

# Assessment of model performance

- Likelihood
- AIC
- Absolute measures of model fit
  - Chi-squared test
  - Qq plots
  - Kolmogorov-Smirnov and Cramér-von Mises tests

# Likelihood

$f(x)$  = probability density function of  $x$

$f(x) dx$  = Pr (animal was between  $x$  and  $x+dx$  from the line, given it was detected between 0 and  $w$ ) for small  $dx$

When distances are exact, the **likelihood** is given by

$$L = \prod_{i=1}^n f(x_i) = f(x_1) \times f(x_2) \times \dots \times f(x_n)$$

$x_i$  = distance of  $i^{\text{th}}$  detected animal from the line.

We fit  $f(x)$  by finding the values for the parameters of  $f(x)$  (or equivalently  $g(x)$ ) that maximize  $L$  (or  $\log_e(L)$ ).

# Akaike's Information Criterion

$$AIC = -2\log_e(L) + 2q$$

$L$  is the maximized likelihood (evaluated at the maximum likelihood estimates of the model parameters)

and  $q$  is the number of parameters in the model.

- Models need not be special cases of one another
- Select the model with smallest AIC
- Gives a relative measure of fit

# Limitations of AIC

Cannot be used to select between models when:

- sample size  $n$  differs
- truncation distance  $w$  differs
- data are grouped, and cutpoints differ
- data are grouped in one analysis and ungrouped in the other

# Goodness-of-Fit

- Chi-squared test for grouped (interval) data; if data are exact, we must specify interval cutpoints for this test
- Q-Q plots and related tests for exact data

# Chi-squared tests

Define  $u$  distance intervals, with  $n_i$  detections in interval  $i$ ,  $i = 1, \dots, u$ .

