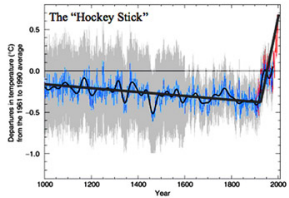



Measuring Things

E80 Spring 2015

Why do we care?

- Who is this man?

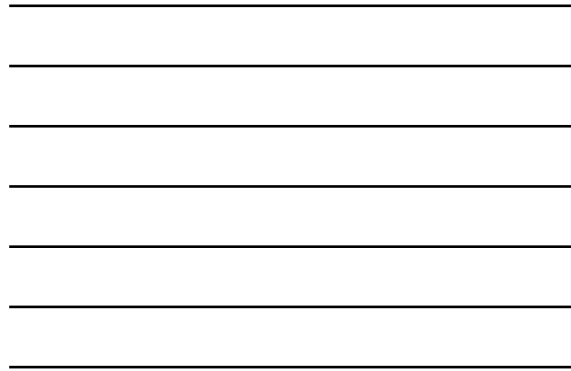
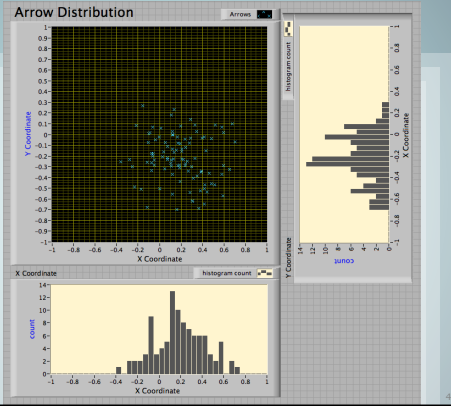


Your Mission

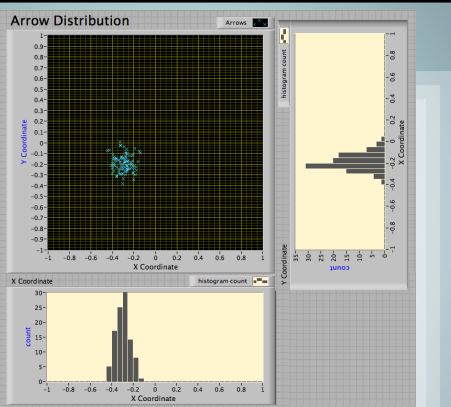
- Mythbuster Tech Advisor
- Testing Arrows for a crossbow



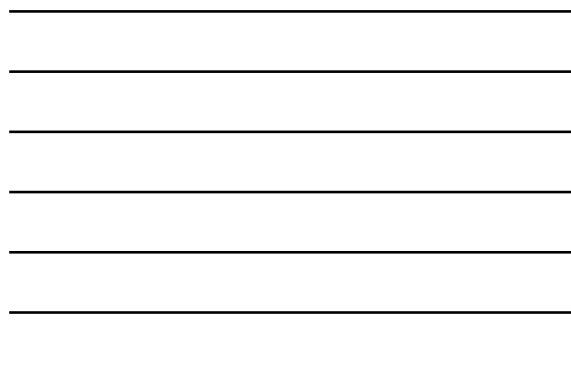
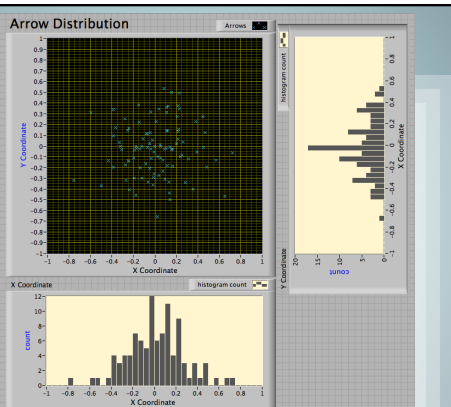
Brand 1 Arrows

