

What is persistence?

TDA group at KTH Stockholm

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Persistence is

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A collection of
data sets

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triangular inequality

symmetry

$d(X,Y)=0$ does not imply $X=Y$

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measurable
functions

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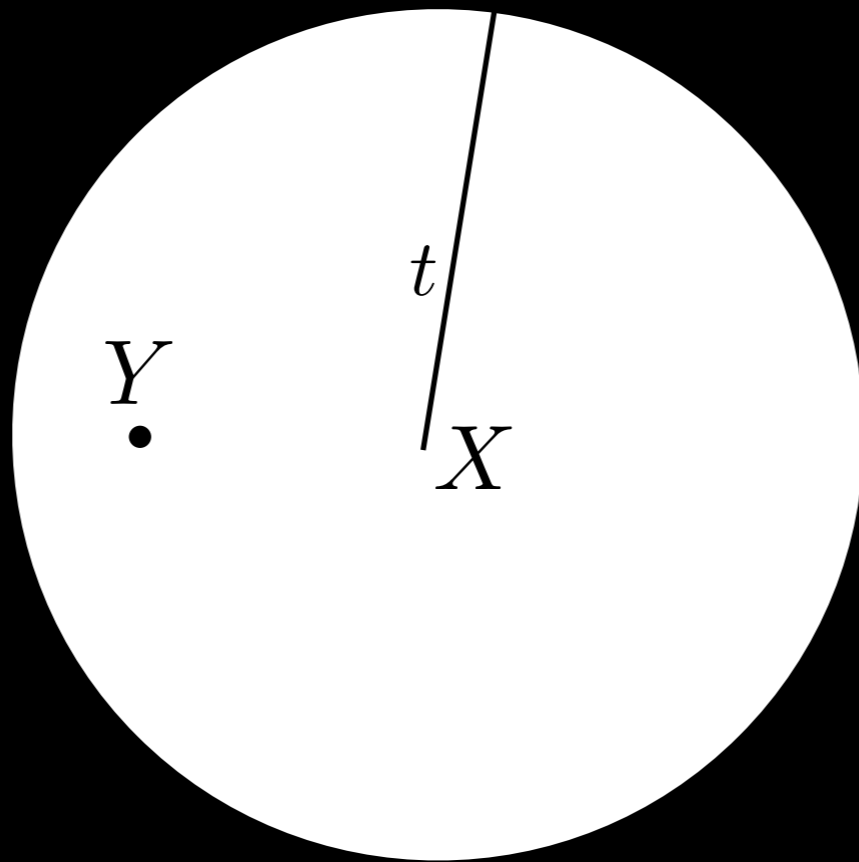
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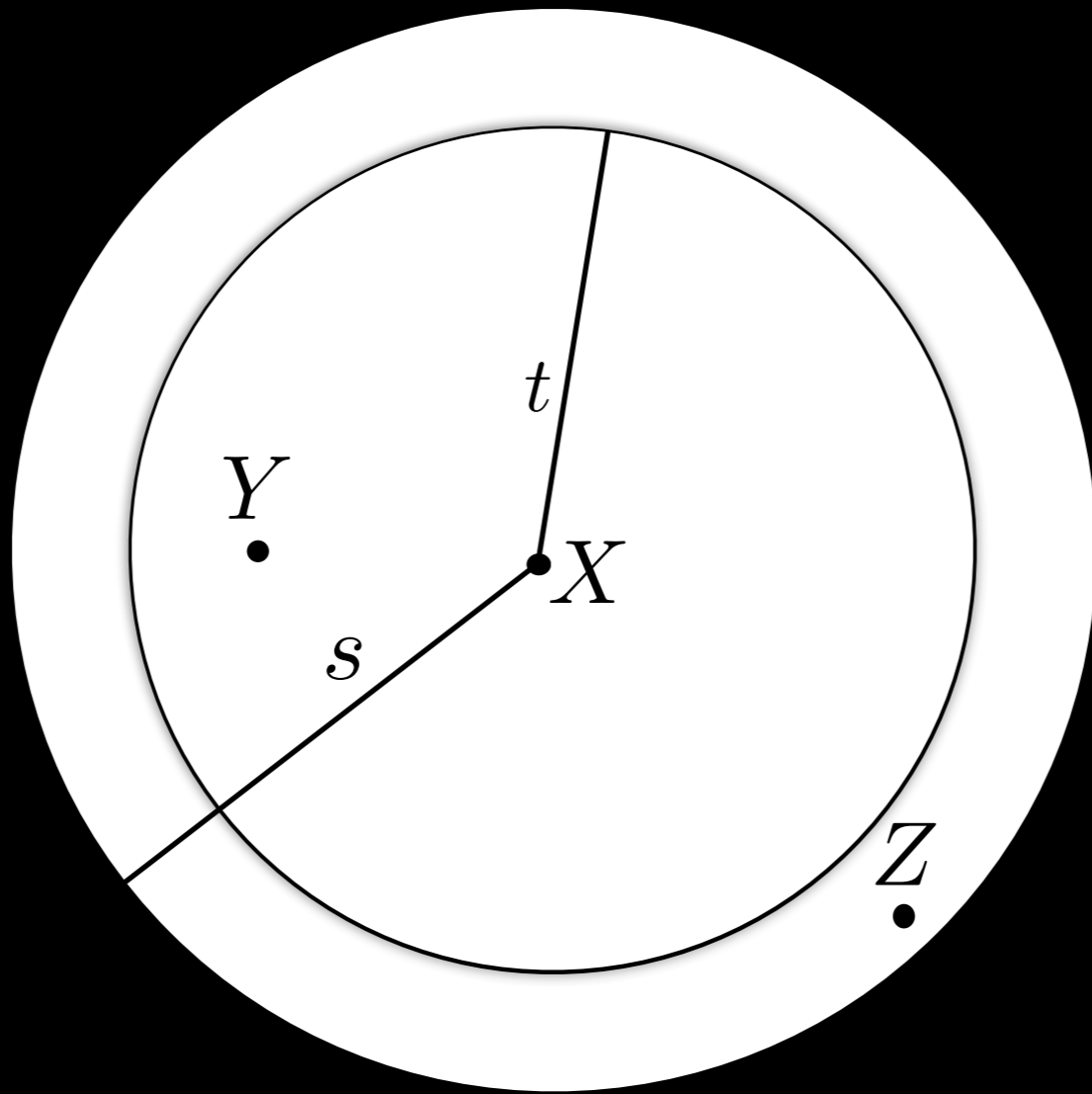


