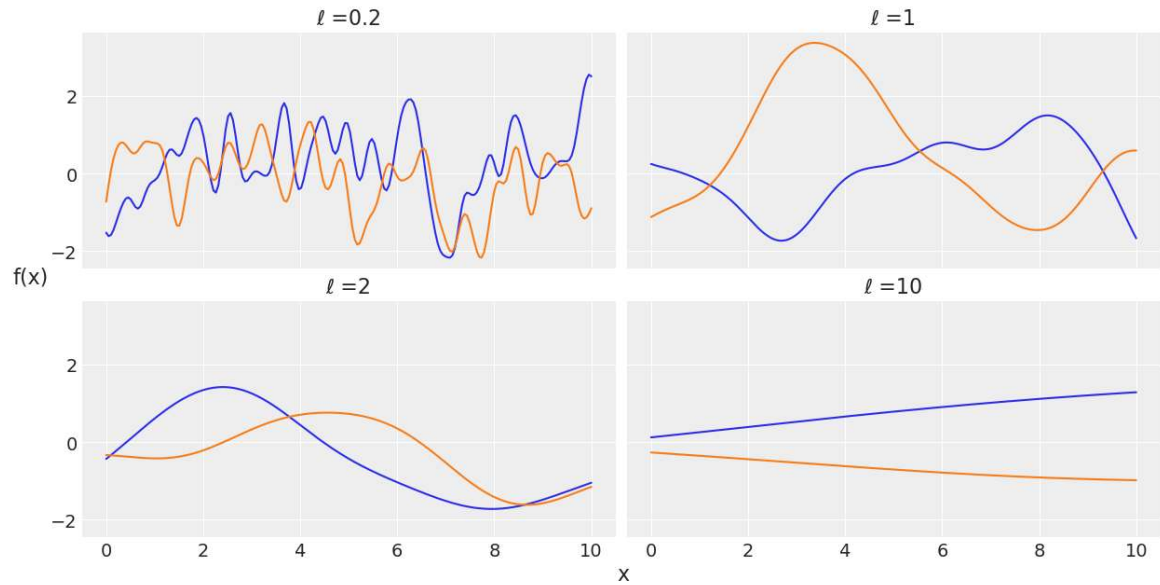


```

In [4]: ▶ np.random.seed(24)
test_points = np.linspace(0, 10, 200)
fig, ax = plt.subplots(2, 2, figsize=(12, 6), sharex=True,
                       sharey=True, constrained_layout=True)
ax = np.ravel(ax)

for idx,  $\ell$  in enumerate((0.2, 1, 2, 10)):
    cov = exp_quad_kernel(test_points, test_points,  $\ell$ )
    ax[idx].plot(test_points, stats.multivariate_normal.rvs(cov=cov, size=2).T)
    ax[idx].set_title(f' $\ell = \{\ell\}$ ')
fig.text(0.51, -0.03, 'x', fontsize=16)
fig.text(-0.03, 0.5, 'f(x)', fontsize=16)
plt.savefig('B11197_07_03.png')