Then

$$\chi^2 = \sum_{i=1}^u \frac{(n_i - n\hat{\pi}_i)^2}{n\hat{\pi}_i}$$

where

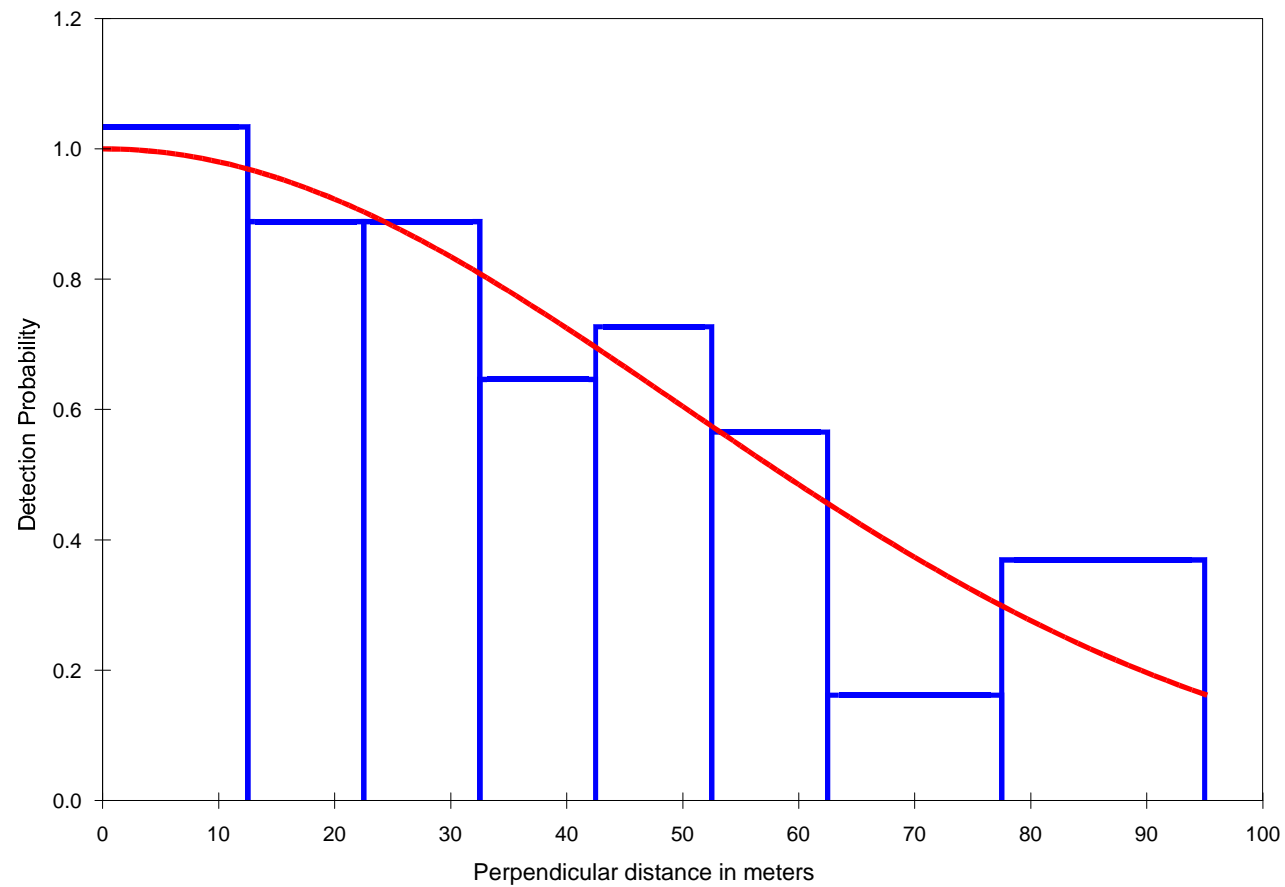
$$n = \sum n_i$$

and  $\hat{\pi}_i$  is the proportion of the area under the estimated pdf,  $\hat{f}(x)$ , that lies in interval  $i$ .

If the model is 'correct':  $\chi^2 \sim \chi_{u-q-1}^2$

$q$  = no. of parameters

# Chaffinch line transect data



# $\chi^2$ goodness-of-fit test

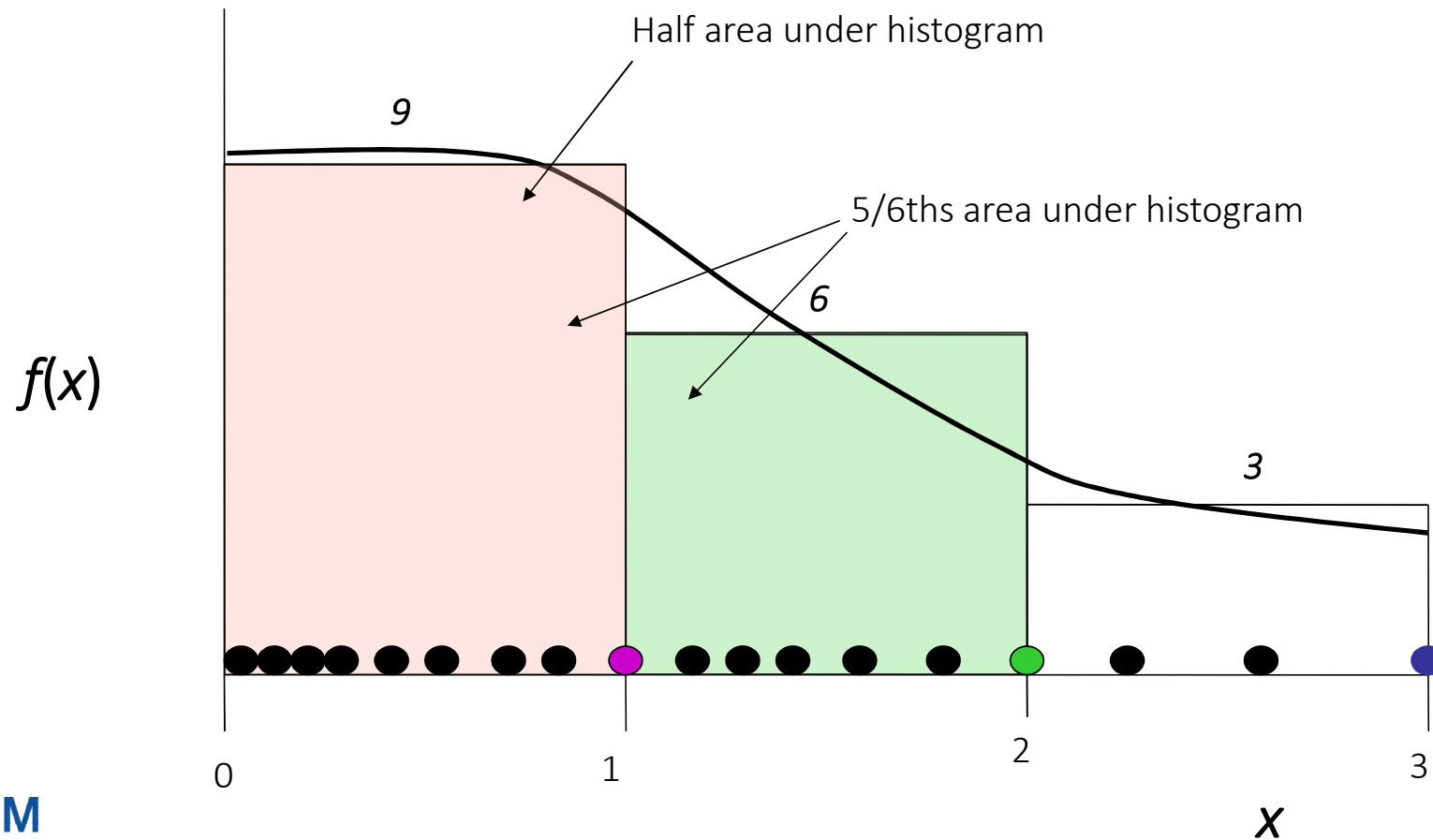
Cell i	Cut Points		Observed Values	Expected Values	Chi-square Values
1	0.000	12.5	16	15.32	0.030
2	12.5	22.5	11	11.63	0.034
3	22.5	32.5	11	10.62	0.013
4	32.5	42.5	8	9.33	0.189
5	42.5	52.5	9	7.87	0.164
6	52.5	62.5	7	6.37	0.062
7	62.5	77.5	3	6.96	2.253
8	77.5	95.0	8	4.91	1.953
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Total Chi-square value =			4.6970	Degrees of Freedom =	6.00