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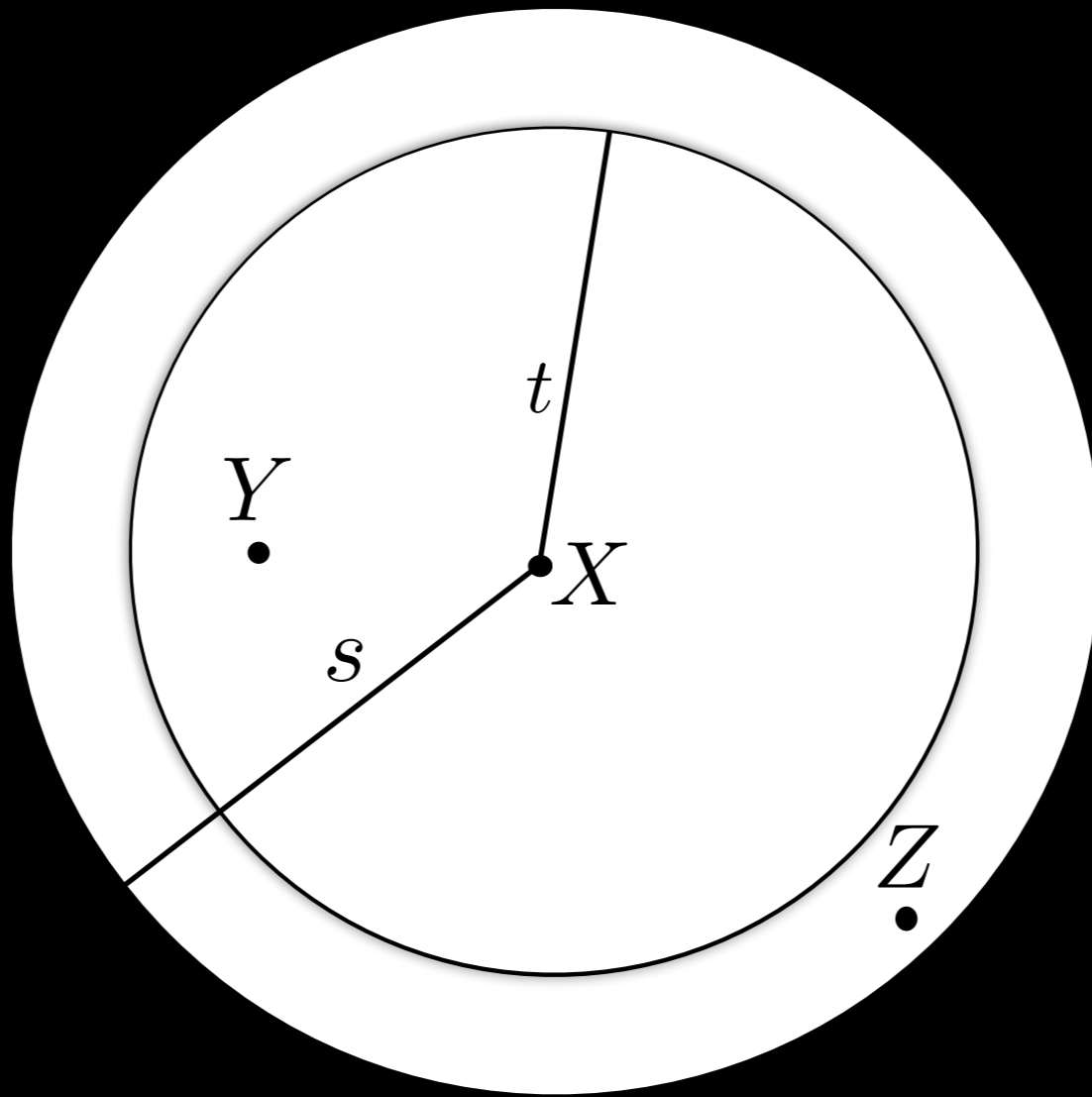
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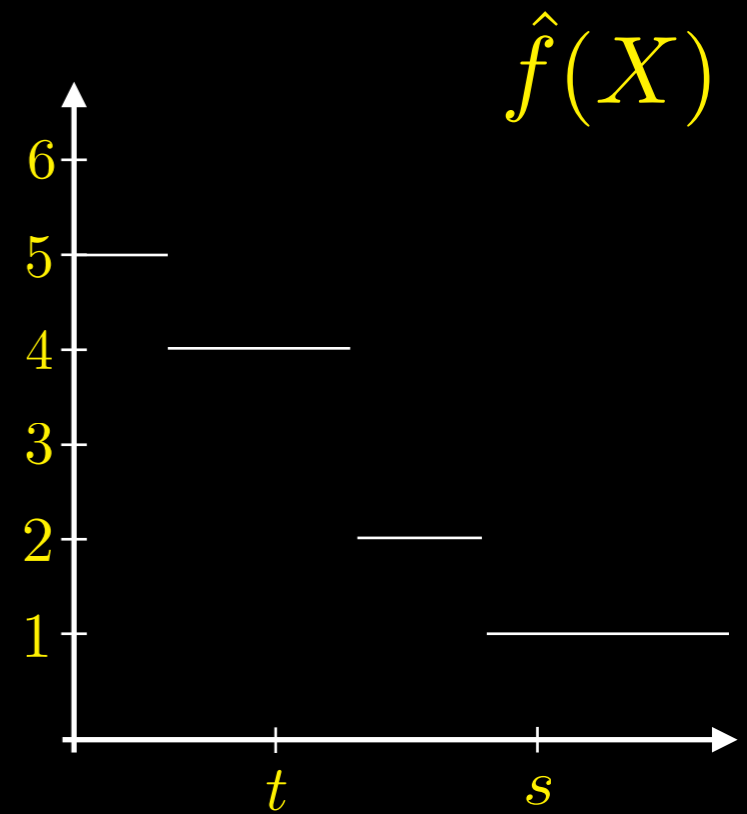
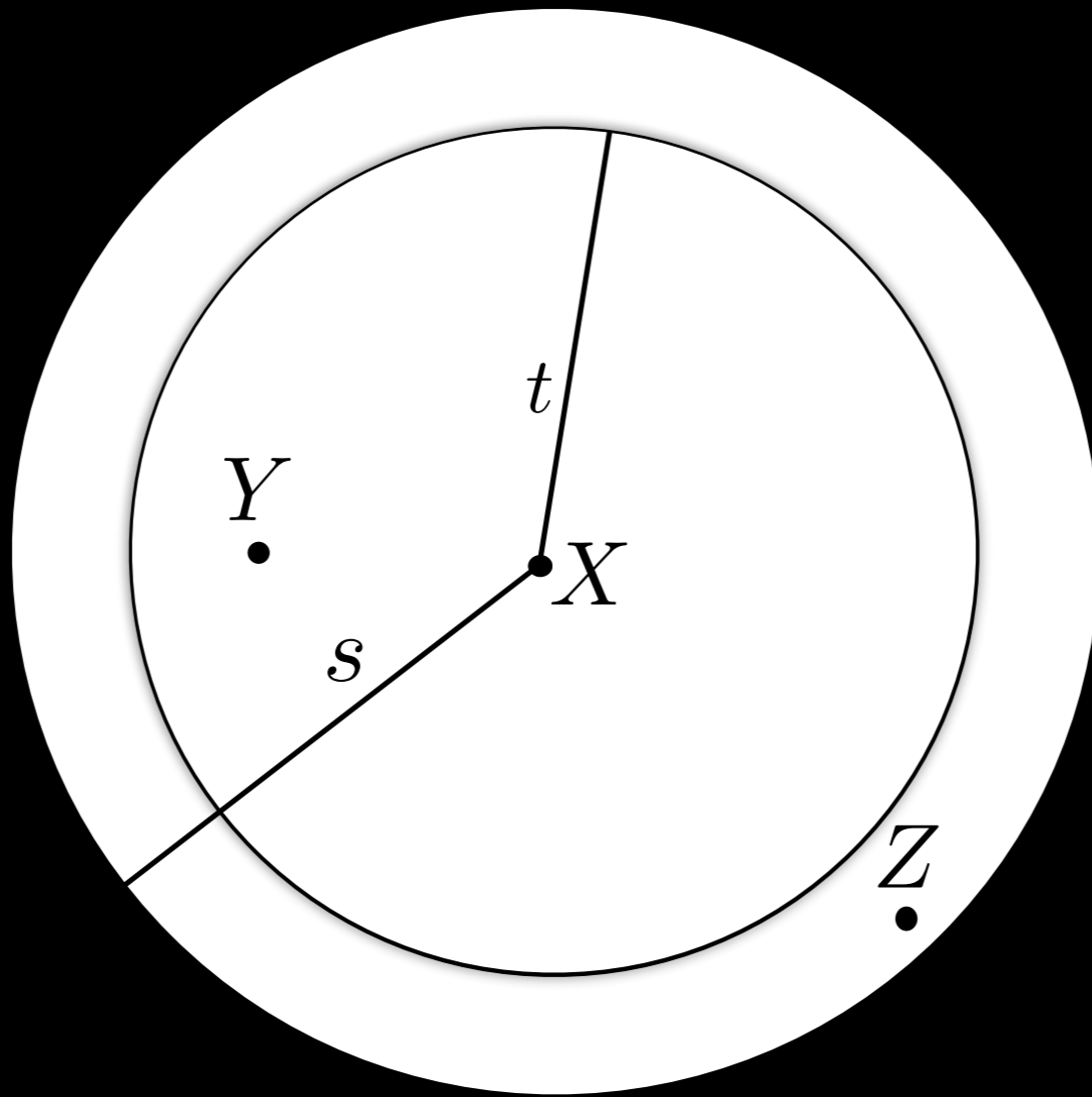
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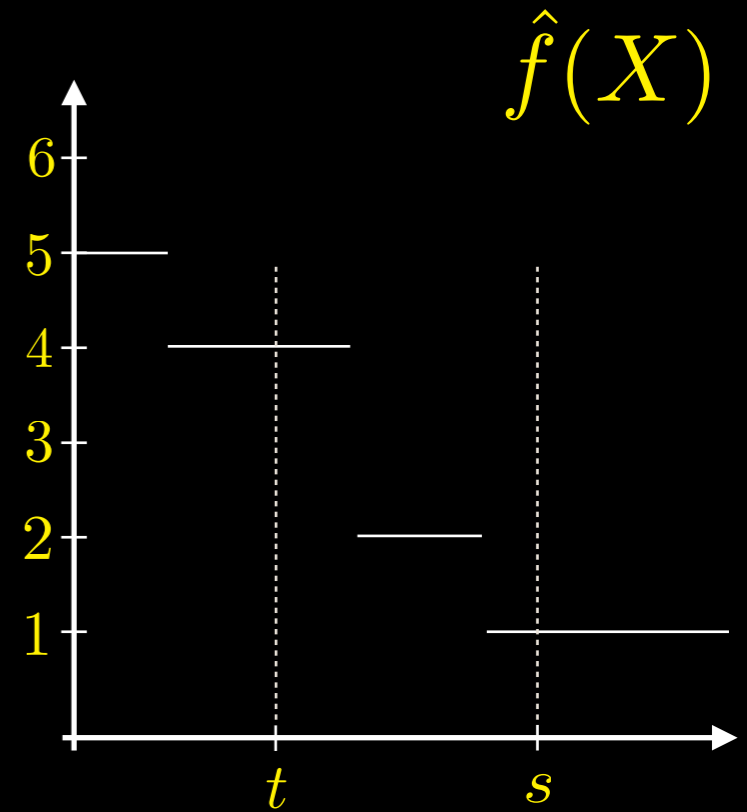
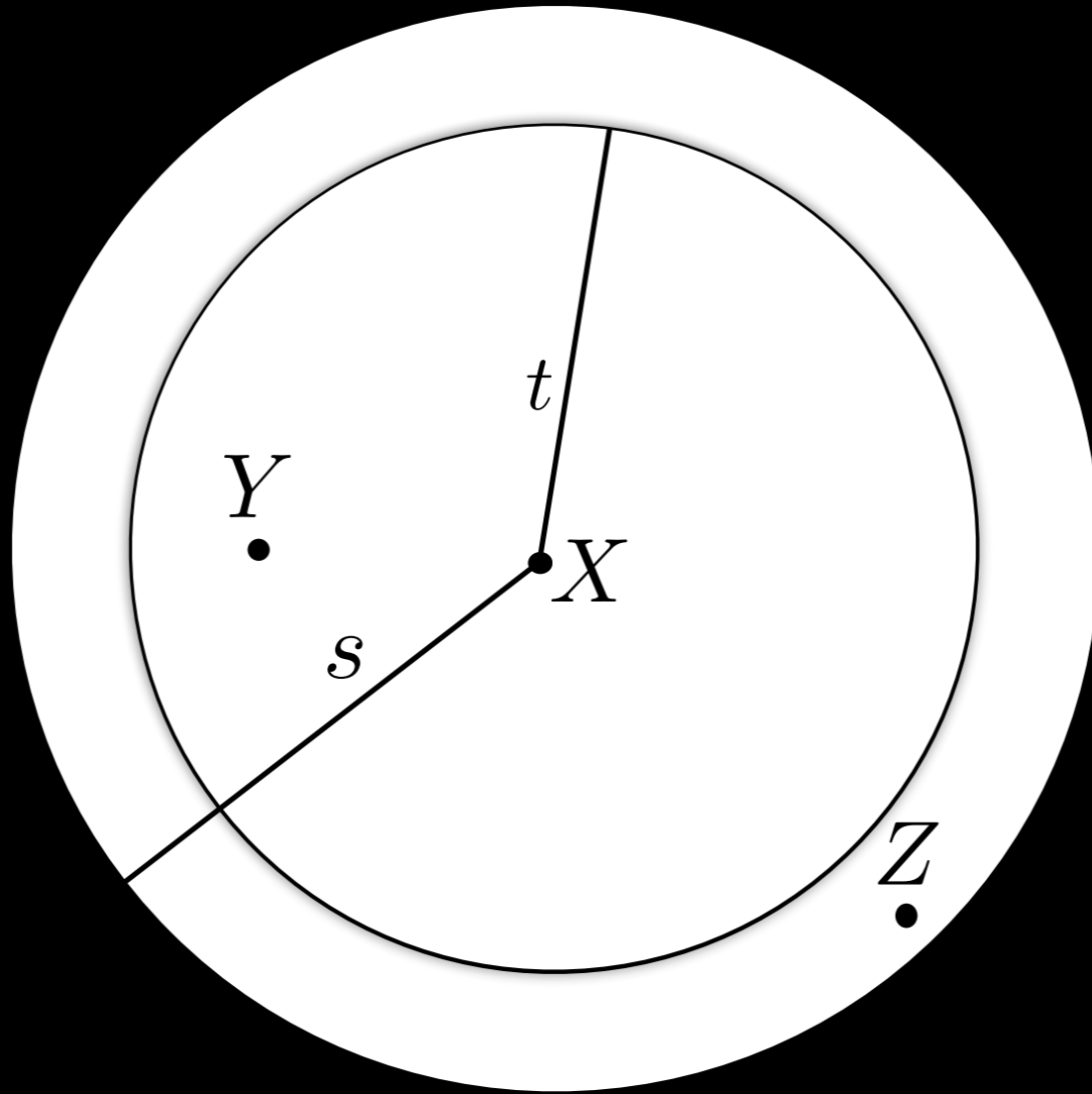
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Metrics on $\text{Func}(\mathbb{R}_{\geq 0}, \mathbb{R}_{\geq 0})$