```



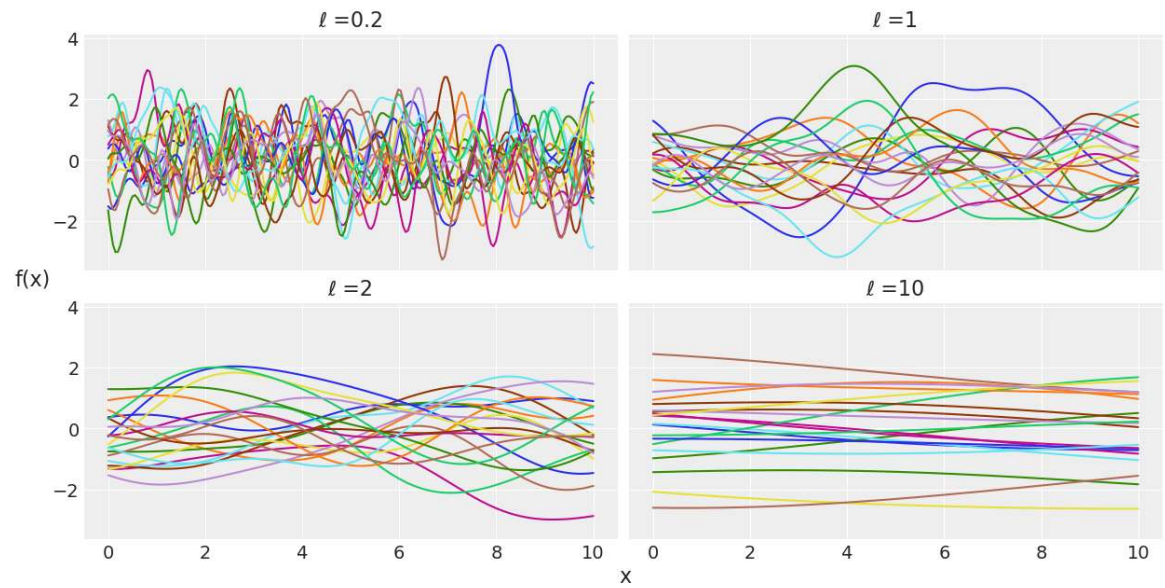
```

In [11]: ▶ np.random.seed(24)
test_points = np.linspace(0, 10, 200)
fig, ax = plt.subplots(2, 2, figsize=(12, 6), sharex=True,
                      sharey=True, constrained_layout=True)
ax = np.ravel(ax)

for idx,  $\ell$  in enumerate((0.2, 1, 2, 10)):
    cov = exp_quad_kernel(test_points, test_points,  $\ell$ )
    ax[idx].plot(test_points, stats.multivariate_normal.rvs(cov=cov, size=20).T)
    ax[idx].set_title(f' $\ell = \{\ell\}$ ')
fig.text(0.51, -0.03, 'x', fontsize=16)
fig.text(-0.03, 0.5, 'f(x)', fontsize=16)

```

Out[11]: Text(-0.03, 0.5, 'f(x)')



```

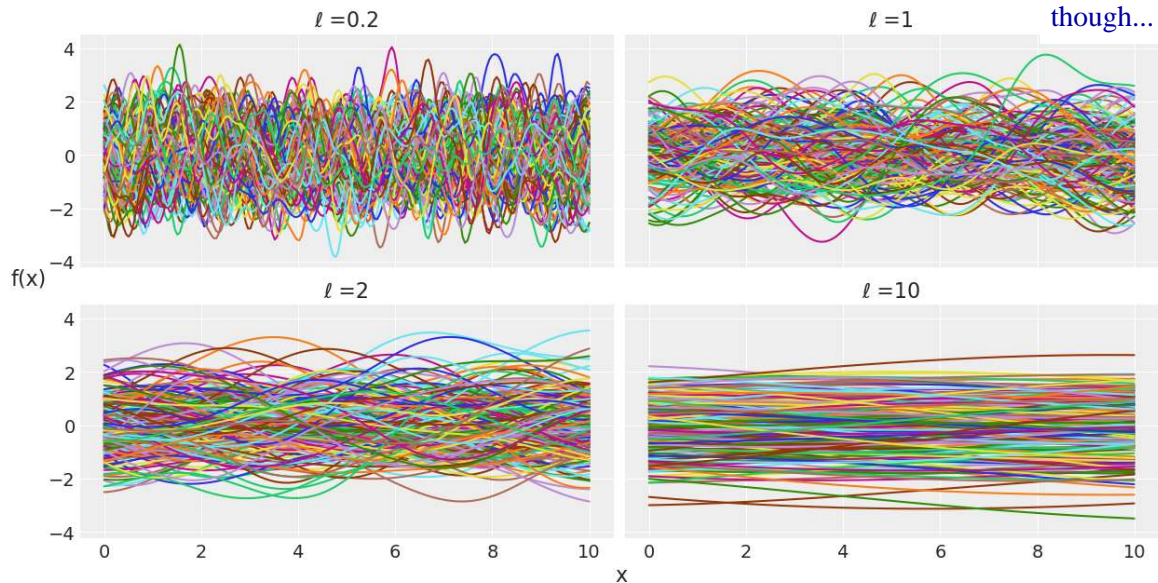
In [5]: ▶ np.random.seed(24)
test_points = np.linspace(0, 10, 200)
fig, ax = plt.subplots(2, 2, figsize=(12, 6), sharex=True,
                      sharey=True, constrained_layout=True)
ax = np.ravel(ax)

for idx, ℓ in enumerate((0.2, 1, 2, 10)):
    cov = exp_quad_kernel(test_points, test_points, ℓ)
    ax[idx].plot(test_points, stats.multivariate_normal.rvs(cov=cov, size=200).T)
    ax[idx].set_title(f'ℓ = {ℓ}')
fig.text(0.51, -0.03, 'x', fontsize=16)
fig.text(-0.03, 0.5, 'f(x)', fontsize=16)