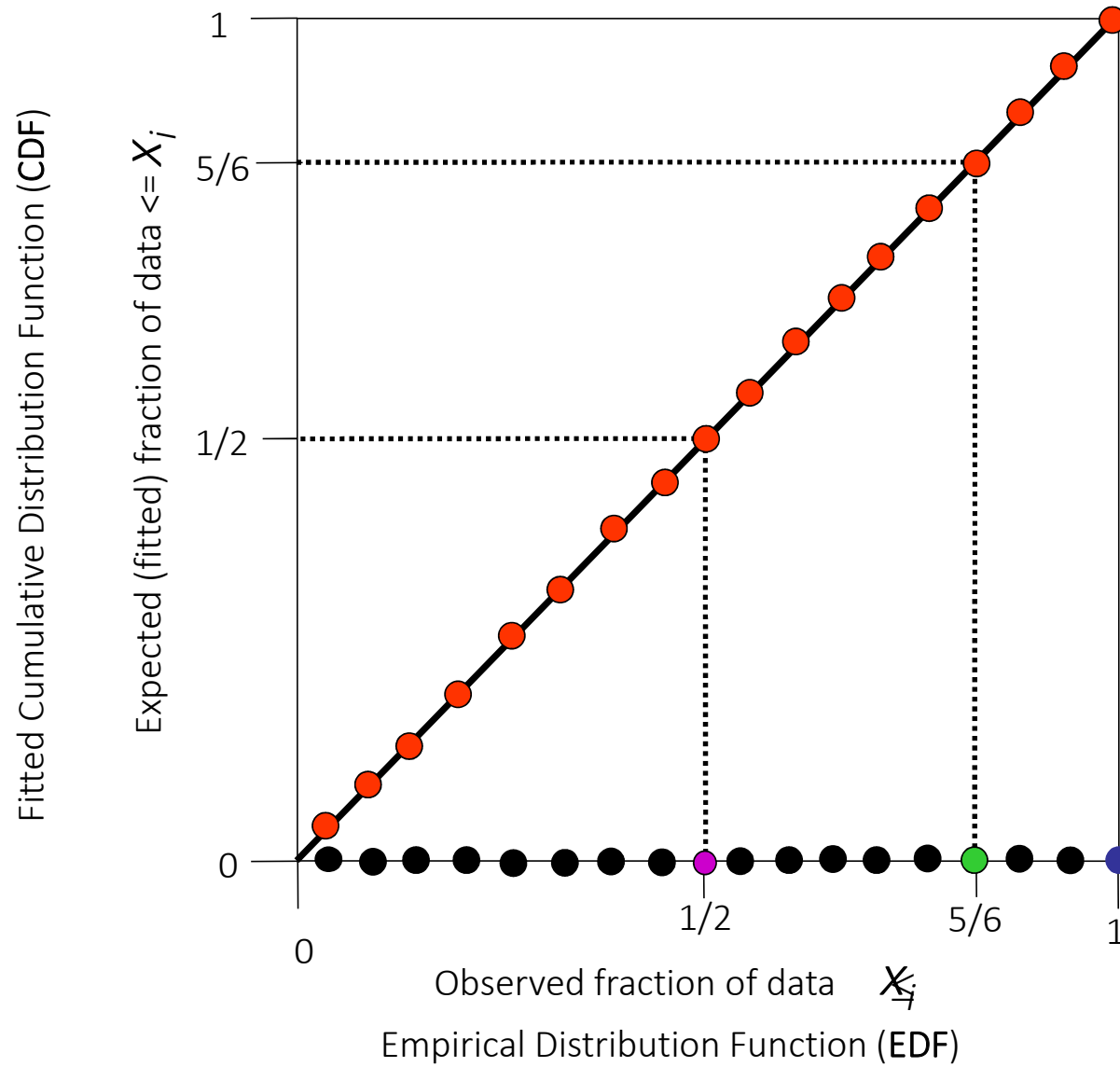
Probability of a greater chi-square value,  $P = 0.58322$