Integral:

$$f(\alpha, \beta) = \int_{t=0}^{\infty} |\alpha - \beta| \quad \int_a^b (\alpha, \beta) = \int_a^b |\alpha - \beta|$$

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integral metric

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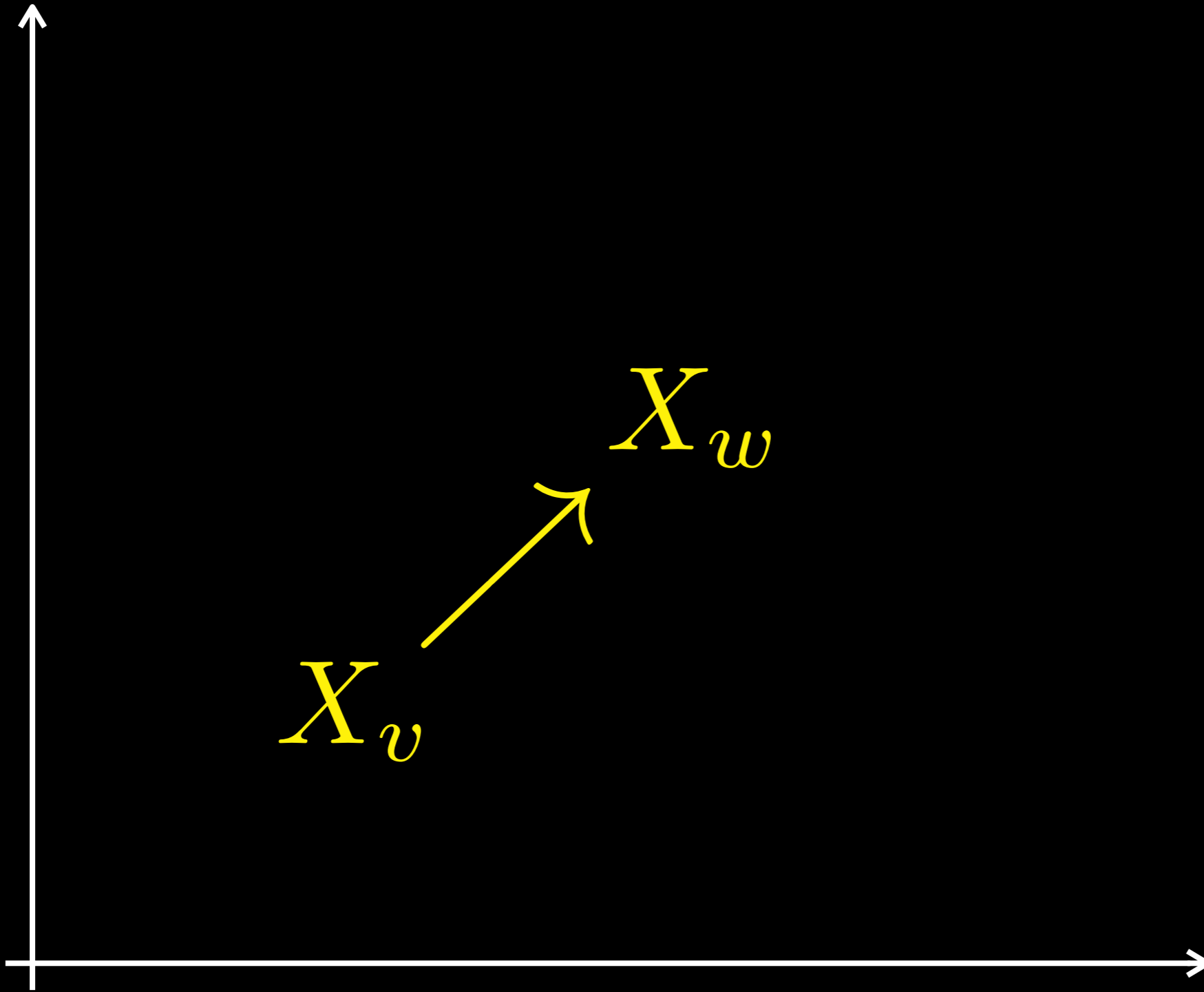
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Compact Tame
Vector Spaces
parametrized by

$$\mathbb{R}_{\geq 0}^n$$

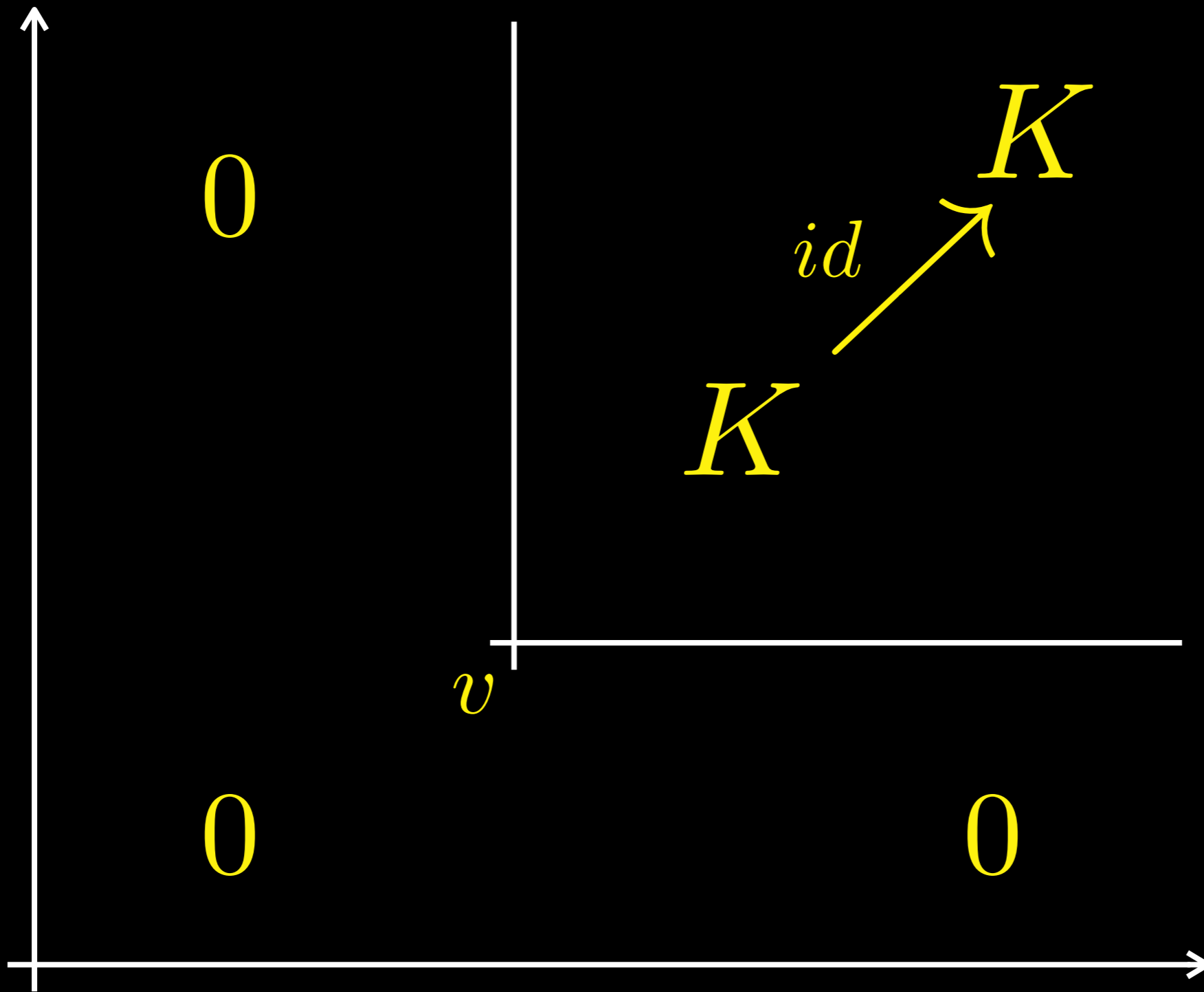
X

$\mathbb{R}_{\geq 0}^2$



$K(v, -)$

$\mathbb{R}_{\geq 0}^2$



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X has rank k if there is a surjection:

$$\bigoplus_{i=1}^k K(v_i, -) \rightarrow X$$

and k is the smallest such number

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Sets with n
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$$H_n(\text{VR}(-), K)$$

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Metrics on T compact tame $\mathbb{R}_{\geq 0}^n$ vector spaces

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Noise Systems



Simple Noise Systems



Contours (actions)

$$C: \mathbb{R}_{\geq 0}^n \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}^n$$

A noise system:

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Theorem. d is a metric on T

Metrics on T compact tame $\mathbb{R}_{\geq 0}^n$ vector spaces



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Contours are called superlinear families by
Bubenik, de Silva, Scott

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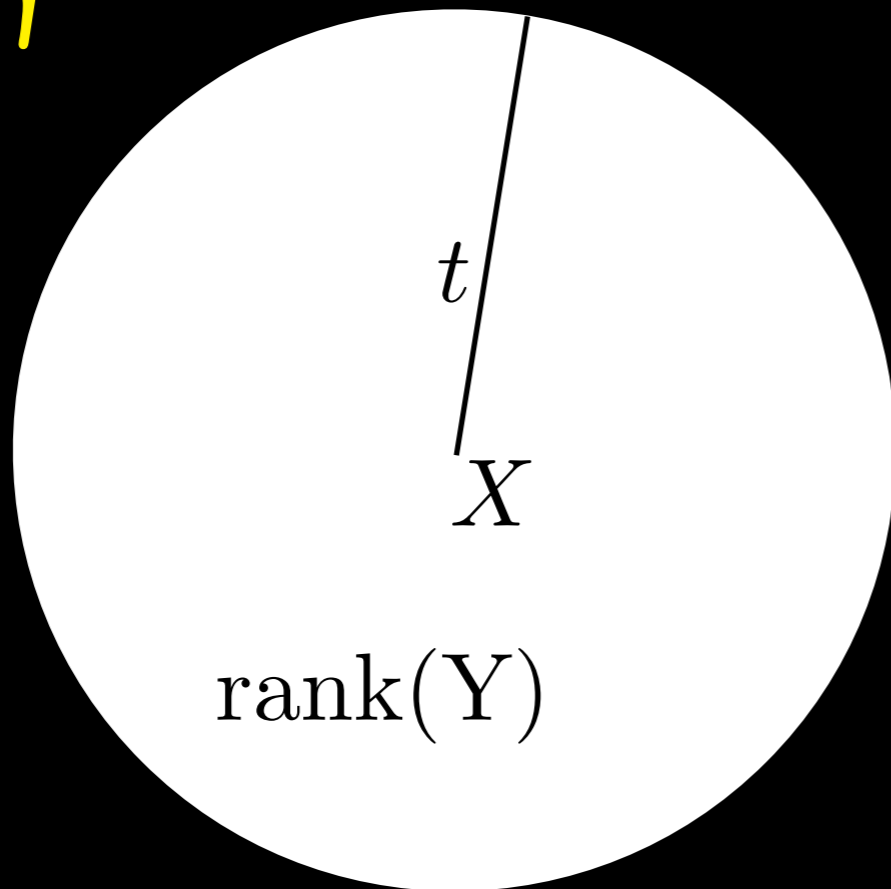


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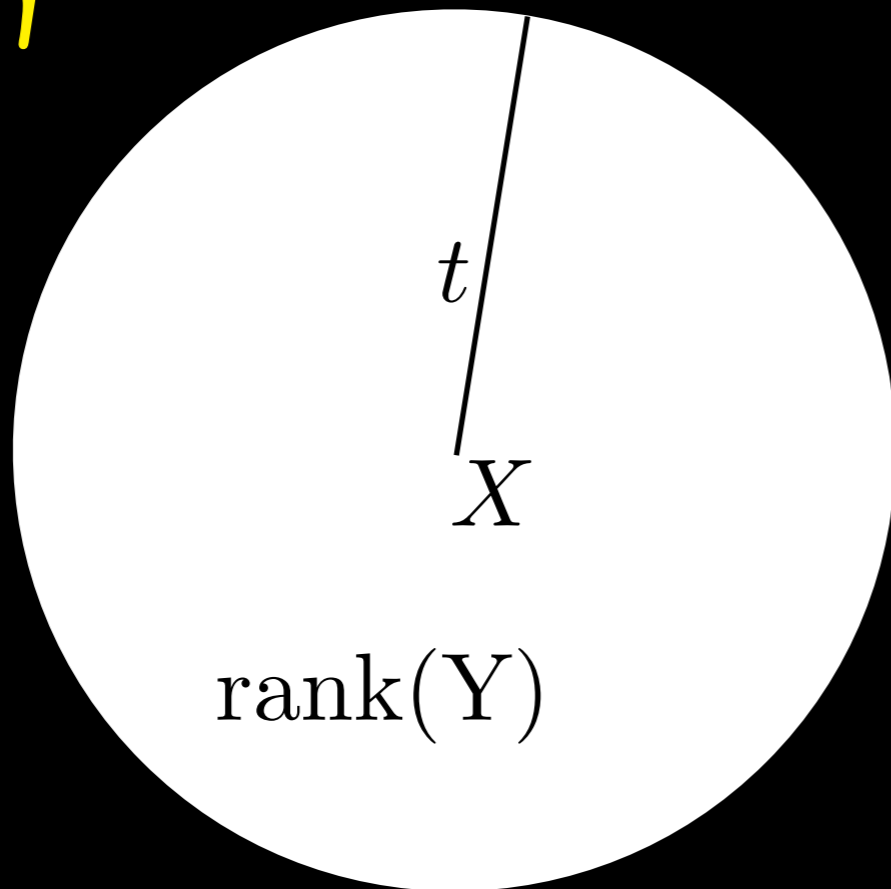
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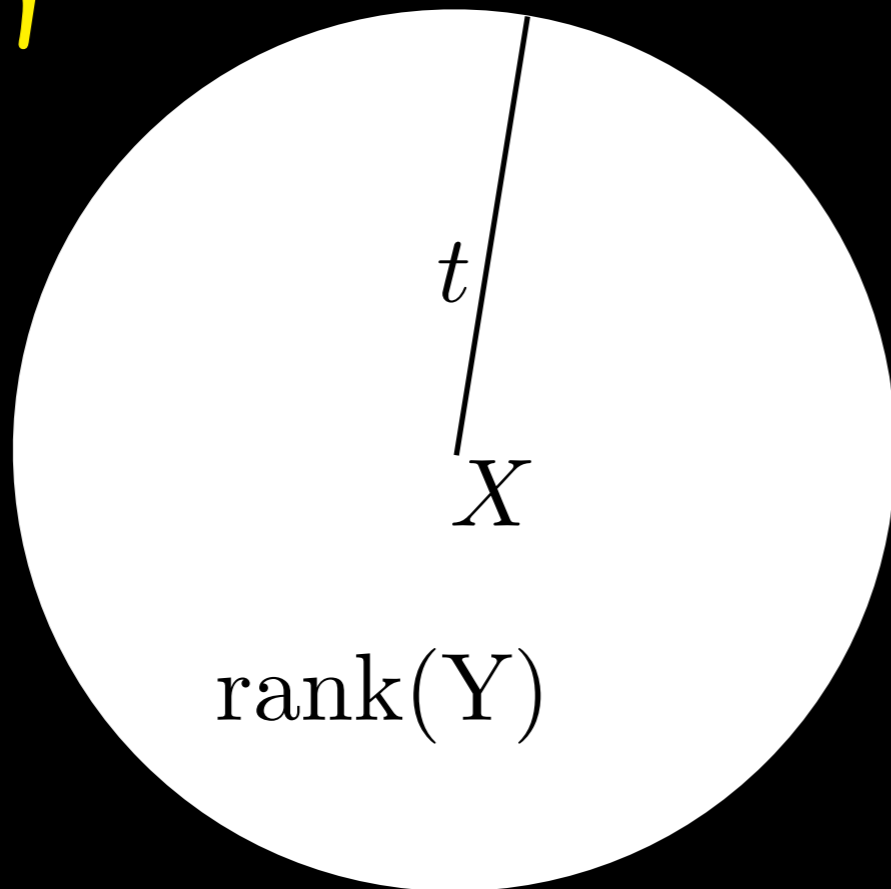


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Theorem

For a simple noise system, if the field is finite,

determining $\widehat{\text{rank}}(X)$

requires only finitely many calculations.

Aim: calculate the stabilization of $\text{rank}: T \rightarrow \mathbb{N}$

Let X, Y be compact tame \mathbb{R}^n parameterized vect sp

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$$\widehat{\text{rank}}(X)(t)$$

\equiv

$$|\{\text{bars } [b, d) \text{ in } X \mid C(b, t) < d\}|$$



parameter

isomorphic

Y

noise system

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$n>1$ parameter

Question

Construct non isomorphic

X, Y

s.t. for any noise system

$$\widehat{\text{rank}}(X) = \widehat{\text{rank}}(Y)$$

O. Gäfvert theorem:

Calculating $\widehat{\text{rank}}$
is an NP hard problem

$n=1$

Implementing contours

$n=1$ Implementing contours

A density is an integrable function $f: \mathbb{R}_{\geq} \rightarrow \mathbb{R}_{>0}$