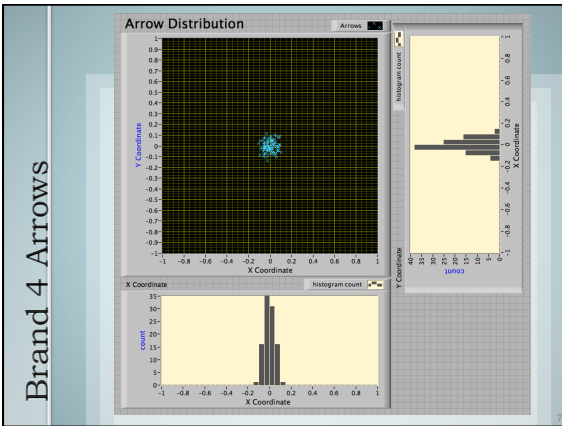


Brand 2 Arrows



Brand 3 Arrows





Crossbow Questions

- Where is the center of the distribution, i.e., what is the aiming point?
- How certain are you?
- If we aim the crossbow at the apple, how likely is Buster™ to get an arrow in the head?

Measured Value

- Where is the center of the distribution, i.e., where should we put the target?
- Use the *sample mean*.

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$$

Error Estimate

- How certain are you? $\pm \lambda$
- Use *Estimated Standard Error* and Student's *t*-test

$$\lambda = tESE = \frac{tS}{\sqrt{N}}$$

$$S = \sqrt{S^2}$$

$$S^2 \equiv \frac{1}{N-1} \sum_{i=1}^N e_i^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$$

Students *t*-test

df	.20	.10	.05	.02	.01	.001
1	3.078	6.314	12.706	31.821	63.657	636.619
2	1.886	2.920	4.303	6.965	9.925	31.598
3	1.638	2.353	3.182	4.541	5.841	12.941
4	1.533	2.132	2.776	3.747	4.604	8.610
5	1.476	2.015	2.571	3.365	4.032	6.859
10	1.372	1.812	2.228	2.764	3.169	4.587
20	1.325	1.725	2.086	2.528	2.845	3.850
30	1.310	1.697	2.042	2.457	2.750	3.646
40	1.303	1.684	2.021	2.423	2.704	3.551
60	1.296	1.671	2.000	2.390	2.660	3.460
120	1.289	1.658	1.980	2.358	2.617	3.373
∞	1.282	1.645	1.960	2.326	2.576	3.291

Spread of Data

- If we aim the crossbow at the apple, how likely is Buster™ to get an arrow in the head?
- Use *sample standard deviation*

$$S = \sqrt{S^2}$$

$$S^2 \equiv \frac{1}{N-1} \sum_{i=1}^N e_i^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$$

Linear Fits

- How about fitting lines to sets of data?

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$$

- Use *linear least squares* fit.

$$\hat{\beta}_1 = \frac{\sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^N (x_i - \bar{x})^2}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

How Certain Are We?

- How good are the betas?

$$S_e = \sqrt{\frac{SSE}{N-2}} = \sqrt{\frac{\sum_{i=1}^N e_i^2}{N-2}} \quad df = N - 2$$

$$S_{\beta_0} = S_e \sqrt{\frac{1}{N} + \frac{\bar{x}^2}{\sum_{i=1}^N (x_i - \bar{x})^2}} \quad \lambda_{\beta_0} = t S_{\beta_0}$$

$$S_{\beta_1} = S_e \sqrt{\frac{1}{\sum_{i=1}^N (x_i - \bar{x})^2}} \quad \lambda_{\beta_1} = t S_{\beta_1}$$

Why? Because!

- How certain is the calculated value of y for a chosen x ?

$$S_y = S_e \sqrt{\frac{1}{N} + \frac{(x - \bar{x})^2}{\sum_{i=1}^N (x_i - \bar{x})^2}} \quad \lambda_y = t S_y$$

- If we experimentally set x , how will the experimental y 's spread (The arrow in the head question)?
- Use S_e like you previously used S .

What About Functions?

