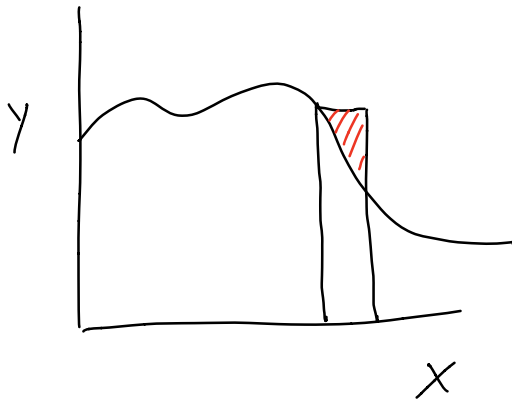
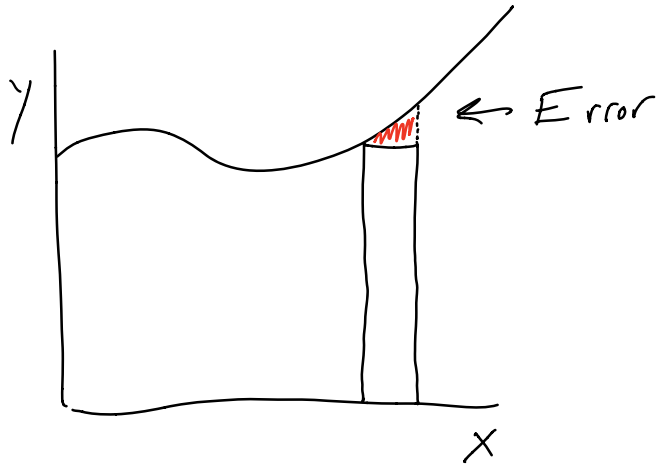
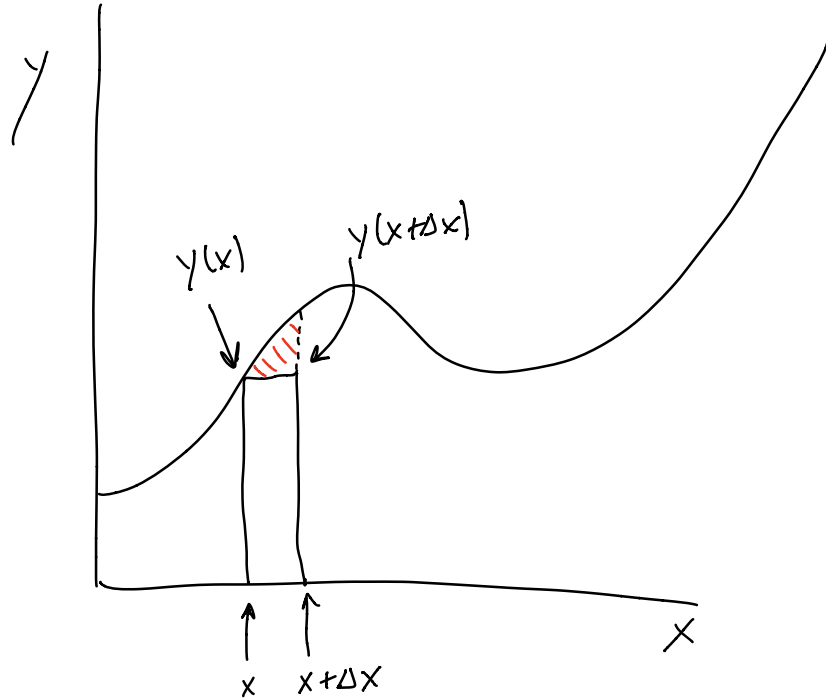


Uncertainty

Where does our error come from?