$n=1$ Implementing contours

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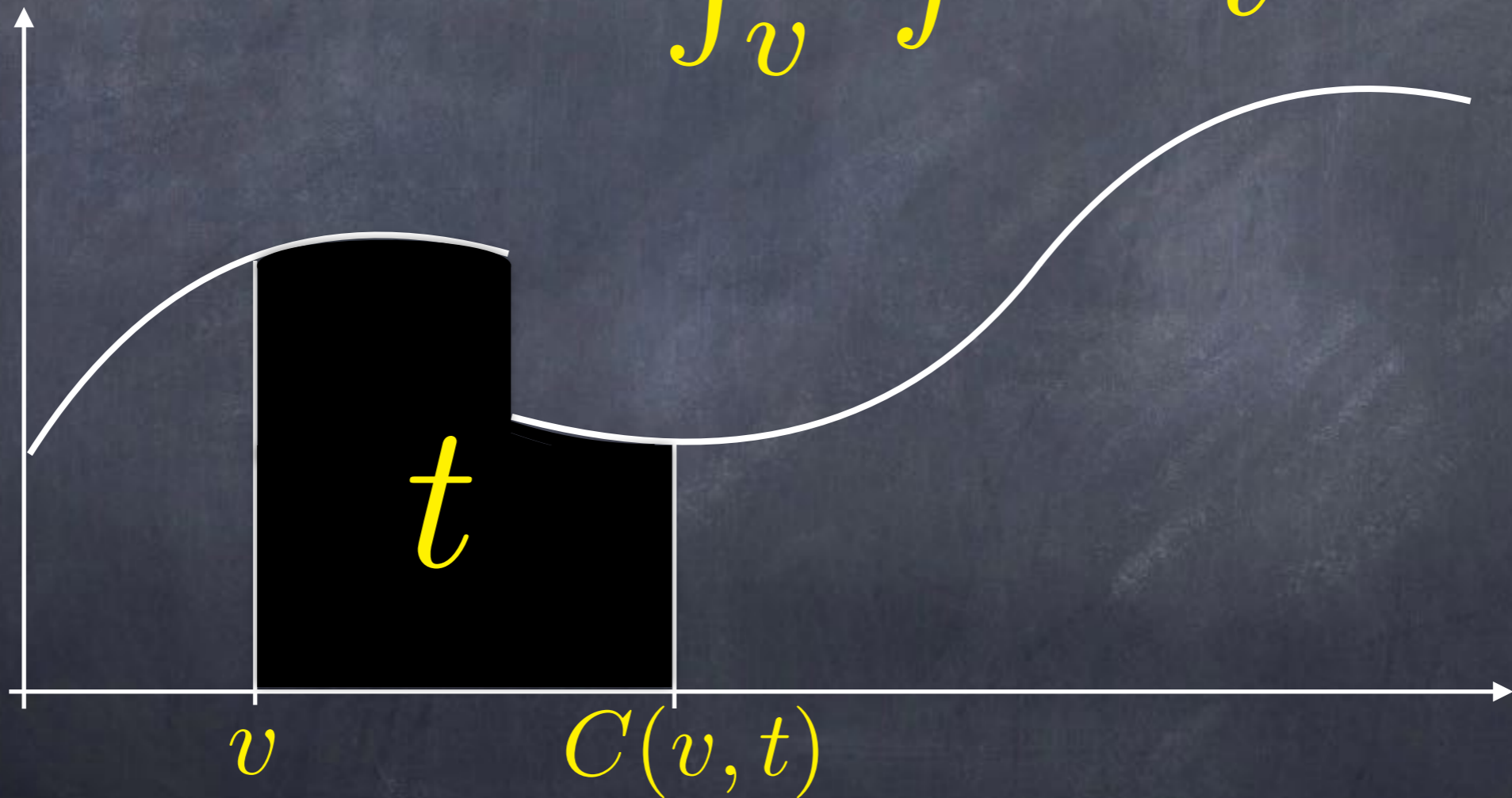
Define $C(v, t)$ to be the solution (unique) x

of the equation: $\int_v^x f = t$

$n=1$ Implementing contours

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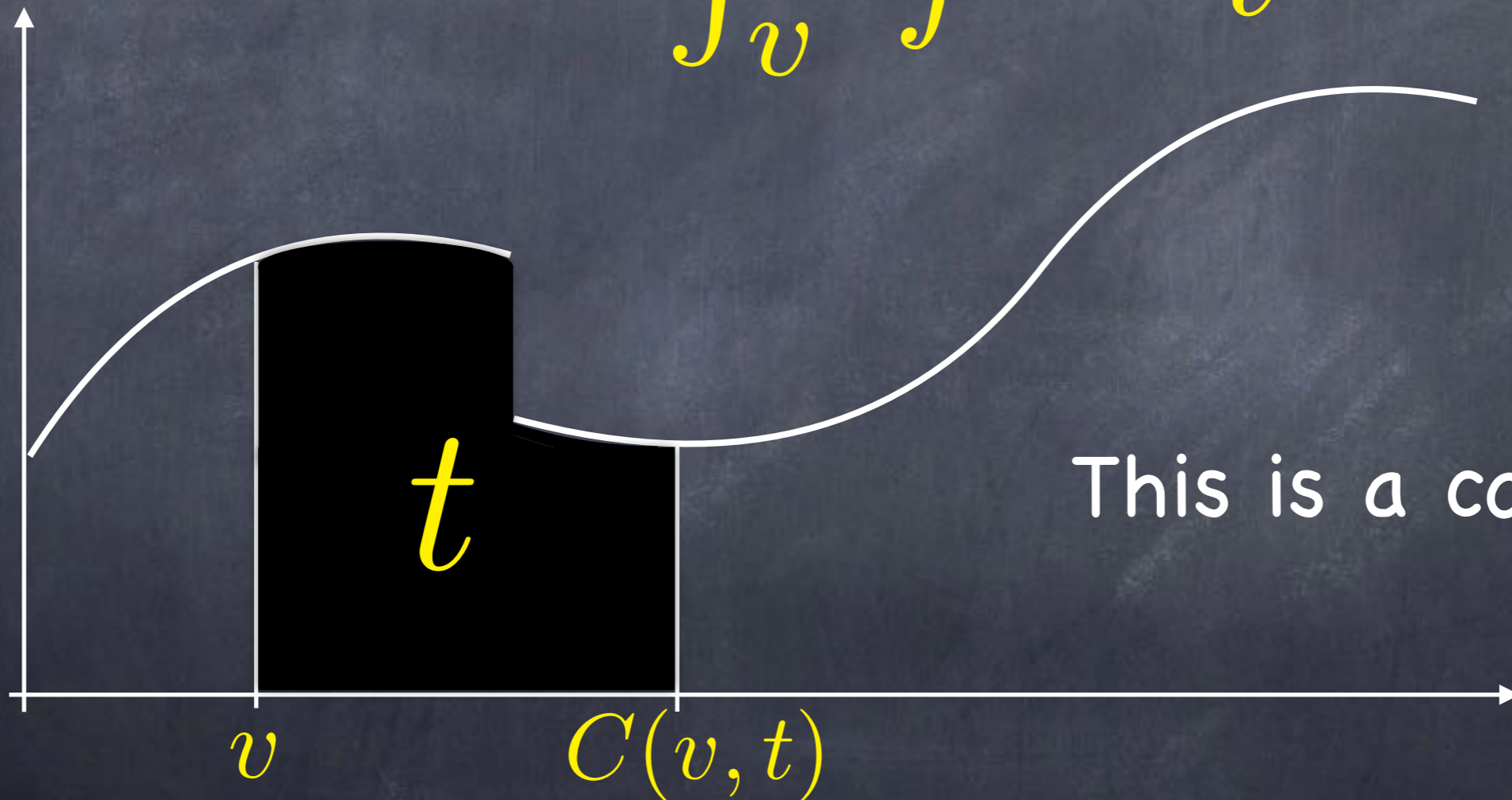