```

I set size=200 so that I could get 200 multi-normal distributions. The plots are not very clear nor informative, though...

Out[5]: Text(-0.03, 0.5, 'f(x)')



```

In [7]: ▶ y_df = pd.DataFrame(stats.multivariate_normal.rvs(cov=cov, size=200))

```

```

In [8]: ▶ y_df.head()

```

Out[8]:

"y" values at each "x"

	0	1	2	3	4	5	6	7	8	
0	0.422100	0.409618	0.397062	0.384436	0.371738	0.358970	0.346135	0.333231	0.320261	0.307
1	-0.778149	-0.779094	-0.780079	-0.781104	-0.782169	-0.783274	-0.784421	-0.785607	-0.786835	-0.788
2	-1.156881	-1.157028	-1.157169	-1.157305	-1.157436	-1.157563	-1.157684	-1.157802	-1.157915	-1.158
3	0.001888	-0.005178	-0.012204	-0.019190	-0.026132	-0.033033	-0.039892	-0.046708	-0.053480	-0.06C
4	0.390947	0.388387	0.385801	0.383188	0.380547	0.377880	0.375186	0.372465	0.369717	0.366

5 rows × 200 columns

```

In [14]: ▶ len(y_df.columns)

```

Out[14]: 200

```
In [17]: y_df_std = []
for i in range(len(y_df.columns)):
    y_df_std.append(round(y_df[i].std(), 2))
```

Since each column represents "y" values at each "x". I computed the standard deviation of each vector in y\_df.

```
In [28]: np.linspace(0, 199, 200)
```

```
Out[28]: array([ 0.,  1.,  2.,  3.,  4.,  5.,  6.,  7.,  8.,  9., 10.,
 11., 12., 13., 14., 15., 16., 17., 18., 19., 20., 21.,
 22., 23., 24., 25., 26., 27., 28., 29., 30., 31., 32.,
 33., 34., 35., 36., 37., 38., 39., 40., 41., 42., 43.,
 44., 45., 46., 47., 48., 49., 50., 51., 52., 53., 54.,
 55., 56., 57., 58., 59., 60., 61., 62., 63., 64., 65.,
 66., 67., 68., 69., 70., 71., 72., 73., 74., 75., 76.,
 77., 78., 79., 80., 81., 82., 83., 84., 85., 86., 87.,
 88., 89., 90., 91., 92., 93., 94., 95., 96., 97., 98.,
 99., 100., 101., 102., 103., 104., 105., 106., 107., 108., 109.,
 110., 111., 112., 113., 114., 115., 116., 117., 118., 119., 120.,
 121., 122., 123., 124., 125., 126., 127., 128., 129., 130., 131.,
 132., 133., 134., 135., 136., 137., 138., 139., 140., 141., 142.,
 143., 144., 145., 146., 147., 148., 149., 150., 151., 152., 153.,
 154., 155., 156., 157., 158., 159., 160., 161., 162., 163., 164.,
 165., 166., 167., 168., 169., 170., 171., 172., 173., 174., 175.,
 176., 177., 178., 179., 180., 181., 182., 183., 184., 185., 186.,
 187., 188., 189., 190., 191., 192., 193., 194., 195., 196., 197.,
 198., 199.]
```

```
In [29]: x=np.linspace(0, 199, 200)
plt.scatter(x, y_df_std)
```

The plot of the std of each column. Stds are around "1". I think this result agree with the the diagonal line of the covariance matrix. But I am not sure if this is what was asked to compare in Question3, and I don't know how to observe the std directly from the "x-y" plots.

```
Out[29]: <matplotlib.collections.Path
```

