernels1a[d],
                # kernels1b[d],
                kernels2a[d],
                kernels2b[d],
                kernels2c[d]
            ]
            if d == 0:
                kernels_to_combine_or_compare.append(kernels3[0])

            if use_multiple_kernel:

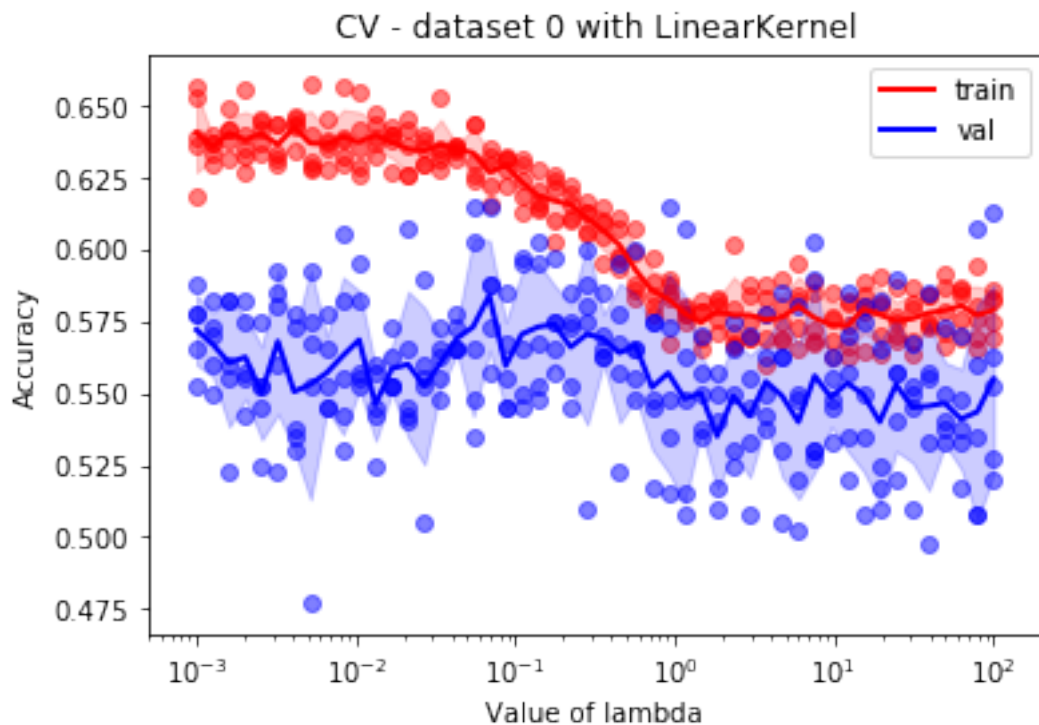
                best_kernel = MultipleKernel(