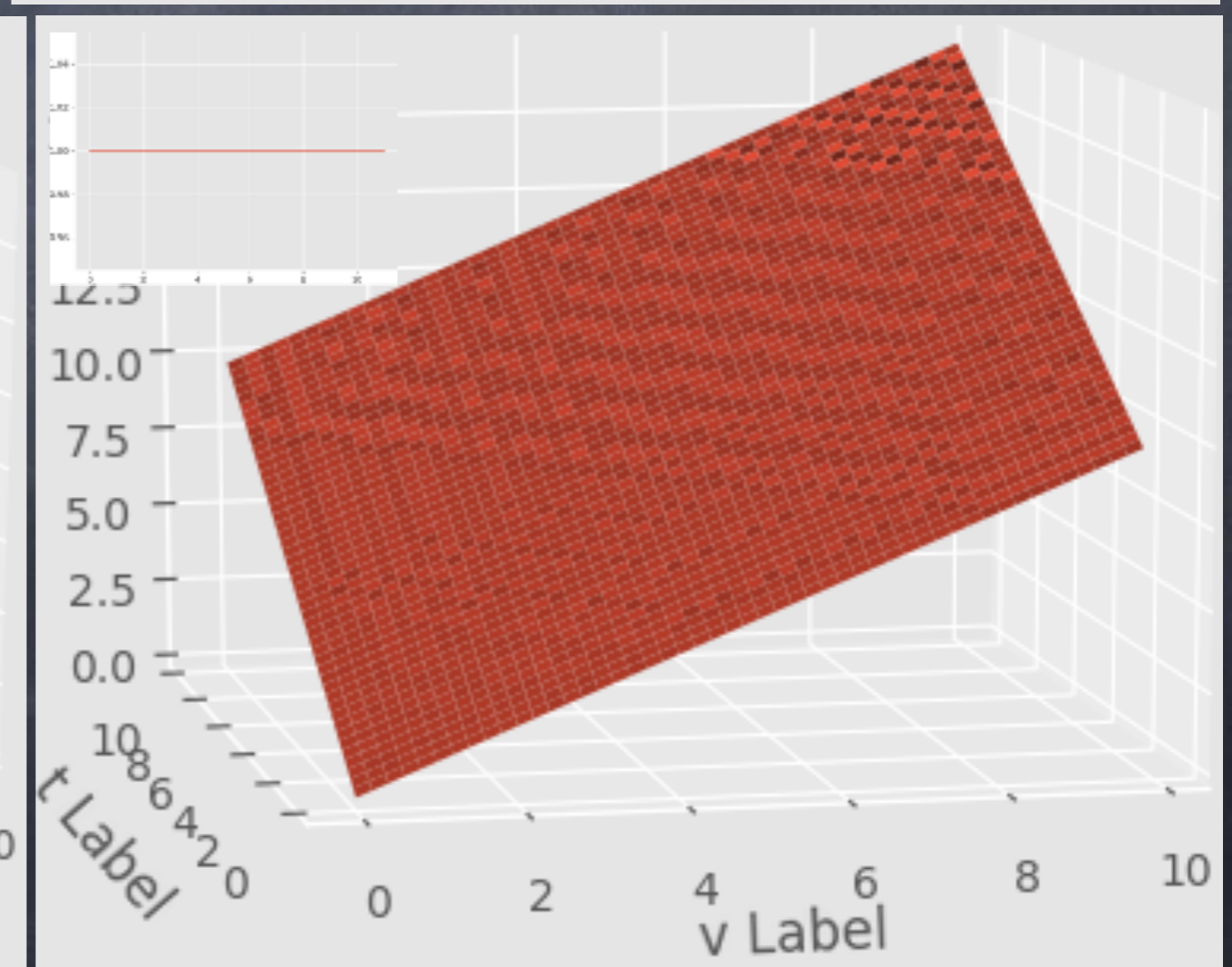
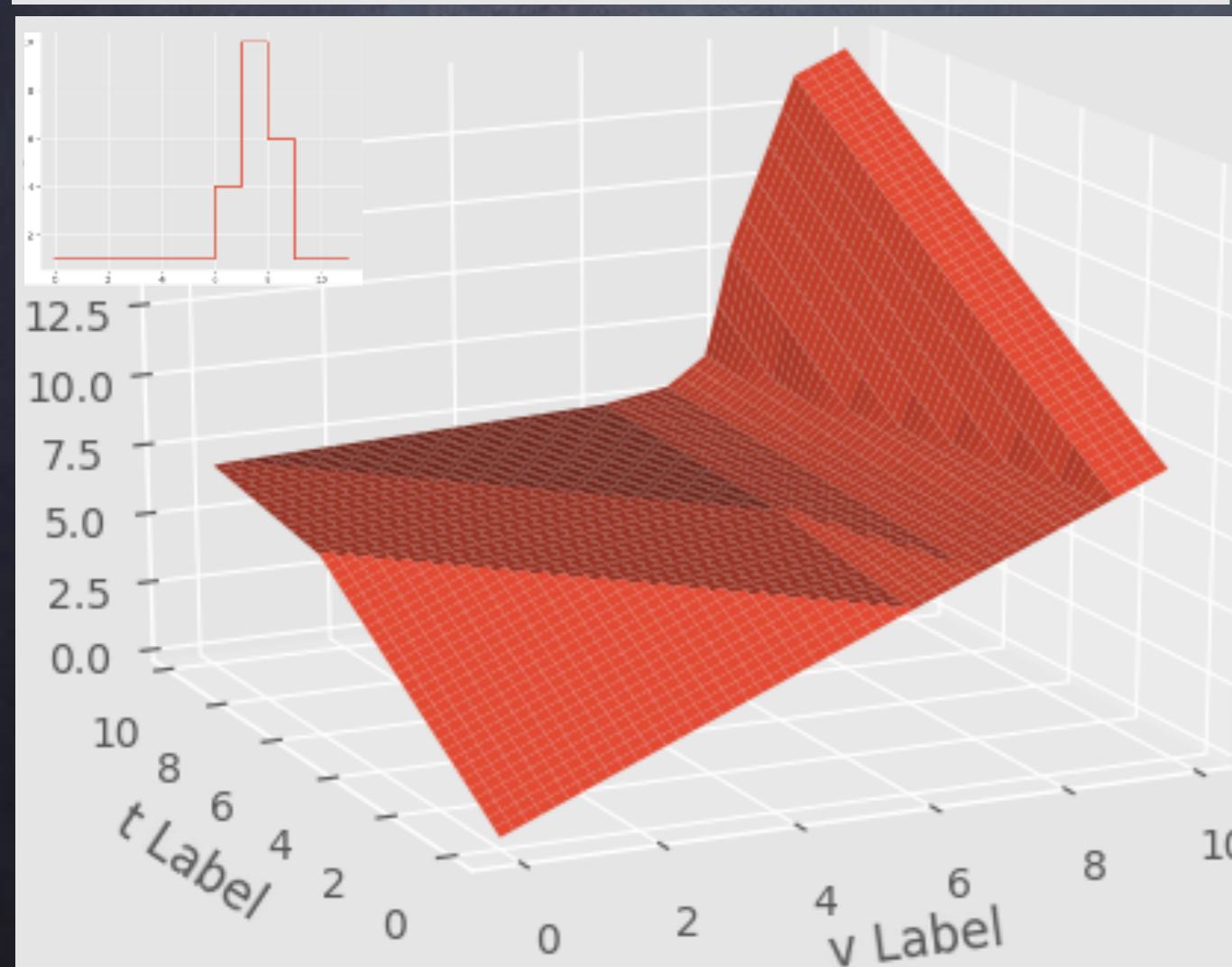
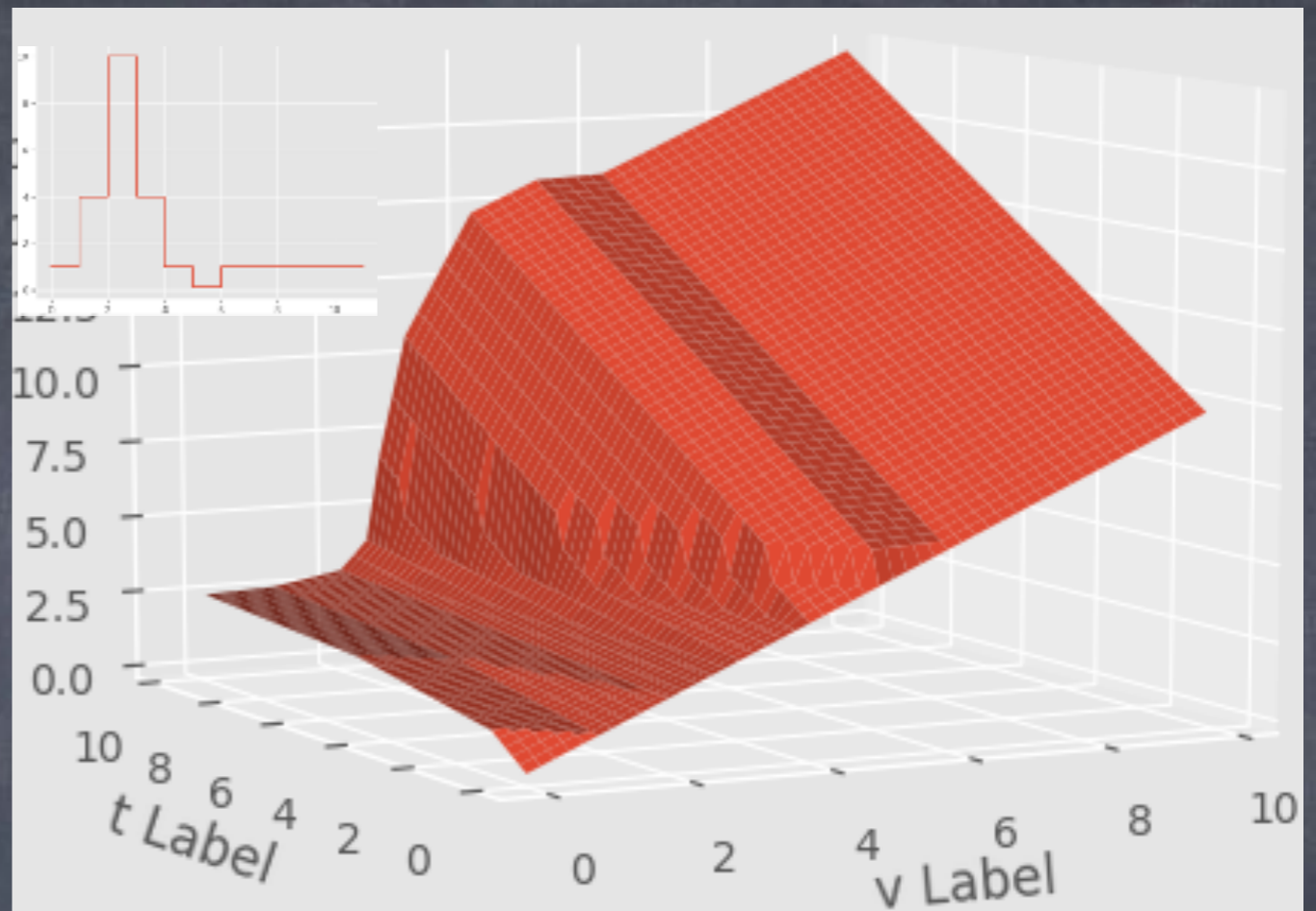
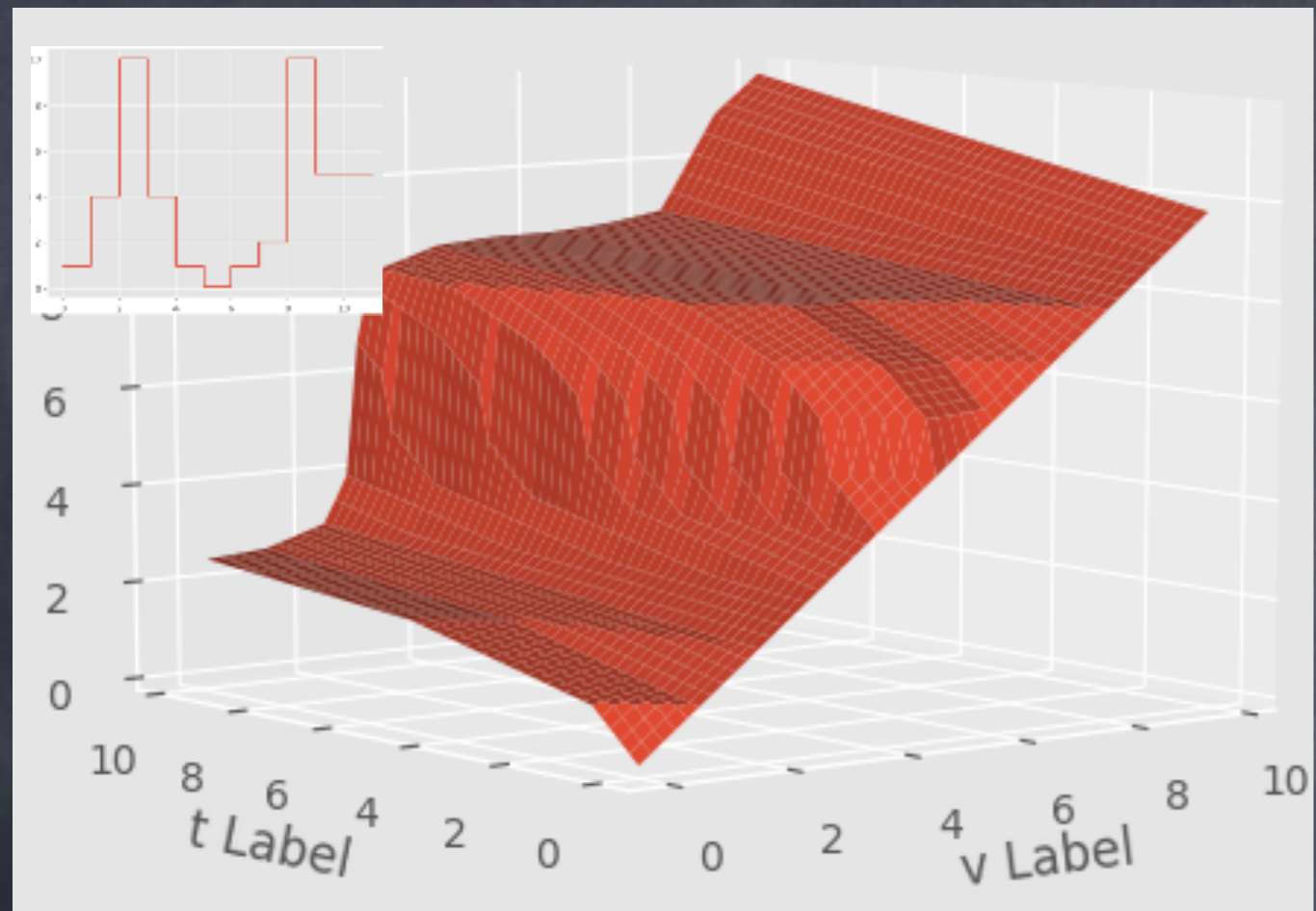
$n=1$ Implementing contours

