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## 1

### 1.1

Generally usefull stuff:

$$f(x) = x \cdot \exp(-x) \quad (1)$$

$$f'(x) = (x - 1) \cdot \exp(-x) \quad (2)$$

$$f''(x) = (x - 2) \cdot \exp(-x) \quad (3)$$

$$F(x) = -\exp(-x) \cdot (x + 1) \quad (4)$$

**Mean  $m$**

$$m = \int_0^{\infty} f_m(x) dx, \text{ with } f_m(x) = x \cdot f(x) \quad (5)$$

$$F_m(x) = -\exp(-x) * (x^2 + 2x + 2) \quad (6)$$

$$m = \lim_{a \rightarrow \infty} F_m(a) - F_m(0) \quad (7)$$

$$= \lim_{a \rightarrow \infty} 2 - \exp(-x) * (x^2 + 2x + 2) = 2 \quad (8)$$

**Mode  $e$**

Three conditions have to be fulfilled for any value  $e$  to be a maximum:

$$f'(e) = (e - 1) \cdot \exp(-e) = 0 \quad (9)$$

$$f''(e) = (e - 2) \cdot \exp(-e) < 0 \quad (10)$$

$$e > 0 \quad (11)$$

$$\Rightarrow e \in \{1\} \quad (12)$$

Since there is only one maximum, it is at the same time the mode.

**Median  $c$**

$$0.5 = \int_0^c f(x) dx = F(c) - F(0) \quad (13)$$

$$\Rightarrow 0.5 = (c + 1) \cdot \exp(-c) \quad (14)$$

$$c > 0 \quad (15)$$

$$c \approx 1.67835 \quad (16)$$

The right hand side of 14 is a transcendental function of  $c$ . See numerical code for solution.

RMS  $r$

$$r^2 = \int_0^{\infty} f_r(x) dx, \text{ with } f_r(x) = (x - m)^2 \cdot f(x) \quad (17)$$

$$F_r(x) = -\exp(-x) * (x^2 - x^2 + 2x + 2) \quad (18)$$

$$r^2 = \lim_{a \rightarrow \infty} F_r(a) - F_r(0) = F_r(0) = 2 \quad (19)$$

$$\Rightarrow r = \sqrt{2} \quad (20)$$

## 1.2

Assuming a binomial process with probability of success (hitting the city)  $p = 0.02$  in each of the  $n = 100$  trials. The probability  $P_{100,0.02,0}$  of zero successes  $k = 0$  is:

$$P_{100,0.02,0} = \frac{n!}{k! \cdot (n - k)!} \cdot p^k \cdot (1 - p)^{n-k} \approx 0.13262 \quad (21)$$

The number of trials necessary to have no successes with a probability of less than  $1 - 0.95$  is bounded by the following inequality:

$$1 - 0.95 > P_{n,0.02,0} = \frac{n!}{k! \cdot (n - k)!} \cdot p^k \cdot (1 - p)^{n-k} \quad (22)$$

$$\Rightarrow 0.05 > (1 - p)^n = 0.98^n \quad (23)$$

$$\Rightarrow n > 148.284 \quad (24)$$

At least  $n = 149$  trials are necessary to have at least one hit with a confidence of 0.95.

## 1.3

The drop in number of deaths  $d$  and injuries  $i$  are:

$$1 - \frac{d_{2012}}{d_{2011}} = 1 - \frac{167}{220} \approx 0.241 = 24.1\% \quad (25)$$

$$1 - \frac{i_{2012}}{i_{2011}} = 1 - \frac{3611}{4039} \approx 0.106 = 10.6\% \quad (26)$$

The variance on two incident numbers  $n_1, n_2$  is (assuming a Poisson process) the same as the numbers themselves. The variance on the difference is the sum of the individual differences. It follows that the error on the difference is  $\sigma_n = \sqrt{|n_1 + n_2|}$ . The ratio of differences and errors for  $d$  and  $i$  are:

$$\frac{|d_1 - d_2|}{\sigma_d} \approx 2.7 \quad (27)$$

$$\frac{|i_1 - i_2|}{\sigma_i} \approx 4.9 \quad (28)$$

This means both drops are likely to reflect a change in rates (with more than 0.99 confidence).

## 2

### 2.1

$$\bar{v} = (97 \pm 4)ms^{-1} \quad (29)$$

$$E_{kin} = (1300 \pm 200)J \quad (30)$$

$$E_{kin,corr} = (1280 \pm 120)J \quad (31)$$

The last measurement deviates about  $2\sigma$  from the mean. This is not a surprising occurrence in a series of 7 measurements.