                    kernels_to_combine_or_compare,
                    grad_step=5e-1, iterations=10)
                best_lambda = backend.tune_parameters(
                    [best_kernel], lambdas,
                    plot=True, result="best_lambdas")[0]

            else:
                best_kernel, best_lambda = backend.tune_parameters(
                    kernels_to_combine_or_compare, lambdas,
                    plot=True, result="best_kernel_lambda")

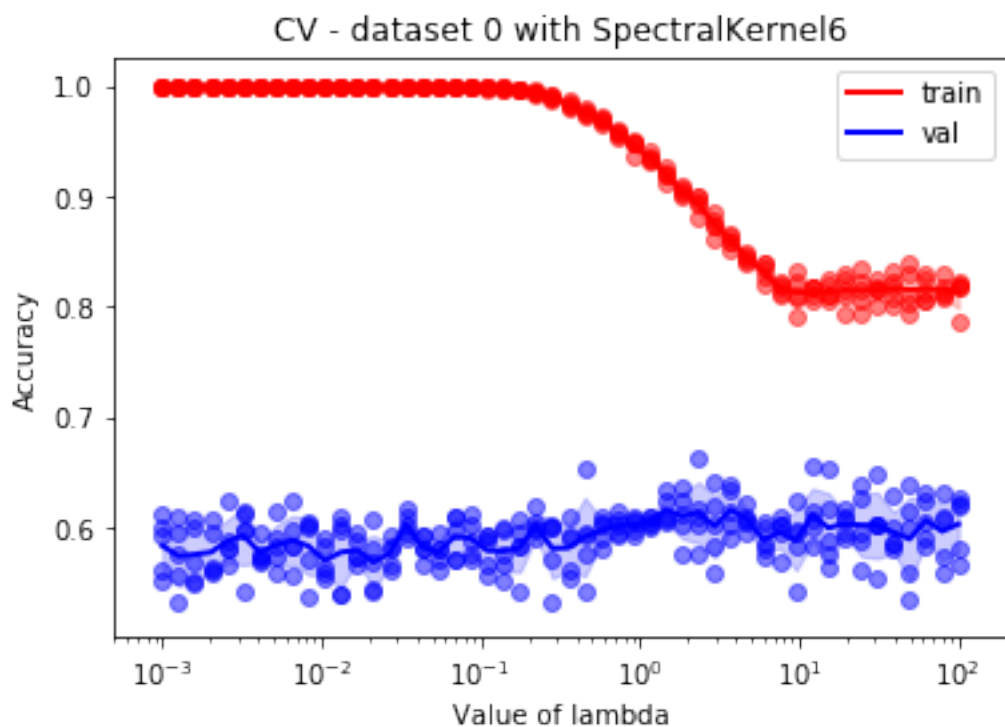
            three_kernels.append(best_kernel)
            three_lambdas.append(best_lambda)

```

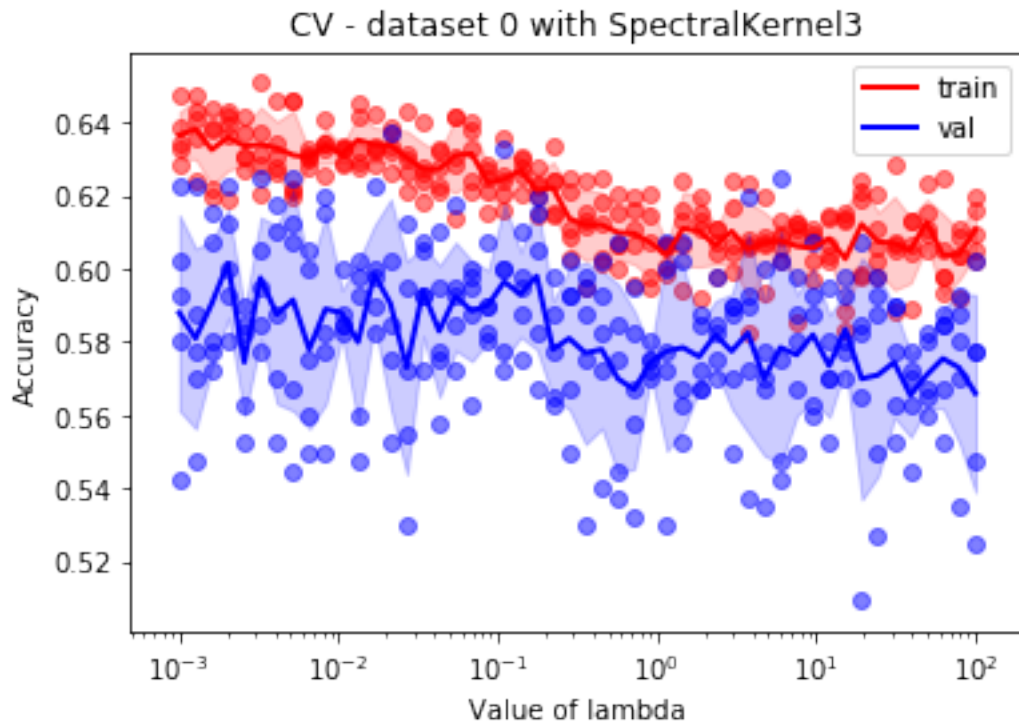