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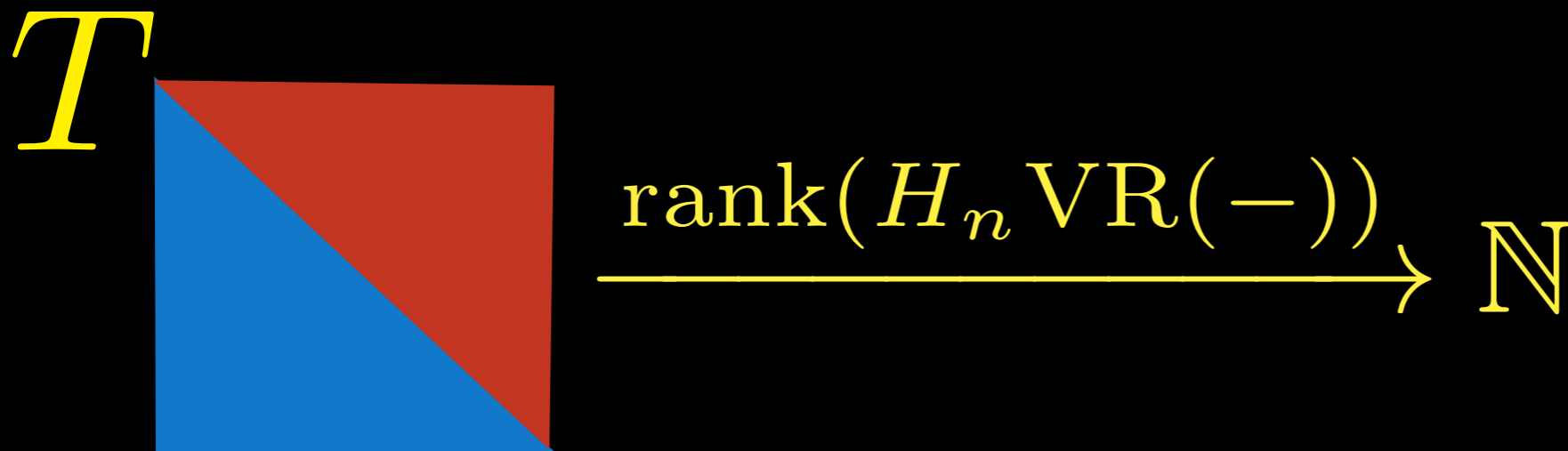
Contour learning

Contour learning

T



Contour learning



Contour learning

Density functions

$$f: \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{> 0}$$

T



$$\xrightarrow{\text{rank}(H_n \text{VR}(-))} \mathbb{N}$$

Contour learning

Density functions

$$f: \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{> 0}$$

T



$$\xrightarrow{\widehat{\text{rank}}(H_n \text{VR}(-))} \text{Funk}(\mathbb{R}_{\geq 0}, \mathbb{R}_{\geq 0})$$

Contour learning

Density functions

$$f: \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{> 0}$$

 ψ $\rightarrow \mathbb{R}$ 

$$\psi(f) = \inf \left\{ \int |\widehat{\text{rank}}(X) - \widehat{\text{rank}}(Y)| \, dX, dY \right\}$$

 T  $\widehat{\text{rank}}(H_n \text{VR}(-))$  $\text{Funk}(\mathbb{R}_{\geq 0}, \mathbb{R}_{\geq 0})$

Contour learning: Maximizing ψ

Density functions

$$f: \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{> 0}$$

ψ

\mathbb{R}



$$\psi(f) = \inf \left\{ \int |\widehat{\text{rank}}(X) - \widehat{\text{rank}}(Y)| \, dX, dY \right\}$$

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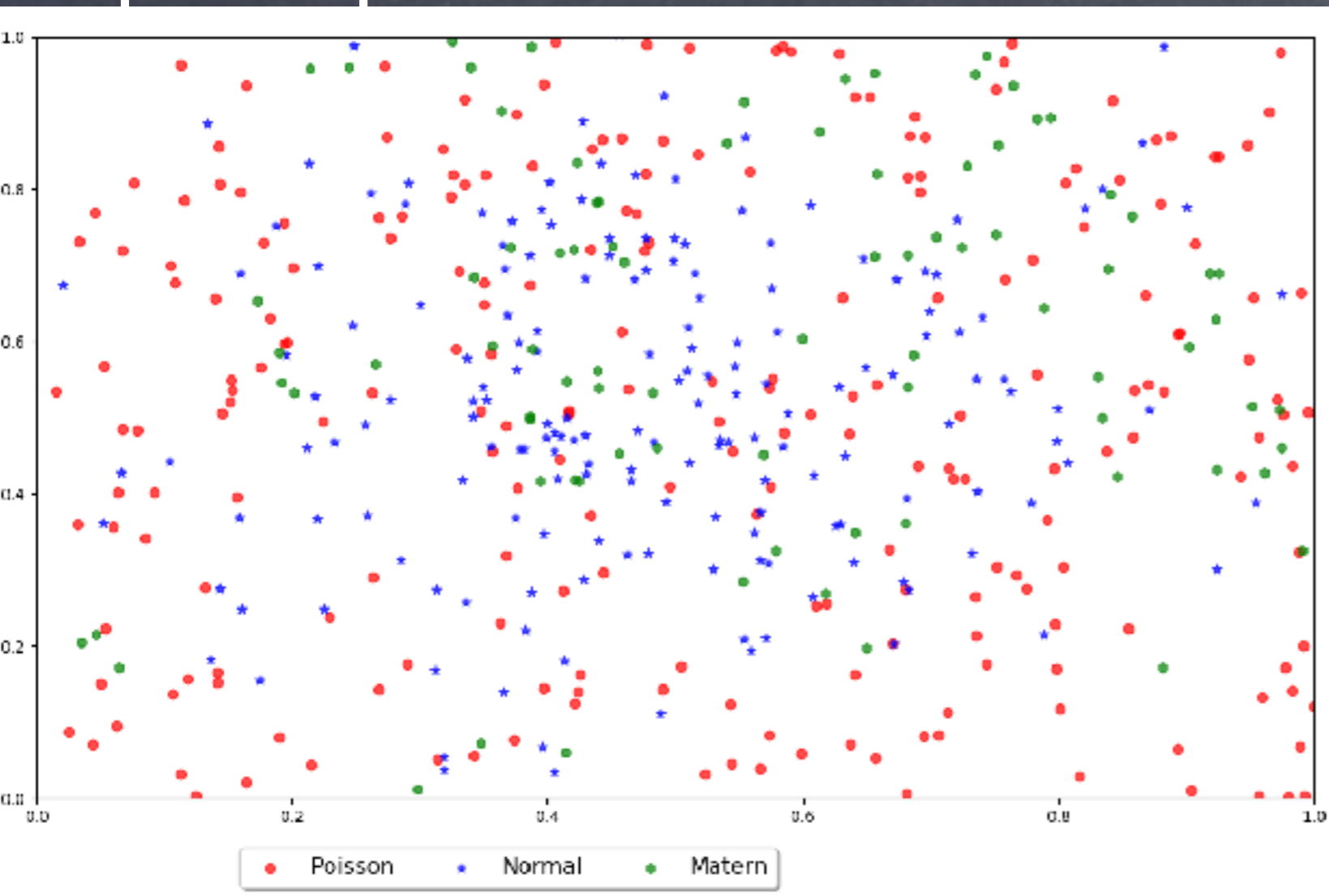
$\widehat{\text{rank}}(H_n \text{VR}(-))$