- What if you want T ?

$$T = \frac{1}{\frac{1}{T_0} + \frac{1}{\beta} \ln \frac{R}{R_0}}$$

- and you measure β and R ?

How error in x affects $F(x)$

Error Propagation

- For $F = F(x, y, z, \dots)$

$$F - F_{true} = \frac{\partial F}{\partial x}(x - x_{true}) + \frac{\partial F}{\partial y}(y - y_{true}) + \frac{\partial F}{\partial z}(z - z_{true}) + \dots$$

- Let $\epsilon_x = x - x_{true}$

$$\epsilon_F = \frac{\partial F}{\partial x} \epsilon_x + \frac{\partial F}{\partial y} \epsilon_y + \frac{\partial F}{\partial z} \epsilon_z + \dots$$

$$\epsilon_F = \sqrt{\left(\frac{\partial F}{\partial x}\right)^2 \epsilon_x^2 + \left(\frac{\partial F}{\partial y}\right)^2 \epsilon_y^2 + \left(\frac{\partial F}{\partial z}\right)^2 \epsilon_z^2 + \dots}$$

Thermistor Example

$$T = \frac{1}{\frac{1}{T_0} + \frac{1}{\beta} \ln \frac{R}{R_0}}$$

$$dT = \frac{\left[\left(\frac{1}{\beta^2} \ln \frac{R}{R_0} \right) d\beta - \frac{1}{\beta R} dR \right]}{\left[\frac{1}{T_0} + \frac{1}{\beta} \ln \frac{R}{R_0} \right]^2}$$

$$dT = T^2 \left[\left(\frac{1}{\beta^2} \ln \frac{R}{R_0} \right) d\beta - \frac{1}{\beta R} dR \right]$$

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Results

$$e_T = T^2 \left[\left(\frac{1}{\beta^2} \ln \frac{R}{R_0} \right)^2 e_\beta^2 + \left(\frac{1}{\beta R} \right)^2 e_R^2 \right]^{1/2}$$

	Nom. Value	error%	error	error term
β	4261	1%	42.61	0.23
R (Ω)	3,700,000	10%	370000	1.75
T (K)	273.14			1.77
T_0 (K)	298.15			
R_0 (Ω)	1000000			

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What if you hate calculus?

$$T = \frac{1}{\frac{1}{T_0} + \frac{1}{\beta} \ln \frac{R}{R_0}}$$

You can do the whole thing on a spreadsheet with no calculus.

Value	Nominal	$\beta + 1\%$	$\beta - 1\%$	$R + 10\%$	$R - 10\%$
T_0 (K)	298.15	298.15	298.15	298.15	298.15
R_0 (Ω)	1,000,000	1,000,000	1,000,000	1,000,000	1,000,000
β	4261	4303.61	4218.39	4261	4261
R (Ω)	3,700,000	3,700,000	3,700,000	4,070,000	3,330,000
T (K)	273.14	273.37	272.91	271.49	275.00
ΔT (K)		0.46			-3.52
error (K)	3.55				
\pm error (K)	1.77				

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Steinhart-Hart Example

$$T = \frac{1}{A_1 + B_1 \ln \frac{R}{R_{ref}} + C_1 \left(\ln \frac{R}{R_{ref}} \right)^2 + D_1 \left(\ln \frac{R}{R_{ref}} \right)^3}$$

$$dT = \frac{\left[dA_1 + (\ln R) dB_1 + (\ln R)^2 dC_1 + (\ln R)^3 dD_1 + \frac{B_1 + 2C_1 \left(\ln \frac{R}{R_{ref}} \right) + 3D_1 \left(\ln \frac{R}{R_{ref}} \right)^2}{R} dR \right]}{\left[A_1 + B_1 \ln \frac{R}{R_{ref}} + C_1 \left(\ln \frac{R}{R_{ref}} \right)^2 + D_1 \left(\ln \frac{R}{R_{ref}} \right)^3 \right]^2}$$

$$dT = -T^2 \left[dA_1 + (\ln R) dB_1 + (\ln R)^2 dC_1 + (\ln R)^3 dD_1 + \frac{B_1 + 2C_1 \left(\ln \frac{R}{R_{ref}} \right) + 3D_1 \left(\ln \frac{R}{R_{ref}} \right)^2}{R} dR \right]$$