Tuning lambda on dataset 0 with kernel LinearKernel and params {'suffix': 'mat100'}: 100%| 50,



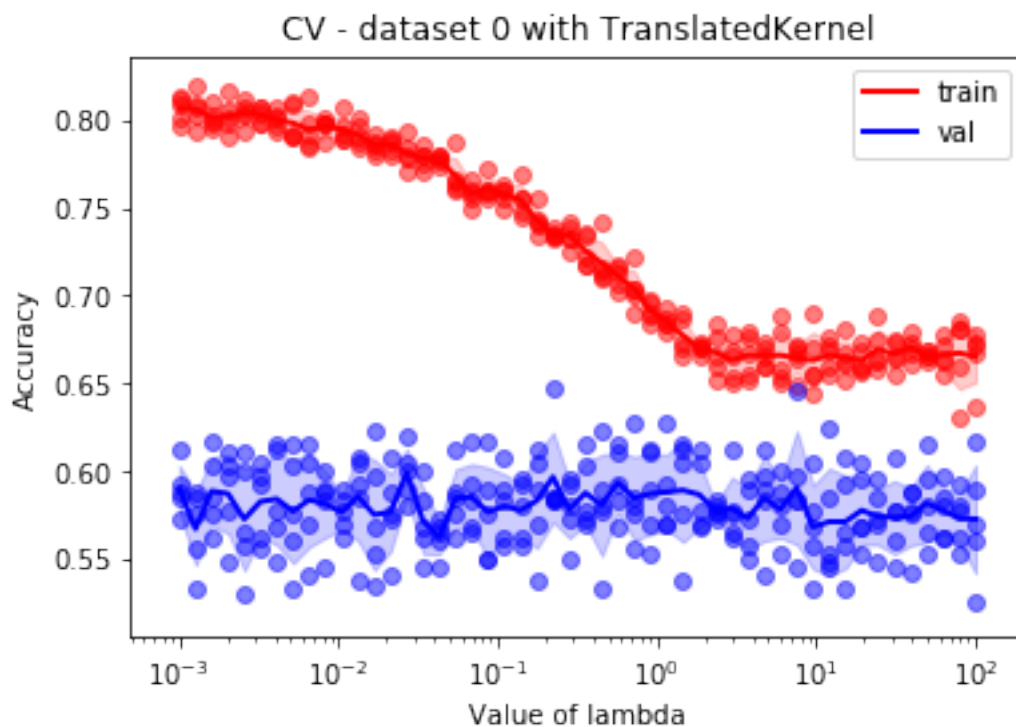
Tuning lambda on dataset 0 with kernel SpectralKernel6 and params {'suffix': 'spectr6'}: 100%



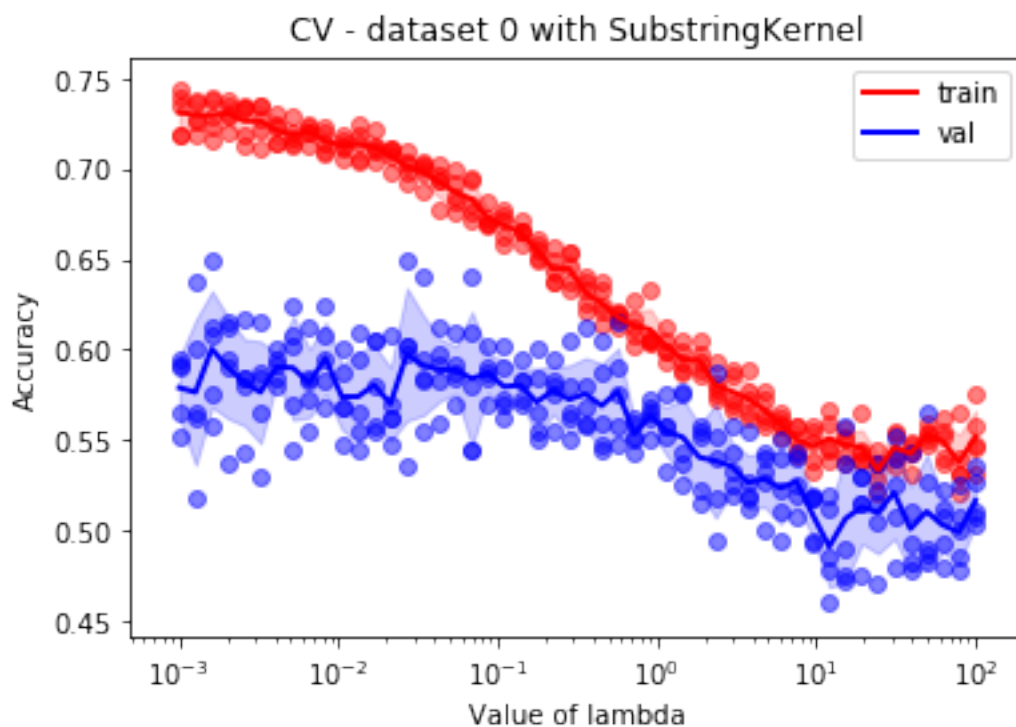
Tuning lambda on dataset 0 with kernel SpectralKernel3 and params {'suffix': 'spectr3'}: 100%