$$A_{\text{true}} = \int_x^{x+\Delta x} y(x) dx$$

$$A_{\text{approx}} = y(x) \Delta x$$

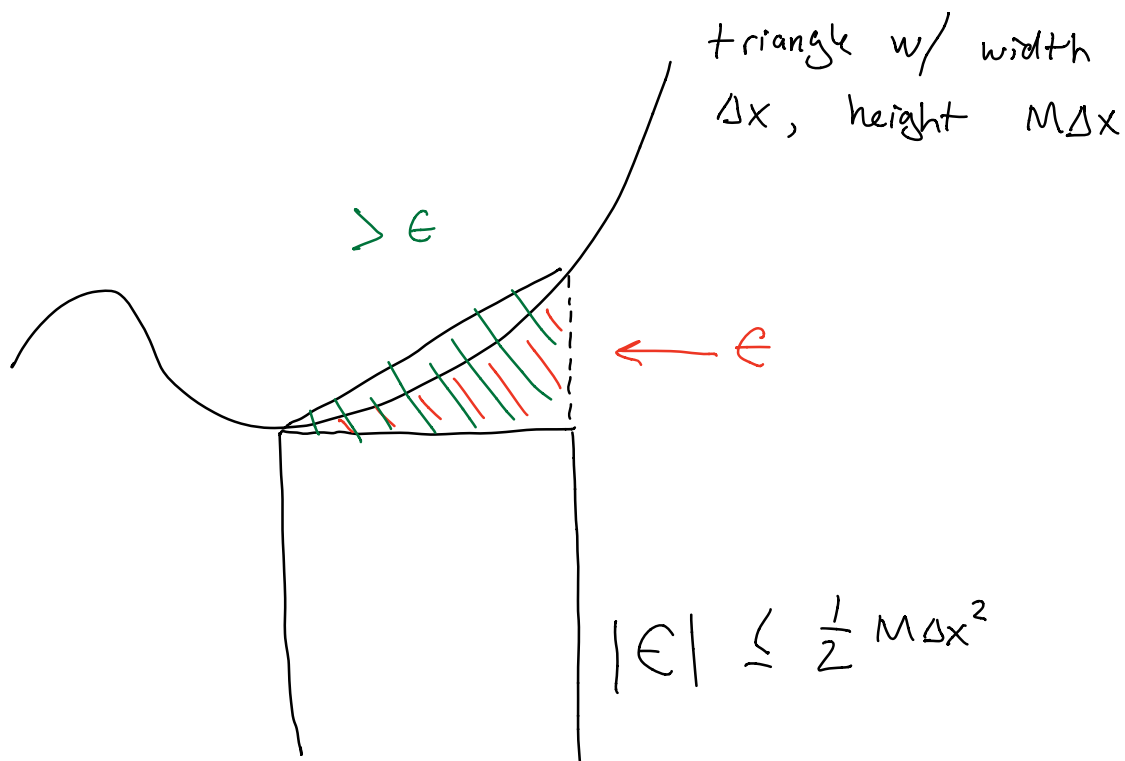
$$E = A_{\text{true}} - A_{\text{approx}}$$

$$\epsilon = A_{\text{true}} - A_{\text{approx}}$$

$$= \int_x^{x+\Delta x} y(x) dx - y(x) \Delta x$$

We can always find a line w/ slope M
such that

$$\left| \int_x^{x+\Delta x} Mx dx - y(x) \Delta x \right| \geq \epsilon$$



$$|\epsilon| \leq \frac{1}{2} M \Delta x^2$$

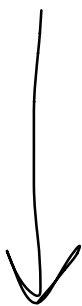
Each rect. we deviate from the true result by a maximum of $\frac{1}{2} M \Delta x^2$,

where M is some number so that

$$M \Delta x \geq \left| y(x + \Delta x) - y(x) \right|$$

$$M \geq \left| \frac{y(x + \Delta x) - y(x)}{\Delta x} \right|$$

$$M \geq \left| y'(x) \right|$$



We deviate this amount each rectangle,

there are n many rectangles

Total error

$$|E| \leq \frac{1}{2} M \Delta x^2 \cdot n$$

$$\Delta x = \frac{x_f - x_i}{n}$$

$$|E| \leq \frac{1}{2} M \frac{(x_f - x_i)^2}{n^2} \cdot n$$

$$\boxed{|E| \leq \frac{1}{2} M \frac{(x_f - x_i)^2}{n}} \quad , \quad M = \max_{x_i \rightarrow x_f} |y'(x)|$$