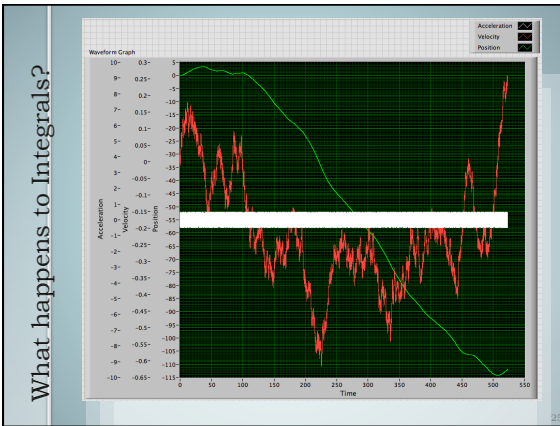
$$e_r = T^2 \left[\left[e_{A_1} \right]^2 + \left[\left(\ln \frac{R}{R_{ref}} \right) e_{B_1} \right]^2 + \left[\left(\ln \frac{R}{R_{ref}} \right)^2 e_{C_1} \right]^2 + \left[\left(\ln \frac{R}{R_{ref}} \right)^3 e_{D_1} \right]^2 + \left[\frac{B_1 + 2C_1 \left(\ln \frac{R}{R_{ref}} \right) + 3D_1 \left(\ln \frac{R}{R_{ref}} \right)^2}{R} e_R \right]^2 \right]^{1/2}$$

	Nom. Value	error%	error	error term
A ₁	3.354016E-03	0.1%	3.35402E-06	0.318
B ₁	2.460382E-04	1%	2.46038E-06	-0.104
C ₁	3.405377E-06	1%	3.40538E-08	0.001
D ₁	1.034240E-07	1%	1.03424E-09	0.000
R (Ω)	64080	3%	1922.4	0.692
T (K)	308.15			0.769
R _{ref}	100,000			

Calculus free style

$$T = \frac{1}{A_1 + B_1 \ln \frac{R}{R_{ref}} + C_1 \left(\ln \frac{R}{R_{ref}} \right)^2 + D_1 \left(\ln \frac{R}{R_{ref}} \right)^3}$$

Value	nominal	R _{ref} = 100,000	±0.1%	±0.1%	±1%	±1%	±1%	±1%	±1%	±1%	±1%	±1%	±1%	±1%	±1%
A ₁	3.354016E-03	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016	0.003354016
B ₁	2.460382E-04	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382	0.0002460382
C ₁	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06	3.405377E-06
D ₁	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07	1.034240E-07
R	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080	64080
T	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15	308.15
error	0.769														



Finite Precision?

- Calculate or determine the quantization range, q , e.g., for a $\pm 5V$, 12-bit DAQ:

$$q = 10V \frac{1}{2^{12}} = \frac{10V}{4096} = 0.027V$$
- If $S > \frac{10q}{\sqrt{12}}$ ignore quantization
- If $\frac{q}{\sqrt{12}} < S < \frac{10q}{\sqrt{12}}$ include as $S_{used} = \sqrt{S^2 + \frac{q^2}{12}}$
- If $S < \frac{q}{\sqrt{12}}$ confidence interval is $\pm q/2$

Resources

- Data Analysis Lecture Notes
 - <http://www.eng.hmc.edu/NewE80/PDFs/DataAnalysisLecNotes2015.pdf>
- NIST Engineering Statistics Handbook
 - <http://www.itl.nist.gov/div898/handbook/>
- ISO Guide to the Expression of Uncertainty in Measurement
 - <http://www.iso.org/sites/JCGM/GUM/JCGM100/C045315e-html/C045315e.html?csnumber=50461>