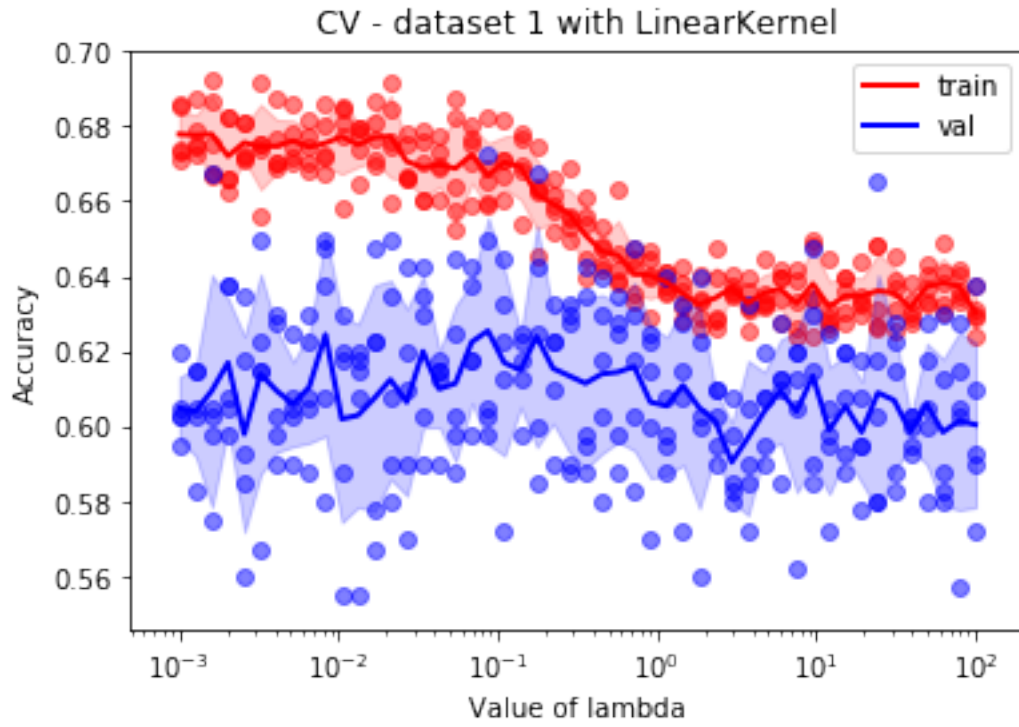
Tuning lambda on dataset 0 with kernel TranslatedKernel and params {'suffix': 'trans6'}: 100%



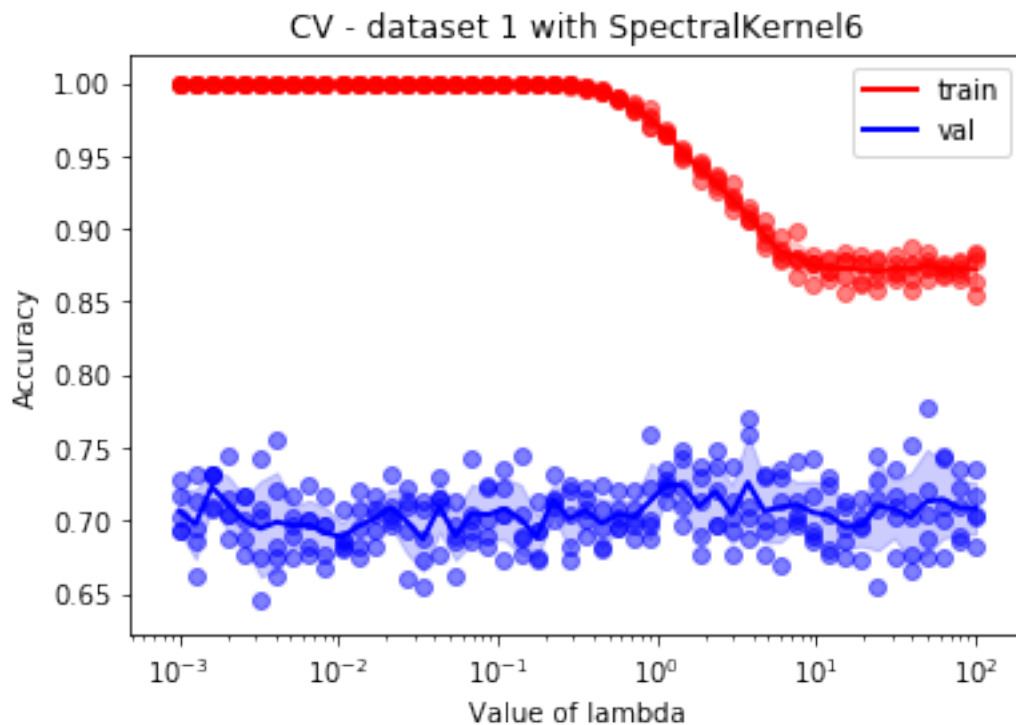
Tuning lambda on dataset 0 with kernel SubstringKernel and params {'suffix': 'substring4_0.7'}



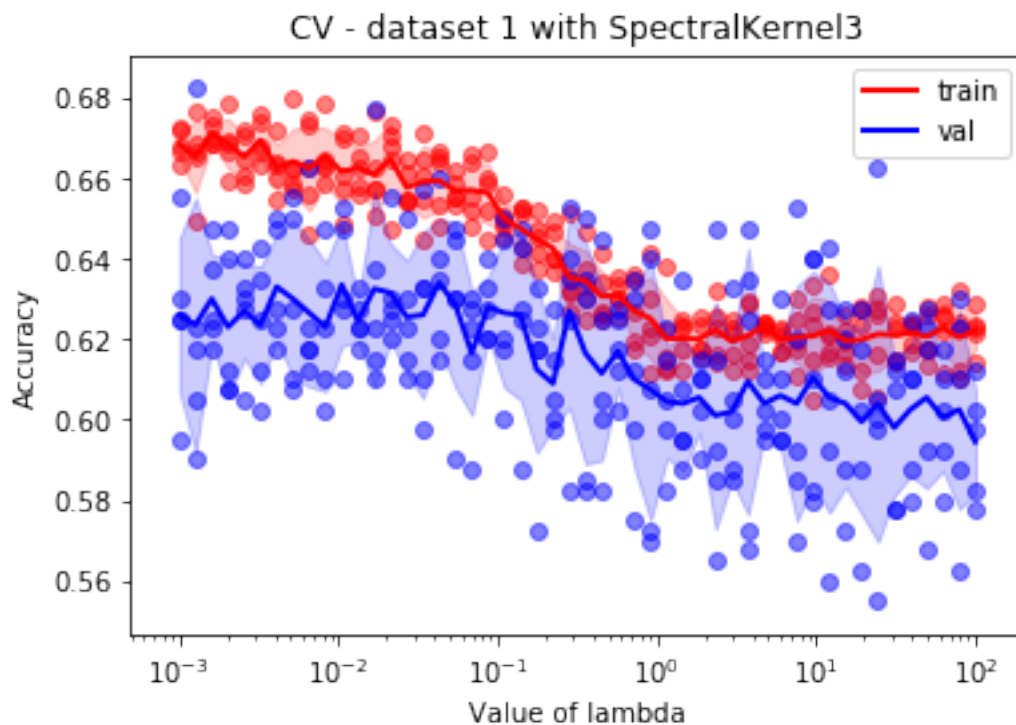