The program has limited capability for pooling. The user should judge the necessity for pooling and if necessary, do pooling by hand.

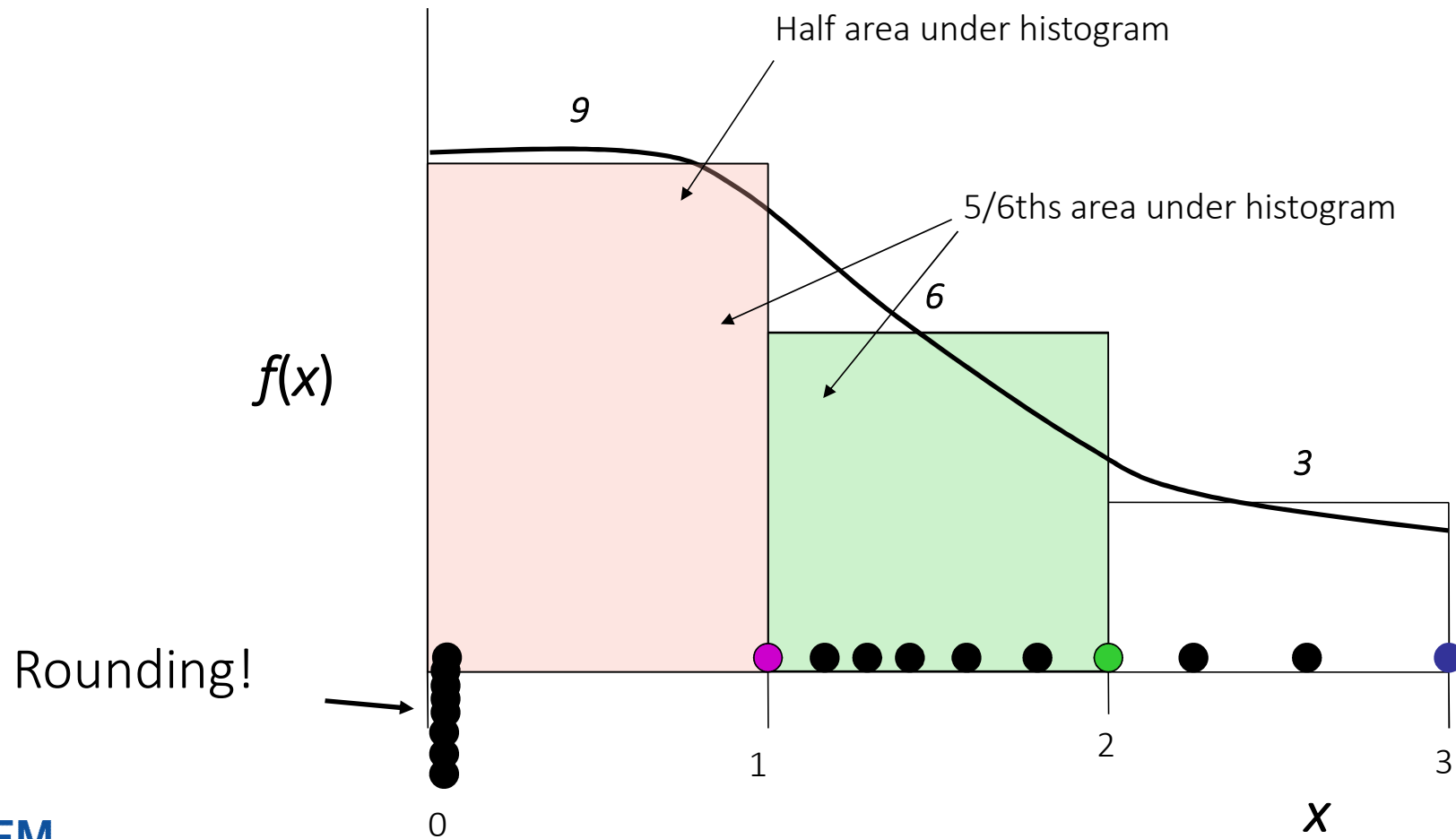


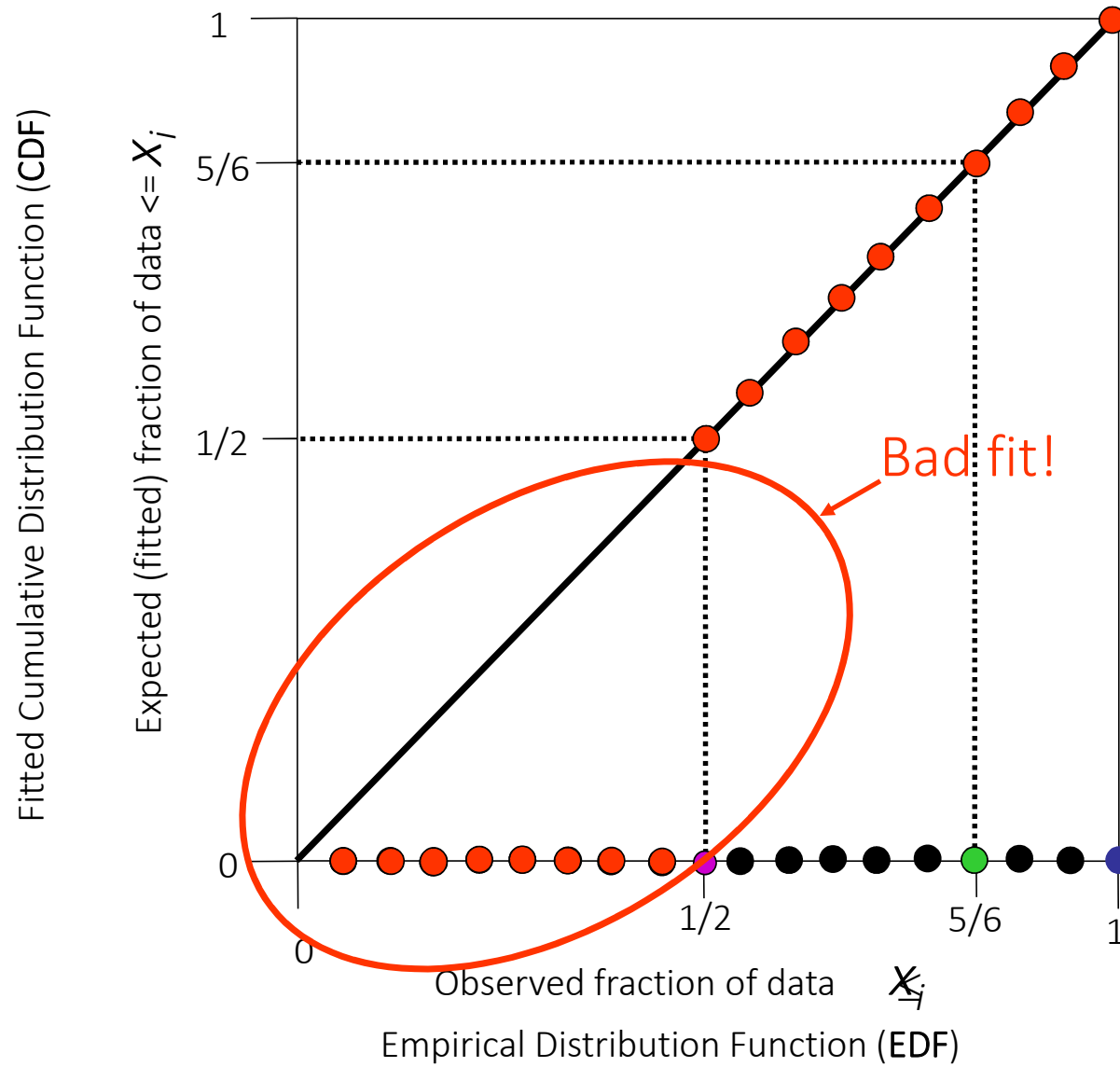
# Q-Q Plots and Related Tests



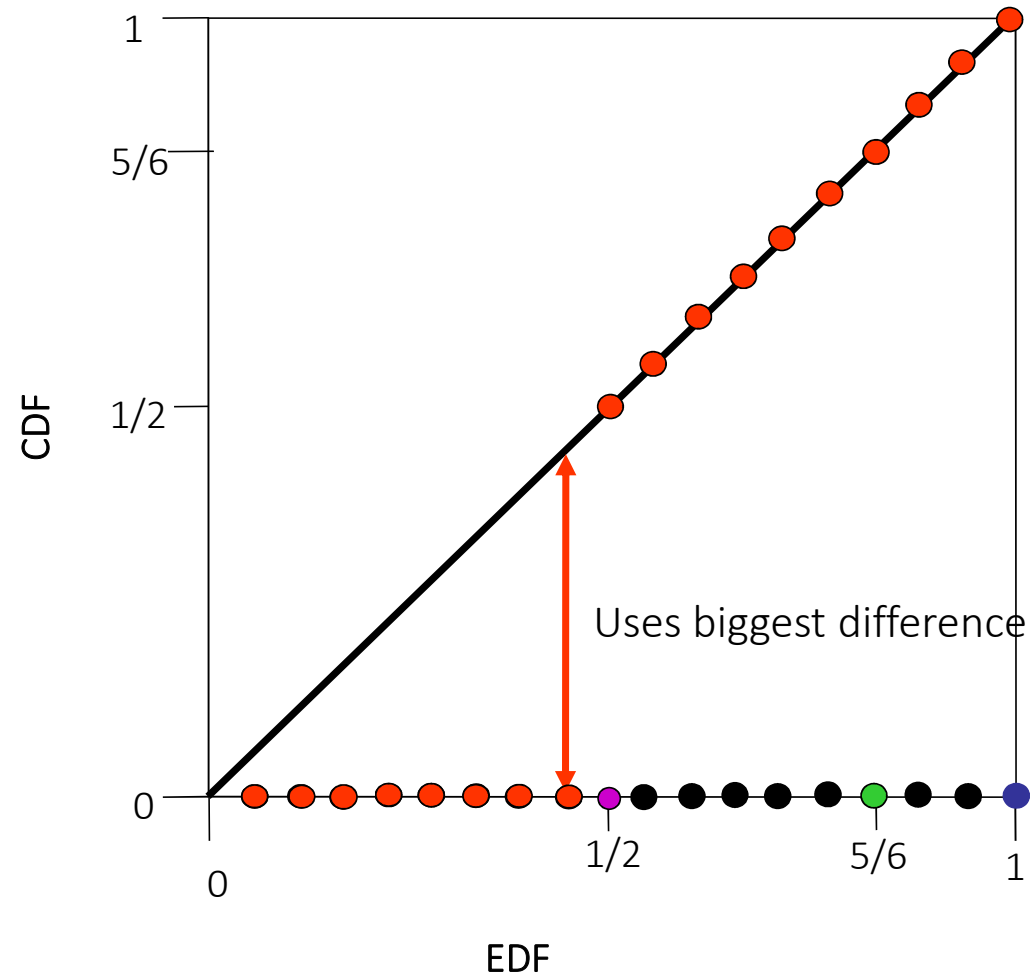


