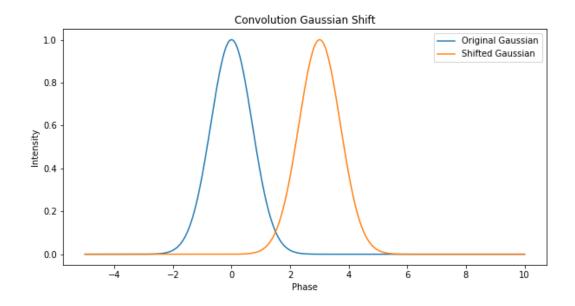
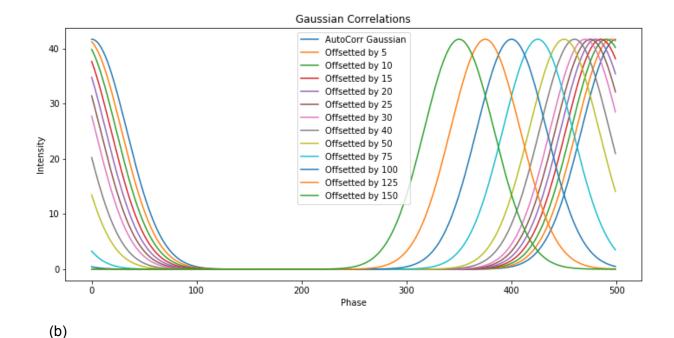
Problem 1)

```
15 def dftOffset(fun:np.ndarray, x0:int):
16
      assert len(fun)%2==0, f"array length {fun.shape} must be even 1-d"
17
      N = len(fun)
18
19
20
      funft = rfft(fun)
      CTFactor = np.exp(-2j * np.pi * x0/N * np.arange(0, N//2 + 1)) #Got reccomenda
21
22
      offsetted = irfft(funft * CTFactor)
23
24
      return offsetted
```



Problem 2)

(a) Since the convolution of two gaussians is a gaussian and they're even functions, we the correlation matches the convolution. Subsequently, the correlation of a gaussian with an offsetted version of itself is the convolution of a gaussian then offsetted by negative that amount. A gaussian's correlation to itself is a gaussian.



26 def correlation(u,v): 27 return irfft(rfft(u) * np.conj(rfft(v))) 28 29 #This function will throw back the correlation of (u,v) 30 31 def offsettedAutoCorrelation(u, offset): 32 #Correlating some fun with its shifted self 33 u_offset = np.roll(u,offset) 34 35 return correlation(u, u_offset) 36

Problem 3)

```
37
38 def wraplessCorrelation(u,v):
39
          #Question of Padding: pad u and v with zeroes to maintain the center of convolution
40
41
      zeroes = np.zeros(len(v))
42
43
          #To avoid circulant conditions, padding input with zeroes should be sufficent.
44
      u_Padding = np.hstack((u, zeroes))
45
      v_Padding = np.hstack((v, zeroes))
46
          #Does the correlation of (u,v) but without the circulant, as it can be problematic...
47
      return correlation(u_Padding, v_Padding)
48
49
```

$$\sum_{x=0}^{N-1} \exp\left[-2\pi i \ln x/N\right] = \frac{1-\exp\left[-2\pi i \ln N\right]}{1-\exp\left[-2\pi i \ln N\right]}$$

$$\begin{array}{lll}
 & \left(1 - \exp\left(-2\pi i \mathbf{k}/\mathbf{N}\right)\right) \sum_{x=0}^{N-1} \exp\left[-2\pi i \mathbf{k}x/\mathbf{N}\right] = 1 - \exp\left[-2\pi i \mathbf{k}\right] = \\
 & = \sum_{x=0}^{N-1} \exp\left[-2\pi i \mathbf{k}x/\mathbf{N}\right] - \sum_{x=0}^{N-1} \exp\left[-2\pi i \mathbf{k}(x,i)/\mathbf{N}\right] \\
 & = \sum_{x=0}^{N-1}$$

=>
$$(1 - e^{2\pi i N/N}) \cdot \sum_{k=0}^{N-1} e^{2\pi i N/N} = 1 - e^{2\pi i N/N}$$

As
$$N \rightarrow 0$$
 $\lim_{N \rightarrow 0} \frac{1 - \exp\left(-2\pi i h\right)}{1 - \exp\left(-2\pi i h\right)} = \lim_{N \rightarrow 0} \frac{\frac{d}{du}\left[1 - \exp\left(-2\pi i h\right)\right]}{1 - \exp\left(-2\pi i h\right)}$

=
$$l_{\text{M}} \left(\frac{2\pi}{2\pi i / N} \right) \left(\frac{exp \left(-2\pi i k \right)}{exp \left(-2\pi i k / N \right)} \right) = N$$

(c)
$$Sm(x) = \frac{e^{ix} - e^{ix}}{2i}$$
 met's scale then arbitrarily:

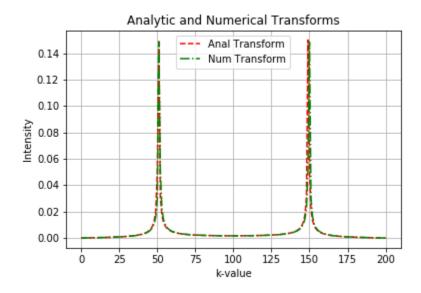
2:
$$sm(\frac{2\pi k'x}{N}) = e^{2\pi i k'x/N} - e^{2\pi i k'x/N}$$

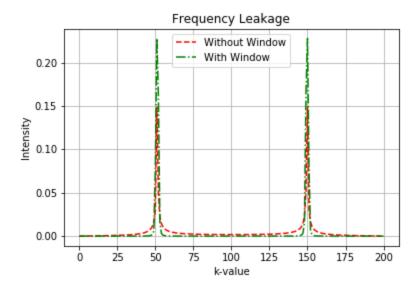
$$\int DFT \Rightarrow F(h) = \sum_{x=0}^{N-1} exp\left[\frac{2\pi i h x}{N}\right] \cdot \left(exp\left[\frac{2\pi i h' x}{N}\right] - exp\left[-\frac{2\pi i h' x}{N}\right]\right)$$

$$= \sum_{x=0}^{N-1} exp\left[\frac{2\pi i \left(h' - h'\right)x}{N}\right] + \sum_{x=0}^{N-1} exp\left[-\frac{2\pi i \left(h' + h'\right)x}{N}\right]$$

$$= \frac{1 - exp[2\pi; (h'-h)]}{1 - exp[\frac{2\pi; (h'-h)}{n}]} + \frac{1 - exp[-2\pi; (h'+h)]}{1 - exp[-\frac{2\pi; (h'+h)}{n}]}$$

Now, we may compare the analytic and numerical estimations of our transforms.





Clearly Less leaking with window (sharper peaks falling to zero more quickly in k-space).

Q4 AND 5 WORK IN PROGRESS (informed TA I would be slightly late)