Tuning lambda on dataset 1 with kernel LinearKernel and params {'suffix': 'mat100'}: 100%|| 50%



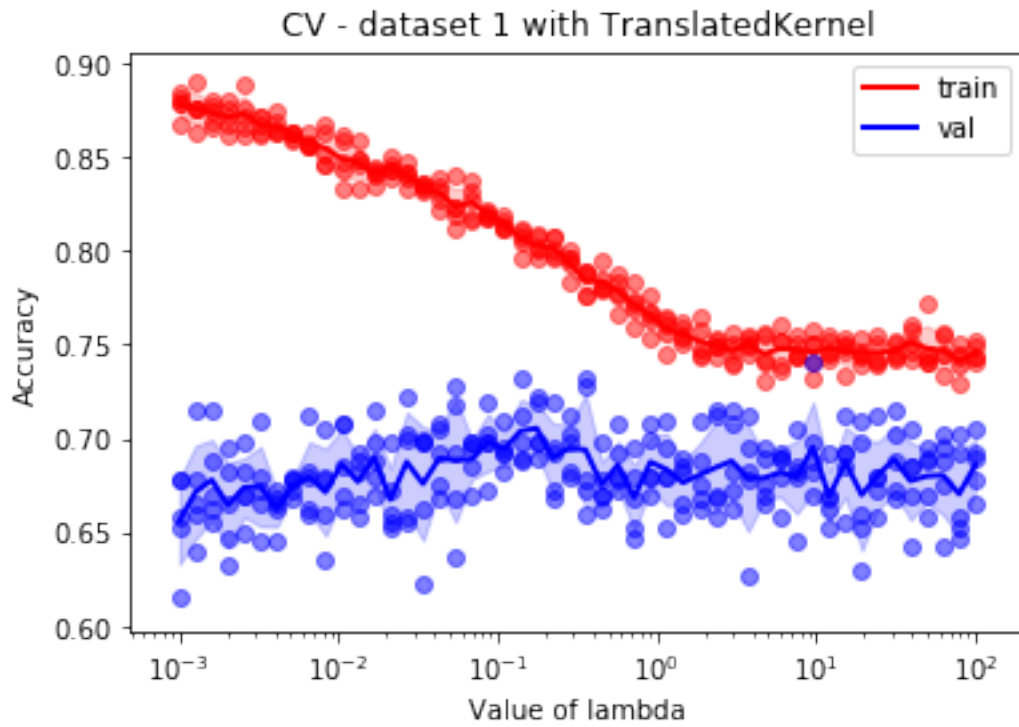
Tuning lambda on dataset 1 with kernel SpectralKernel6 and params {'suffix': 'spectr6'}: 100%|



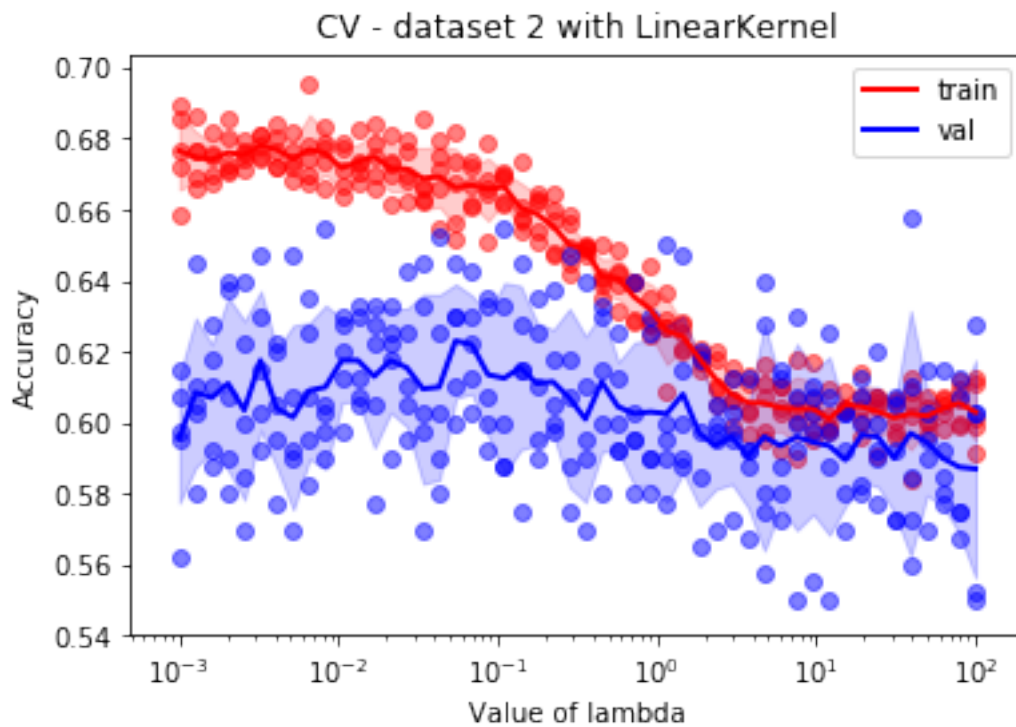
Tuning lambda on dataset 1 with kernel SpectralKernel3 and params {'suffix': 'spectr3'}: 100%



