# Estimating the width of uniform distribution under measurement errors 

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## Convolution model

- $X=U+\varepsilon$
- $U \mathrm{i} \varepsilon$ independent, random
- $U$ uniform, $\varepsilon$ normal error

Example: $U \sim \mathcal{U}(-a, a), \varepsilon \sim \mathcal{N}\left(0, \sigma^{2}\right)$

$$
f_{X}(x ; a, \sigma)=\frac{1}{2 a}\left(F_{\varepsilon}\left(\frac{x+a}{\sigma}\right)-F_{\varepsilon}\left(\frac{x-a}{\sigma}\right)\right)
$$



## Applications in metrology

- Usually assumed that $\varepsilon$ is normally distributed.
- Known as the Flatten-Gaussian distribution
- The basis for calculating the measurement uncertainty
- Blázquez, J., García-Berrocal, A., Montalvo, C., Balbás, M. (2008). The coverage factor in a Flatten-Gaussian distribution, Metrologia 45, 503-506.
- Fotowicz, P. (2014). Methods for calculating the coverage interval based on the Flatten-Gaussian distribution. Measurement, 55, 272-275


## Applications in geology

- Cox, A.V., Dalrymple, G.B. (1967). Statistical analysis of geomagnetic reversal data and the precision of potassium-argon dating, J Geophys Res 72:2603-2614
- Agterberg, F.P. (1988). Quality of time scales - a statistical appraisal. In: Merriam, D.F. (ed) Current trends in geomathematics, Plenum, New York, pp 57-103
- Agterberg, F. P. (2014). Geomathematics: Theoretical foundations, applications and future developments. Quantitative geology and geostatistics (Vol. 18), Springer, Heidelberg.
- Ex. - for estimating the age of stage boundaries.


## Applications in electrical engineering

- For modeling the transmission losses data Tolić, I., K. Miličević, N. Šuvak, I. Biondić (2017). Non-linear Least Squares and Maximum Likelihood Estimation of Probability Density Function of Cross-Border Transmission Losses, IEEE Transactions on Power Systems 33/2 (2018), 2230-2238


## Fitting a line segment to noisy data

- Uniform latent variable and normal noise Davidov, O., Goldenshluger, A. (2004). Fitting a line segment to noisy data. Journal of Statistical Planing and Inference 119, 191-206.
- Linear structural relationship

Chan, N.N. (1982). Linear structural relationships with unknown error variances, Biometrika, 69, No.1, 277-279

## Estimating the size of an object in a noisy image

- M. B., K. Sabo, Border estimation of a Two-dimensional Uniform Distribution if Data are Measured with Additive Error, Statistics, 41 (2007), 4, 311-319.
- K. Sabo, M. B., Border estimation of a disc observed with random errors solved in two steps, Journal of Computational and Applied Mathematics, 229 (2009)


## Black fungi colonies



Garlipp, T., Müller, C. H., Detection of linear and circular shapes in image analysis, Computational Statistics \& Data Analysis 51 ( 2006), 1479-1490


## R package

## https : //cran.r - project.org/web/packages/LeArEst/

- M. Benšić, P. Taler, S. Hamedović, E.K. Nyarko, K. Sabo, LeArEst: Length and Area Estimation from Data Measured with Additive Error, The R Journal 9/2 (2017), 461-473
- Includes web application for estimating the size of an object from a noisy image
- Gaussian error model - tested in simulations


## LeArEst

(i) localhost:5656/ocpu/library/LeArEst/www/index_esttest.htral


## Length Estimator



## LeArEst



## LeArEst

Data successfully prepared.
Histogram of data


Click for raw R output

## LeArEst

```
0. (i) localhost:5656/ocpu/library/LeArEst/www/index_esttest.htm
```


Length Estimator



Levels of grey. 8, Box size. 10. Line thickness. 5
Error distribution: laplace, Error standard deviation: ML, Confidence level: 0.95
Length. 311.28 pixel width ( $52.85 \%$ of the image wiath). Green line length. 428.12 pixel width Standard deviation: 18.24 (ML estimated)
Method. Asymptotic distribution of LR statistic
Conficience interval: $(310.04,312.58)$
Click for raw R output

## Detection of the circle edge in a noisy image

Image of the can, $r=51.5 \mathrm{~mm}$, area $=8332.3 \mathrm{~mm}^{2}$


## Detection of the circle edge in a noisy image

- Hough transformation
P. V. Hough, Method and means for recognizing complex patterns.