$|E|$ is the maximum possible
error

We could be closer to the true value, but we are guaranteed to be within $\pm E$

If our estimate is A , then the true answer must be:

$$A - |E| \leq A_{\text{true}} \leq A + |E|$$

```
def force(t):  
    return 5000*(1-np.exp(-t**3/5-t**2))
```

```
mcar = 1100
```

```
ti = 0
```

```
tf = 8
```

```
n = 10
```

```
dt = (tf - ti) / n
```

```
total = 0
```

```
for i in range(n):
```

```
    t = ti + i * dt
```

```
    f = force(t)
```

```
    area = f * dt
```

```
    total += area
```

```
pfinal = total
```

```
vfinal = pfinal / mcar
```

```
print(vfinal)
```

```

def force(t):
    return 5000*(1-np.exp(-t**3/5-t**2))

def calc_uncertainty(xi,xf,M,n):
    """We need to find M somehow?"""
    return 0.5 * M * (xf-xi)**2 / n

mcar = 1100
ti = 0
tf = 8
n = 10

dt = (tf - ti) / n
total = 0
for i in range(n):
    t = ti + i * dt
    f = force(t)
    area = f * dt
    total += area

pfinal = total
vfinal = pfinal / mcar
print(vfinal)

```


$E_x:$

$$F(t) = 5000N \cdot (1 - e^{-0.2t^3 - t^2})$$

$$M = \max(F'(t), 0 \leq t \leq 8s)$$

$$F'(t) = 5000N \left[0 - (-3 \cdot 0.2t^2 - 2t) e^{-0.2t^3 - t^2} \right]$$

$$F'(t) = 5000N (0.6t^2 + 2t) e^{-0.2t^3 - t^2}$$

Plot

Technically, we find max of $F'(t)$

by finding the critical points ($F''(t) = 0$)

- I don't feel like doing that!

```

"""Plot F'(t)"""
def force_deriv(t):
    return 5000*(0.6*t**2+2*t)*np.exp(-t**3/5-t**2)
t = np.linspace(0,8,1000)
dfdt = force_deriv(t)
plt.plot(t,dfdt)

```

```

"""Plot F'(t)"""
def force_deriv(t):
    return 5000*(0.6*t**2+2*t)*np.exp(-t**3/5-t**2)
t = np.linspace(0,8,100000)
dfdt = force_deriv(t)
plt.plot(t,dfdt)
M = dfdt.max()
plt.axhline(ftmax)
print(ftmax)

```

```

def force(t):
    return 5000*(1-np.exp(-t**3/5-t**2))

def force_deriv(t):
    return 5000*(0.6*t**2+2*t)*np.exp(-t**3/5-t**2)

def calc_uncertainty(xi,xf,M,n):
    """We need to find M somehow?"""
    return 0.5 * M * (xf-xi)**2 / n

mcar = 1100
ti = 0
tf = 8
n = 100
t = np.linspace(ti,tf,10000)
dfdt = force_deriv(t)
M = dfdt.max()

dt = (tf - ti) / n
total = 0
for i in range(n):
    t = ti + i * dt
    f = force(t)
    area = f * dt
    total += area

pfinal = total
pfinal_uncert = calc_uncertainty(ti,tf,M,n)
vfinal = pfinal / mcar
vfinal_uncert = pfinal_uncert / mcar
print('The speed after {} seconds is {}+-{}'.format(tf-ti,vfinal,vfinal_uncert))

```

```

def func(x):
    pass

def deriv(x):
    pass

def calc_uncertainty(xi,xf,n):
    """We need to find M somehow?"""
    x = np.linspace(xi,xf,10000)
    dydx = deriv(x)
    M = dydx.max()
    return 0.5 * M * (xf-xi)**2 / n

def integrate(xi,xf,n):
    #First get estimate
    dx = (xf - xi) / n
    total = 0
    for i in range(n):
        x = xi + i * dx
        f = func(x)
        area = f * dx
        total += area

    #Now get uncertainty
    uncert = calc_uncertainty(xi,xf,n)
    return total,uncert