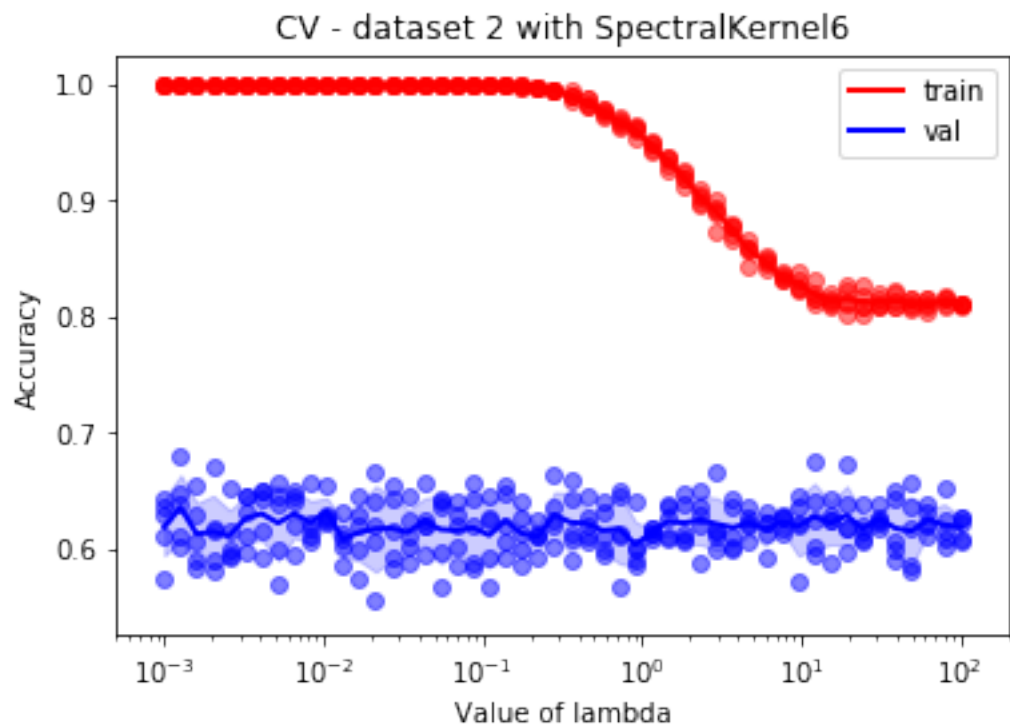
Tuning lambda on dataset 1 with kernel TranslatedKernel and params {'suffix': 'trans6'}: 100%



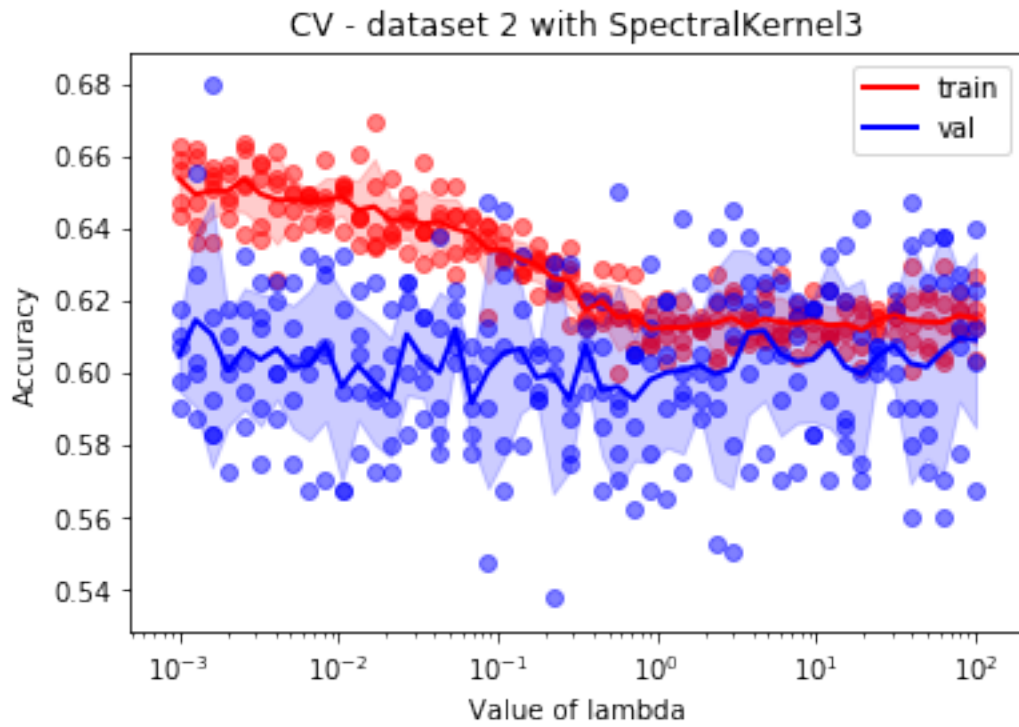
Tuning lambda on dataset 2 with kernel LinearKernel and params {'suffix': 'mat100'}: 100%|| 50%



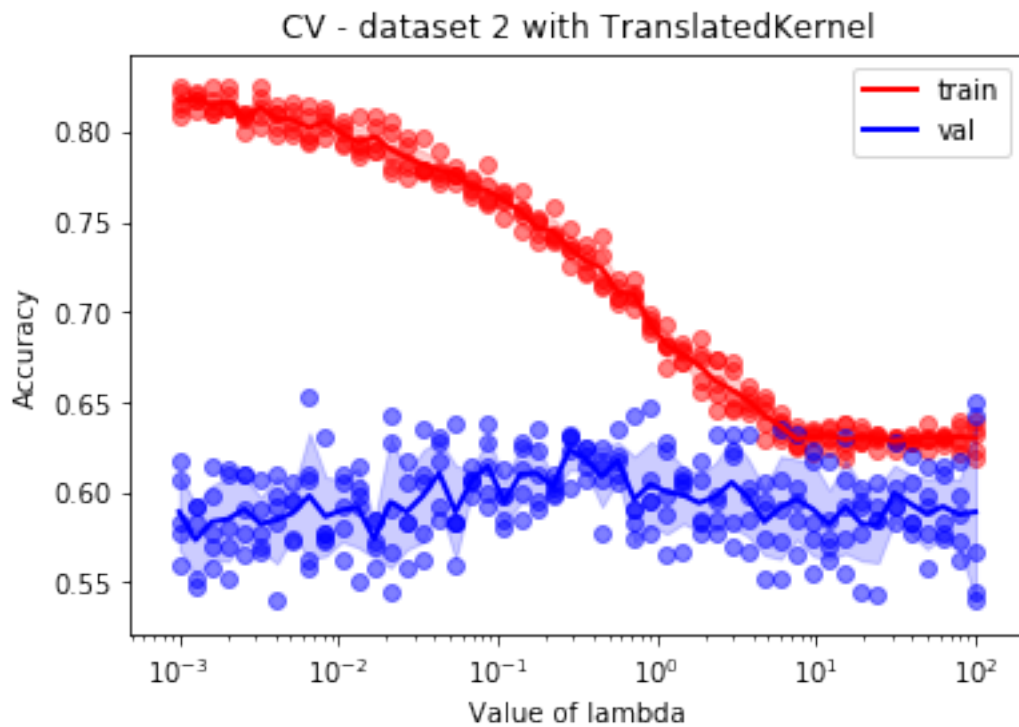
Tuning lambda on dataset 2 with kernel SpectralKernel6 and params {'suffix': 'spectr6'}: 100%



Tuning lambda on dataset 2 with kernel SpectralKernel3 and params {'suffix': 'spectr3'}: 100%



Tuning lambda on dataset 2 with kernel TranslatedKernel and params {'suffix': 'trans6'}: 100%



```
In [9]: backend.final_prediction(three_kernels, three_lambdas)
```

DATASET 0

DATASET 1

DATASET 2