Patent U.S. Patent No. 3,069,654, 1962.
R. D. Duda i P. E. Hart, Use ofthe Hough transform to detectlinesand curvesin pictures, Commun. ACM, 15, br. 1, 1972.
A. Rosenfeld, Picture processing by computer, ACM Computing Surveys (CSUR), svez. 1, br. 3, pp. 147-176, 1969.

- EDCirlces
C. Akinlar i C. Topal, EDCircles: A real-time circle detector with a false detection control, Pattern Recognition, 46, br. 3, pp. 725-740, 2013.
- Fornaciari
M. Fornaciari, A. Prati i R. Cucchiara, A fast and effective ellipse detector for embedded vision applications, Pattern Recognition, 47, br. 11, pp. 3693-3708, 2017.


## Simulation results - can example

- $r=51.5 \mathrm{~mm}$, area $=8332.3 \mathrm{~mm}^{2}$
- 20 simulations for each sd
- RMSE for area estimation

| $s d / r(\%)$ | LeArEst | Hough | EDCircles | Fornaciari |
| :---: | :---: | :---: | :---: | :---: |
| 0 | 229 | 1858 | 681 |  |
| 5 | 587 | 1353 | 723 |  |
| 13 | 523 | 1365 | 852 |  |
| 26 | 243 | 2315 | 659 |  |
| 39 | 289 | 1955 | 755 |  |

- For 1000 simulations we have similar results but withouth EDCircles.


## Problem with real images



## Laplace error model - improvement



## General one-dimensional model

- $X=U+\sigma \varepsilon$
- $f_{U}(t)=\frac{1}{2 a} I_{[-a, a]}(t)$
- $\varepsilon$ absolutely continuous random variable with distribution $F_{\varepsilon}$ and density function $f_{\varepsilon}$, which is even.

$$
\Rightarrow \quad f_{X}(t)=\frac{1}{2 a}\left[F_{\varepsilon}\left(\frac{a+t}{\sigma}\right)-F_{\varepsilon}\left(\frac{a-t}{\sigma}\right)\right]
$$

- $\varepsilon \sim \mathcal{N}(0,1)$

$$
f_{X}(t \mid a, \sigma)=\frac{1}{2 a \sqrt{2 \pi}} \int_{-\frac{a+t}{\sigma}}^{\frac{a-t}{\sigma}} e^{-\frac{u^{2}}{2}} d u
$$

- $\varepsilon \sim \mathcal{L}(\lambda), \sigma=1$

$$
f_{X}(t \mid a, \lambda)= \begin{cases}\frac{1}{2 a} \sinh \frac{a}{\lambda} e^{-\frac{|t|}{\lambda}}, & a \in(0,|t|] \\ \frac{1}{2 a}\left(1-e^{-\frac{a}{\lambda}} \cosh \frac{t}{\lambda}\right), & a \in(|t|, \infty)\end{cases}
$$

## General one-dimensional model - density


red - normal error
blue - Student's $t(1)$ error

## Maximum likelihood estimation

- Data $\left(x_{1}, \ldots, x_{n}\right)$, log-likelihood:

$$
I(a)=-n \log 2 a+\log \sum_{i=1}^{n}\left[F_{\varepsilon}\left(\frac{x_{i}+a}{\sigma}\right)-F_{\varepsilon}\left(\frac{x_{i}-a}{\sigma}\right)\right]
$$

- Differentiable function (model with absolutely continuous error).
- Optimization should be easy with a good initial approximation that is not difficult to recognize in applications.
- Regularity conditions???


## MLE - regularity conditions

- Easy to check sufficient conditions
- S. Hamedović, MB, K. Sabo, Estimating the width of a uniform distribution under symmetric measurement errors, submitted 2019
- Examples: Normal, Logistic, Student's $t(\nu), \nu \geq 1$
- Student's distribution with small degrees of freedom could be a particularly good choice for the error. (We can adjust the number of degrees of freedom to the amount of outliers in the data.)


