



```
2-5
Precision limited by random errors
     - cluseness of grouping
     - may or may not get 'correct' value
     - estimate by sepetition (Std. deu.)
Accorney measures how close to 'corred' value
 includes both random & systematic
Consider simple physics lab to mecsure
Consider Simple physics has
acceleration due to gravity of

Drop ball from height (h)

Time to fall (t)
Messure time to fall (t)
 Repeat N times
Assume we know h=1m exactly
Stopwatch claims according to 5 ms
 L. = 30
           t, = 52
             t3 = 47
             ty = 41
             ts = 52
```

How do we report our answer for g? 2-3 We define mean and standard deviation of our sample Sample mean $\overline{t} = \frac{1}{N} \sum_{i=1}^{N} t_i$ (44.4 ms) Sample variance $s'(ti) = \frac{1}{N-1} \sum_{i=1}^{N} (t_i - \overline{t})^2$ Sample Standard deviction: (5°(ti) = S(ti) s(te)=9.2 ms en scasonable? Theory: $y = 1/3 gt^2 = D g = \frac{\partial y}{t^2}$ How could we estimate our 'error' or uncertainty in 9 based on our messured uclues ti.? Method: Les measure t.F.S... and want a = f (t,1,s) to compute some velve a g = 2h/t2 7 (like g = f(t,h) since

2-4 We estimate $\bar{g} = \frac{\partial h}{(\bar{t})^2}$ $\bar{a} = f(\bar{t}, \bar{r}, \bar{s}, \dots)$ I Not the only way] Remember Taylor Series Write fixt around a point Xo polynomial $f(x) = f(x_0) + (x_0) f'(x_0) + \frac{(x_0)^2 f''(x_0)}{2!}$ $+ \frac{3!}{(x-x^0)^2} \int_{111}^{111} (x^0) + \cdots$ Example: $f(x) = (1-x)^{-1/2}$ expand around X=0 $f(0) = (1-0)^{-1/2} = 1$ f'(0) = 1/2 $f'(x) = -\frac{1}{2}(1-x)^{-\frac{3}{2}}(-1)$ f"(0) = 3/1 $f''(x) = -3/4(1-x)^{-5/5}(-1)$ $f(x) \approx 1 + x(1/2) + x^{2}(3/4) = 1 + x^{2} + \frac{3x^{2}}{8}$ [See plot]

Uncertainty Propogation We measure t with some uncertainty St We want uncertainty in f(t) [like 24/t') f(to) - Sf Sf=f'(to) St f'(t, c, s, ...)For Many Variables: $Sf' \simeq \left(\frac{3f}{3f}\right)^2 Sf' + \left(\frac{3f}{3f}\right)^2 Ss' + \dots$ //// W Ex: Area = I W l± El > Area = lowot Messure ω, ± δω $SA^2 = \left(\frac{\partial A}{\partial e}\right)^2 Sl^2 + \left(\frac{\partial A}{\partial w}\right)^2 f\omega^2$

More useful:
$$(SA)^2 = (SL)^2 + (SW)^2$$

Let's say $lo = 100 \pm (mm \quad A_0 = 1000 mm^2)$
 $loo = 10 \pm 1 mm \quad A_0 = 1000 mm^2$
 $(SA)^2 = (\frac{1}{100})^2 + (110)^2 = (196)^2 + (1076)^2$
 $SA/A = 1096$

Graity lab: What is our uncertainty in L^2 .

Is it sample standard deviation $S(E)^2$.

 $L = \frac{1}{5} \left[L_1 + L_5 + ... + L_5 \right]$
 $(SL)^2 = (\frac{3L}{3})^2 SL^2 + (\frac{3L}{3})^2 SL^2 + ... + (\frac{3L}{3})^2 SL^2$

Assume $SL = SL = ... = SL = SL$

We estimate $SL = SL = ... = SL = SL$
 $(SL)^2 = (115)^2 S^2 + (115)^2 S^2 + ... + (115)^2 S^2$

7-7 (St) = 5 (1/51) s' = 5/5 Generally St = S(E) Uncertainty in mean Nu merical compating 'essois' - approximation (we cut off taylor series after finite terms) computer doesn't have infinite memory to store 52 uses floating point representation python floats are double precision (doubles) X double = (-1) x 1. f x 2 n-1003 n= exponent s = sign 1 bit f'= Mantusa 11 61/5 5) 6,45 Precision range + 2 -1074 (-) + 2 1004 de 0 66 5 ±4.9e-304 (-> ± 1.8e 308