# Example: Rounding to zero

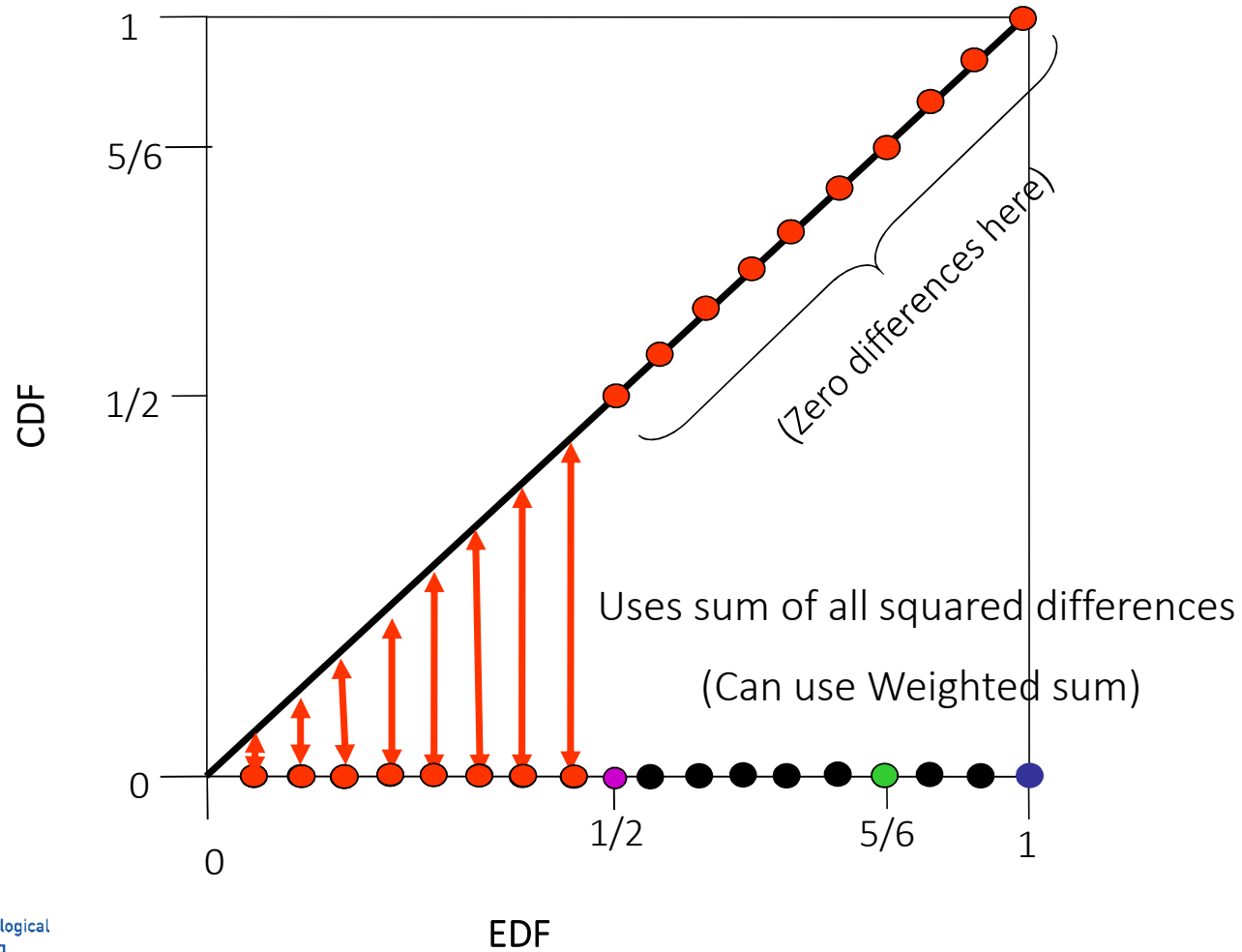




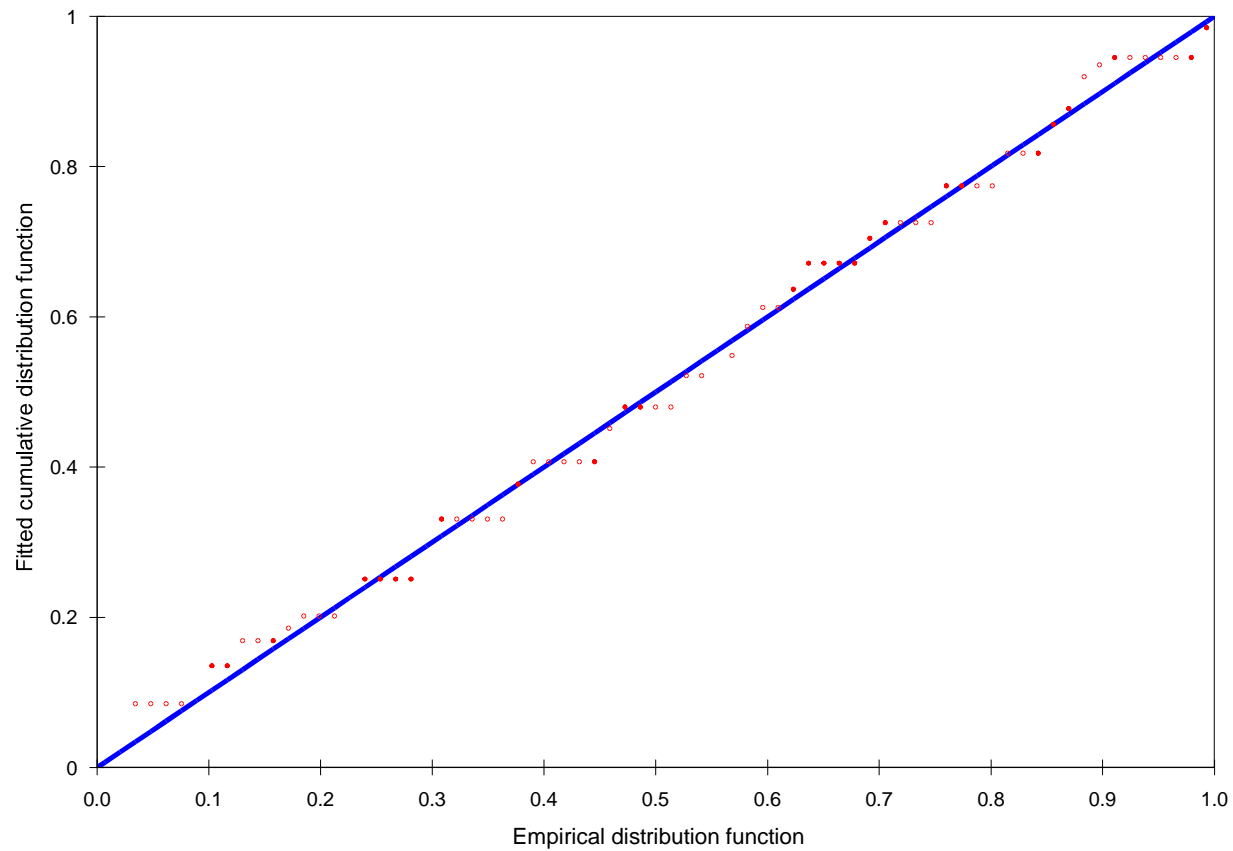
# Kolmogorov-Smirnov test



# Cramér-von Mises test



# Chaffinch line transect Q-Q plot



# K-S test and Cramer-von Mises test

## Kolmogorov-Smirnov test

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D\_n = 0.0573 p = 0.9703

## Cramer-von Mises family tests

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W-sq (uniform weighting) = 0.0368 0.900 < p <= 1.000

Relevant critical values:

W-sq crit(alpha=0.900) = 0.0000

C-sq (cosine weighting) = 0.0257 0.900 < p <= 1.000

Relevant critical values:

C-sq crit(alpha=0.900) = 0.0000



# Q-Q Plot Summary

- Q-Q plots show goodness-of-fit at “high resolution” – without requiring grouping into intervals
- Kolmogorov-Smirnov test and Cramér-von Mises test are goodness-of-fit tests that do not require grouping
- Cramér-von Mises test can be weighted, to give higher weight to  $x$  near zero