## General regular model

## ML confidence intervals

Fisher information

$$
\begin{gathered}
I(a)=\frac{-1}{a^{2}}+\frac{1}{2 a \sigma^{2}} \int_{-\infty}^{\infty} \frac{\left(f_{\varepsilon}\left(\frac{x+a}{\sigma}\right)+f_{\varepsilon}\left(\frac{x-a}{\sigma}\right)\right)^{2}}{F_{\varepsilon}\left(\frac{x+a}{\sigma}\right)-F_{\varepsilon}\left(\frac{x-a}{\sigma}\right)} d x \\
\left(\hat{a}_{M L}-\frac{z_{1-\alpha / 2}}{\sqrt{n I\left(\hat{a}_{M L}\right)}}, \hat{a}_{M L}+\frac{z_{1-\alpha / 2}}{\sqrt{n I\left(\hat{a}_{M L}\right)}}\right)
\end{gathered}
$$

## General model

## LR test

$$
\begin{aligned}
& H_{0}: a=a_{0} \\
& H_{1}: a \neq a_{0}
\end{aligned}
$$

Critical region (significance level $\alpha, L$ likelihood)

$$
\left\{\mathbf{y} \left\lvert\,-2 \log \frac{L\left(a_{0} ; \mathbf{y}\right)}{L\left(\hat{a}_{M L} ; \mathbf{y}\right)} \geq \chi_{1}^{2}(1-\alpha)\right.\right\}
$$

## Updated R package

## https : //cran.r - project.org/web/packages/LeArEst/

## Eye pupil



## Area Estimator

Loaci Prcture... eye3.jpeg

## Data

| Levels of <br> groy | Box size | Line <br> thickness |
| :--- | :--- | :--- |
| 4 | 20 | 1 |
|  |  |  |
| Number of <br> slices | Slicing | Parallelization |
| 10 | Star | Off |


| Object <br> ongntness | Represent <br> ooject as |
| :--- | :--- |
| dark | clrcle |

## Estimation

| Error <br> distribution | Error <br> standard deviation |
| :--- | :--- | :--- |
| T 4 | ML estimator |

$$
\mathrm{E} \text { Estimate }
$$




## Eye pupil



## Area Estimator

```
Load Picture... eye3.jpeg
```


## Data

| Levels of <br> gray | Boxslze | Line <br> thickness |
| :--- | :--- | :--- |
| 4 | 20 | 1 |


| Number of <br> slices | Slicing | Parallelization |
| :--- | :--- | :--- |
| 10 | Star | Off |


| Ooject <br> brighthess | Represent <br> object as |
| :--- | :--- |
| dark | v circle |

Estimation


Welcomel Click on Load picture (must be JPEG tormat), choose upper left and lower right points of the rectangle surrounding the measured


## Eye pupil




## Eye pupil





| Area Estimator |
| :--- |
| Data |
| Levels of <br> gray |
| 4 Box size Line <br> thickness   |


| Number of slices | Slicing | Parallelization |
| :---: | :---: | :---: |
| 10 | Star $\quad$ - | Off |
| Object brightness | Represent object as |  |
| dark * | circle * |  |

## Estimation



Welcomel Cick on Load picture imust be JPEG format), choose upper left and lower right points of the rectangle surrubnding the measured object, set data parameters and cllck on Estimate.

Please use proportional screen resolution, eg $1920 \times 1080$ if you use display with 16.9 aspect fatio. or $1920 \times 1200$ in the case of 1610 asoet
$\square$

## Blood artery



## Blood artery



## Blood artery



## Length Estimator

| Laad Picture .. s7.jpg |
| :--- |
| Data |
| Levels of |
| gray |
| 10 |

c. Prepare data

## Estimation

|  | Error <br> Error | standard <br> deviation | Confidence <br> Ievel |
| :--- | :--- | :--- | :--- |
| T1 | ML estirr | - | 0.95 |

Ea Estimate

## Testing

Ho value Unit
$10 \quad$ pixel wioth

## Aternative

two-slde .