```

```
def func(x):  
    return x**2
```

```
def deriv(x):  
    return 2*x
```

```
def calc_uncertainty(xi,xf,n):  
    """We need to find M somehow?"""  
    x = np.linspace(xi,xf,10000)  
    dydx = deriv(x)  
    M = dydx.max()  
    return 0.5 * M * (xf-xi)**2 / n
```

```
def integrate(xi,xf,n):  
    #First get estimate  
    dx = (xf - xi) / n  
    total = 0  
    for i in range(n):  
        x = xi + i * dx  
        f = func(x)  
        area = f * dx  
        total += area  
  
    #Now get uncertainty  
    uncert = calc_uncertainty(xi,xf,n)  
    return total,uncert
```

```
nvals = np.logspace(1,4,20,dtype=int) #10**np.linspace(1,4,50)  
avals = []  
dvals = []  
for n in nvals:  
    a,da = integrate(0,10,n)  
    avals.append(a)  
    dvals.append(da)
```

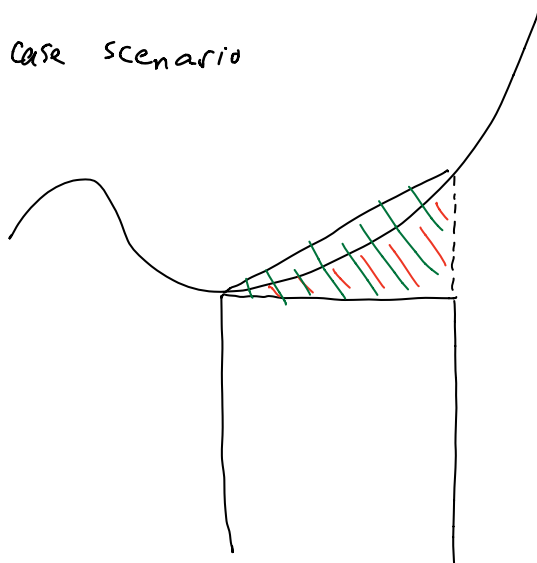
```
atru = 1/3 * 10**3
plt.errorbar(nvals,avals,yerr=davals,fmt='o',color='blue',capsize=2)
plt.xscale('log')
plt.axhline(atru,c='k',ls='--')
plt.xlabel('n')
plt.ylabel('A_est')
```