### 2.2

Note that for  $\theta = 1.54 \pm 0.02$  error propagation using the derivative is not suitable for the tan, because the derivative changes to quickly.

$$\text{For } \theta = 0.54 \pm 0.02 : \quad (32)$$

$$\sin(\theta) = 0.51 \pm 0.02 \quad (33)$$

$$\cos(\theta) = 0.857 \pm 0.010 \quad (34)$$

$$\tan(\theta) = 0.60 \pm 0.03 \quad (35)$$

$$\text{For } \theta = 1.54 \pm 0.02 : \quad (36)$$

$$\sin(\theta) = 0.9995 \pm 0.00006 \quad (37)$$

$$\cos(\theta) = 0.03 \pm 0.02 \quad (38)$$

$$\tan(\theta) = 32 \pm_{12}^{61} \quad (39)$$

### 2.3

$$n_2 = 1.50 \pm 0.02 \quad (40)$$

### 2.4

$$\Delta N_{N_0} = \Delta N_0 \exp(-t/\tau) \quad (41)$$

$$\Delta N_\tau = \Delta \tau \cdot N_0 \cdot t \exp(-t/\tau)/\tau^2 \quad (42)$$

$$\Delta N_{N_0} = \Delta N_\tau \quad (43)$$

$$\Rightarrow \Delta N_0 \exp(-t/\tau) = \Delta \tau \cdot N_0 \cdot t \exp(-t/\tau)/\tau^2 \quad (44)$$

$$\Rightarrow t/\tau = \frac{\tau \Delta N_0}{N_0 \Delta \tau} = 1 \quad (45)$$

### 3

$$1 = \int_{-1}^2 a x^2 dx = 3a \quad (46)$$

$$\Rightarrow a = \frac{1}{3} \quad (47)$$

$$\text{mean: } m = \int_{-1}^2 a x^3 dx = \frac{15a}{4} = \frac{5}{4} \quad (48)$$

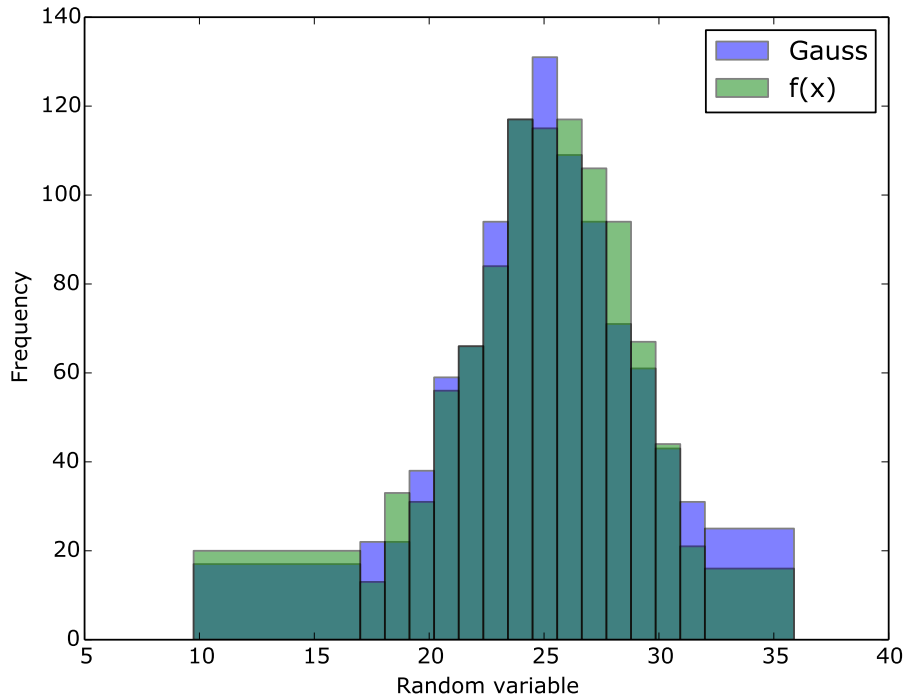
$$\text{square of width: } w^2 = \int_{-1}^2 (x - m)^2 a x^2 dx = m^2 - \frac{5}{2}m + \frac{11}{5} = \frac{51}{80} \quad (49)$$

$$(50)$$

To generate random numbers according to this distribution, one can calculate the cumulative distribution  $c(x) = \frac{1}{9}(x^3 + 1)$  and invert it:

$$x(c) = \begin{cases} (9c - 1)^{\frac{1}{3}} & \text{if } x > \frac{1}{9} \\ -(1 - 9c)^{\frac{1}{3}} & \text{otherwise} \end{cases} \quad (51)$$

A uniformly distributed random variable  $c$  on the interval  $[0, 1]$  can now be transformed into the desired result using  $x(c)$ .