Improving the approximation

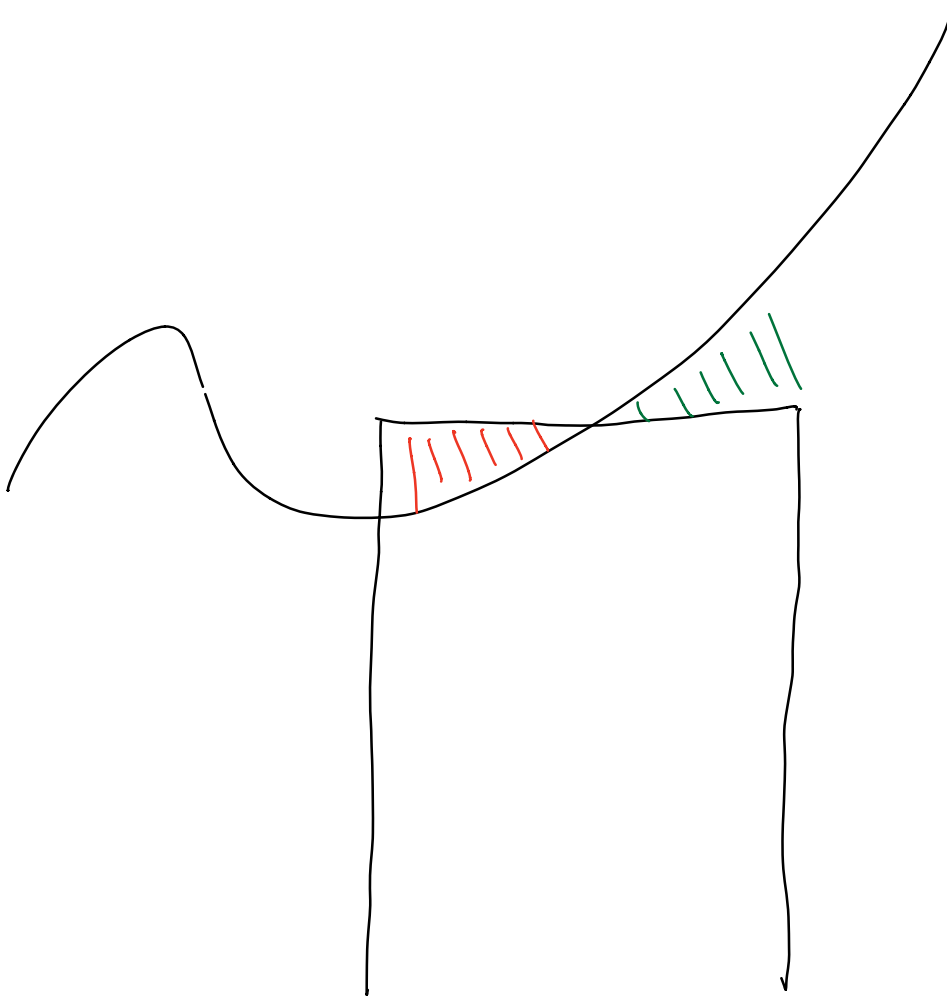


We get errors because the function deviates from our rectangle at the edge

- Worst case scenario



What if I shift each rectangle so that instead of the left edge touching the curve, the center does



$$A = \gamma\left(x_i + \frac{\Delta x}{2}\right) \cdot \Delta x + \gamma\left(x_i + \frac{3\Delta x}{2}\right) \cdot \Delta x + \dots$$

$$= \sum_{k=0}^{n-1} \gamma\left(x_i + \left(k + \frac{1}{2}\right) \Delta x\right) \Delta x$$

```
def integrate_midpoint(xi,xf,n):
```

```
    #First get estimate
```

```
    dx = (xf - xi) / n
```

```
    total = 0
```

```
    for i in range(n):
```

```
        x = xi + (i+1/2) * dx
```

```
        f = func(x)
```

```
        area = f * dx
```

```
        total += area
```

```
    return total
```

```
def func(x):  
    return x**2
```

```
def deriv(x):  
    return 2*x
```

```
def calc_uncertainty(xi,xf,n):  
    """We need to find M somehow?"""  
    x = np.linspace(xi,xf,10000)  
    dydx = deriv(x)  
    M = dydx.max()  
    return 0.5 * M * (xf-xi)**2 / n
```

```
def integrate_midpoint(xi,xf,n):  
    #First get estimate  
    dx = (xf - xi) / n  
    total = 0  
    for i in range(n):  
        x = xi + (i+1/2) * dx  
        f = func(x)  
        area = f * dx  
        total += area  
    return total
```

```
def integrate(xi,xf,n):  
    #First get estimate  
    dx = (xf - xi) / n  
    total = 0  
    for i in range(n):  
        x = xi + i * dx  
        f = func(x)  
        area = f * dx  
        total += area
```

```

#Now get uncertainty
uncert = calc_uncertainty(xi,xf,n)
return total,uncert

nvals = np.logspace(1,4,20,dtype=int) #10**np.linspace(1,4,50)
vals_left = []
vals_mid = []
#dvals = []
for n in nvals:
    yleft,dyleft = integrate(0,10,n)
    ymid = integrate_midpoint(0,10,n)
    vals_left.append(yleft)
    vals_mid.append(ymid)

atrue = 1/3 * 10**3
plt.scatter(nvals,vals_left,c='blue')
plt.scatter(nvals,vals_mid,c='green')
plt.xscale('log')
plt.axhline(atrue,c='k',ls='--')
plt.xlabel('n')
plt.ylabel('A_est')

```