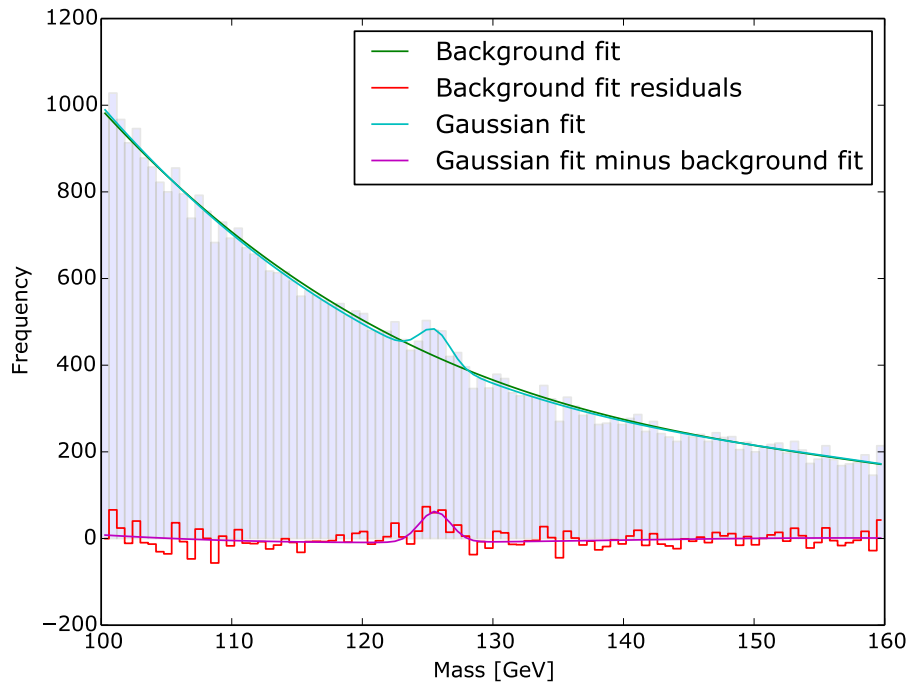
The histogram above shows a sample of a thousand values from the the sum of 20 distributions  $f(x)$  and a gaussian with the same width and mean. The p-value for the hypothesis of independence of a Chi-square contingency test of the two samples is 0.41 and the respective means are  $25.02 \pm 0.12$  for the gaussian and  $25.04 \pm 0.11$  for the convolution of  $f(x)$ . Neither the difference and uncertainties of the means nor the Chi-square test indicate that the hypothesis of equality of the two underlying pdfs can be rejected with a confidence of at least 0.95.

## 4

### 4.1

The Wald-Wolfowitz test on the residuals of the background fit returns 103 and an expected number of runs of  $100 \pm 7$ . This does not indicate, that the fit is insufficient (runs and expectation are consistent).

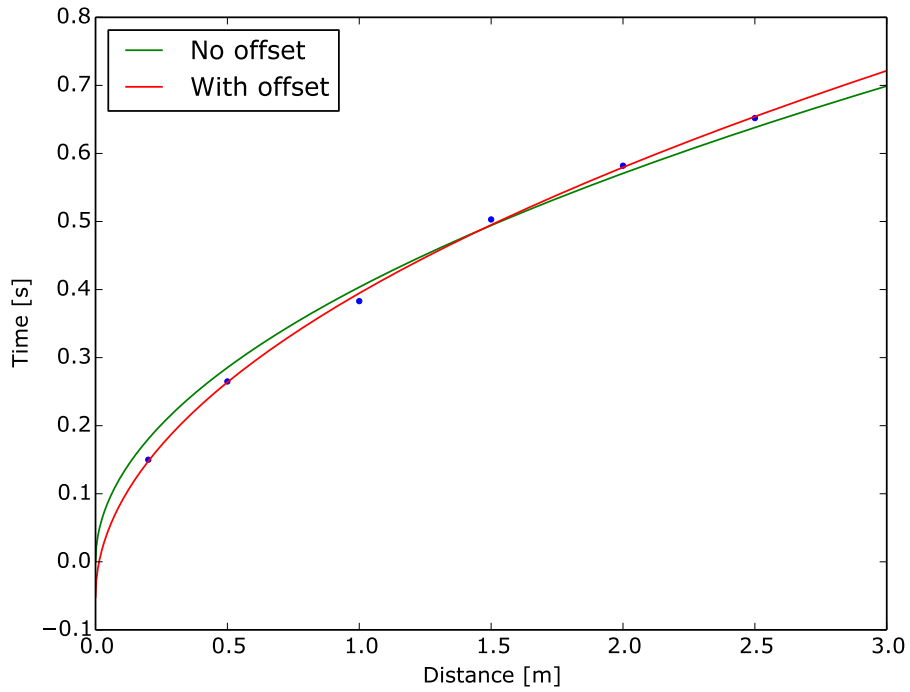
The significance of the largest gaussian (shown in the graph below) is:  $3.3\sigma$



## 4.1

The third result was removed, because its distance to the mean corresponds to  $\approx 20\sigma$ . The true value is  $2.2\mu s$ , the unweighted mean of the results is  $(1.99 \pm 0.06)\mu s$ , the weighted mean is  $(1.91 \pm 0.03)\mu s$ . Both are completely inconsistent with the true value (by  $\approx 10\sigma$ ). Using the weighted mean and uncertainty as best combined measurement, yields  $\chi^2 \approx 21.2$  and  $p \approx 0.006$ .

5



The fit quality of the first hypothesis is very bad ( $\chi^2 \approx 21.6$ ,  $p \approx 0.00061$ ). Adding an offset time as a parameter improves the fit enough to be reasonable ( $\chi^2 \approx 2.16$ ,  $p \approx 0.71$ ). Weighing the two probabilities for obtaining a  $\chi^2$  this bad or worse against each other makes me  $\approx 0.99914$  certain that the hypothesis of an exact release time  $t_0 = 0$  should be rejected.