$\text{Funk}(\mathbb{R}_{\geq 0}, \mathbb{R}_{\geq 0})$

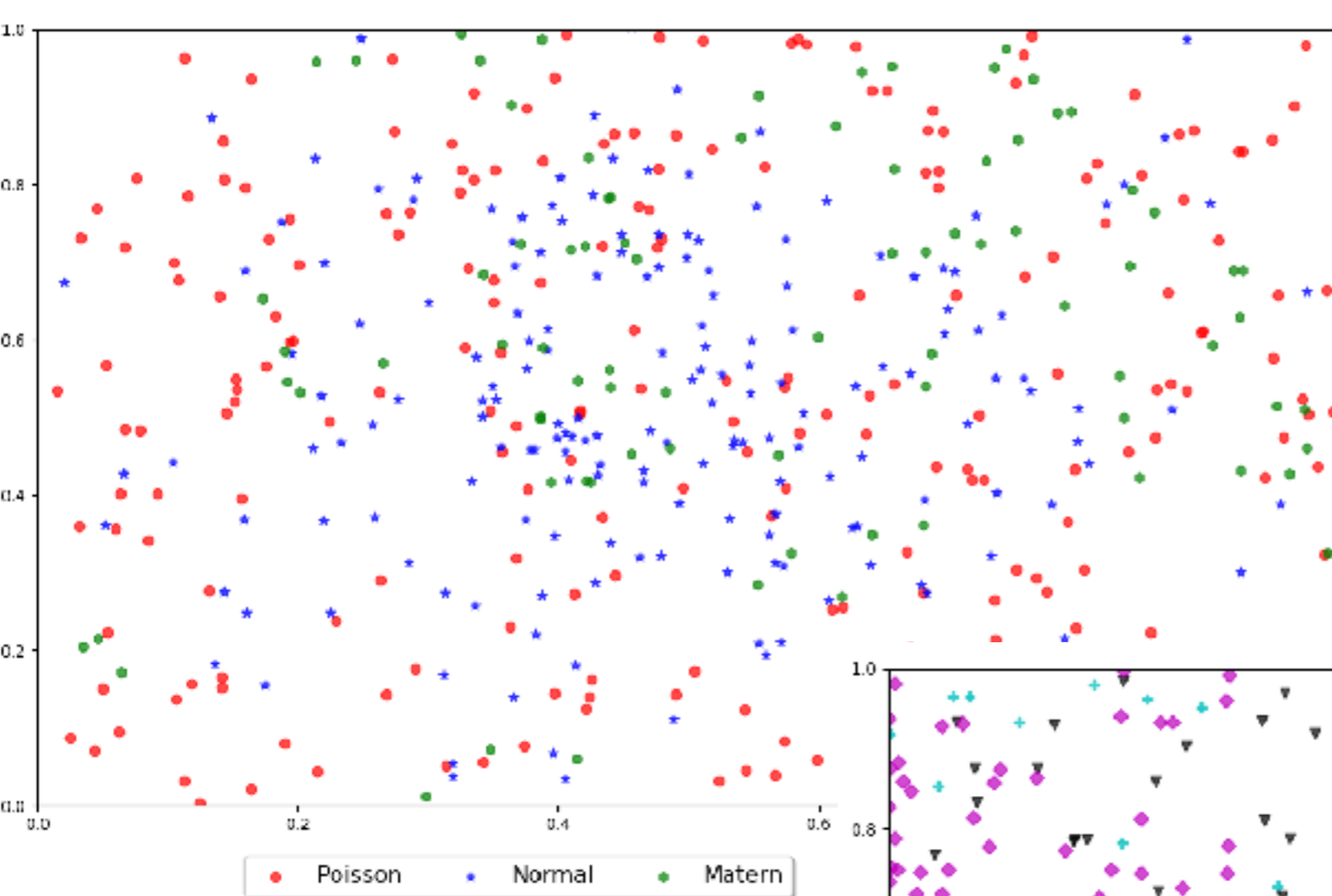
6 point processes on the unit square, containing 200
points on average

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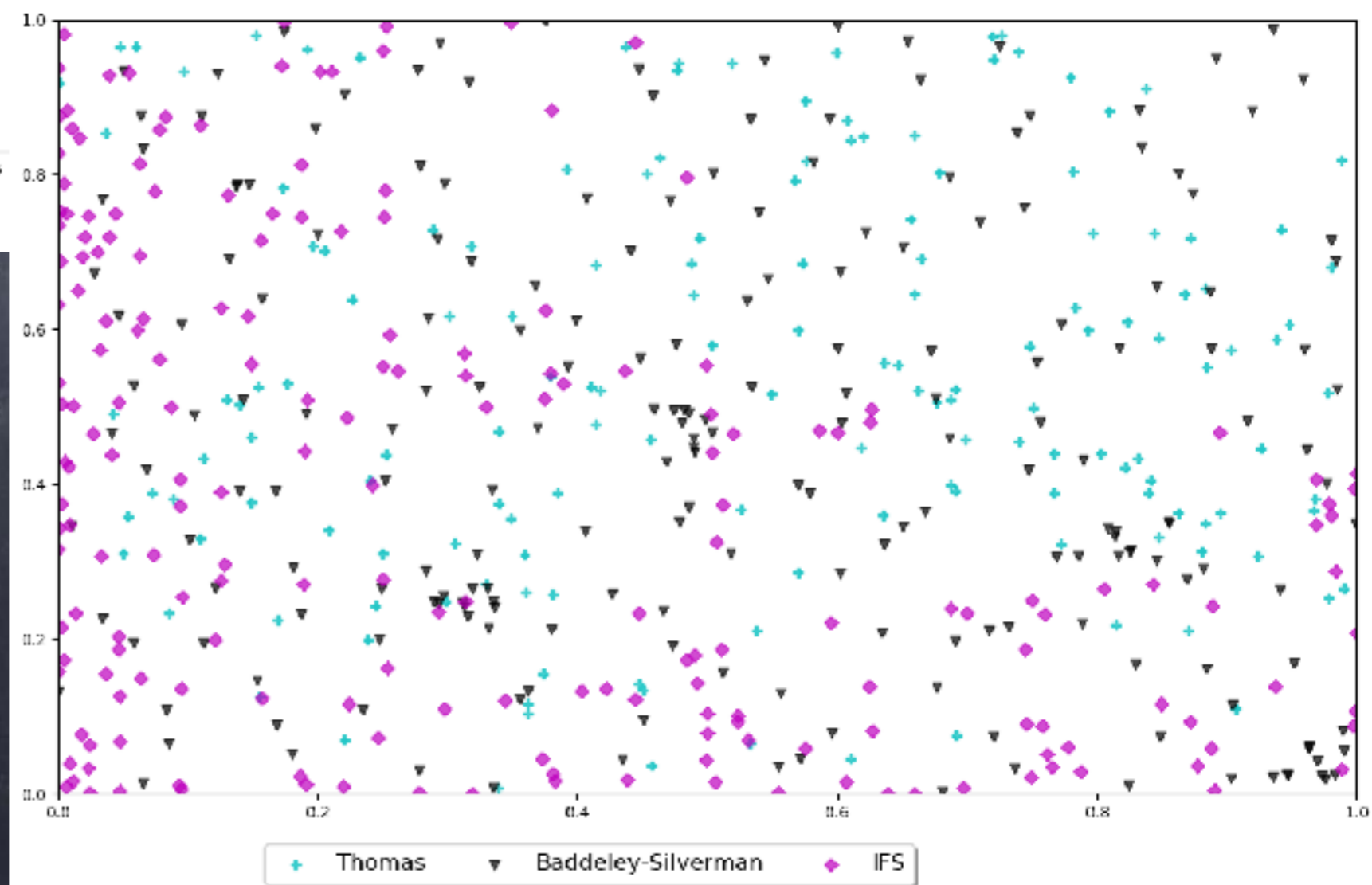


points on average

6 point processes on the unit square, containing 200 points on average



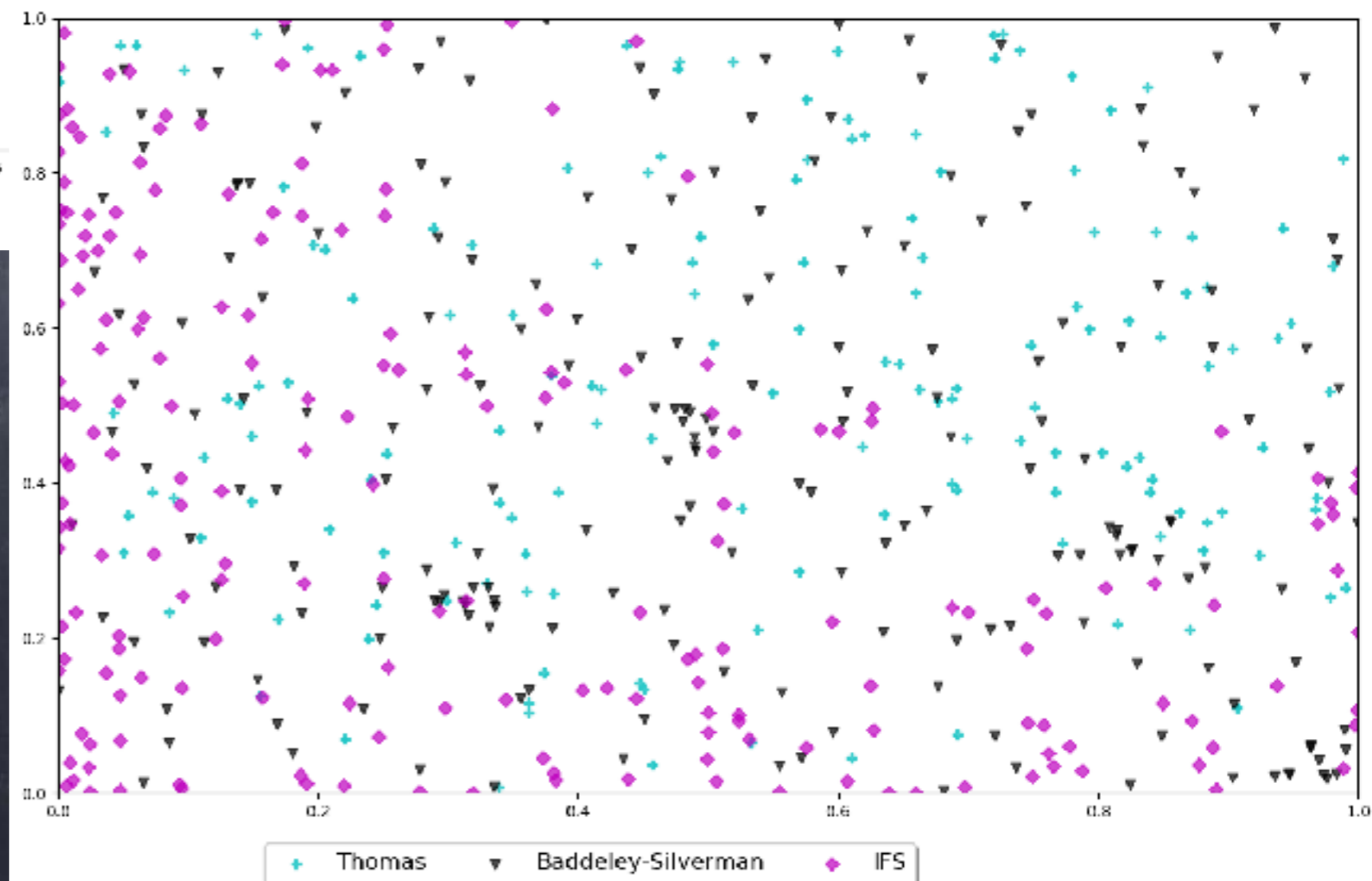
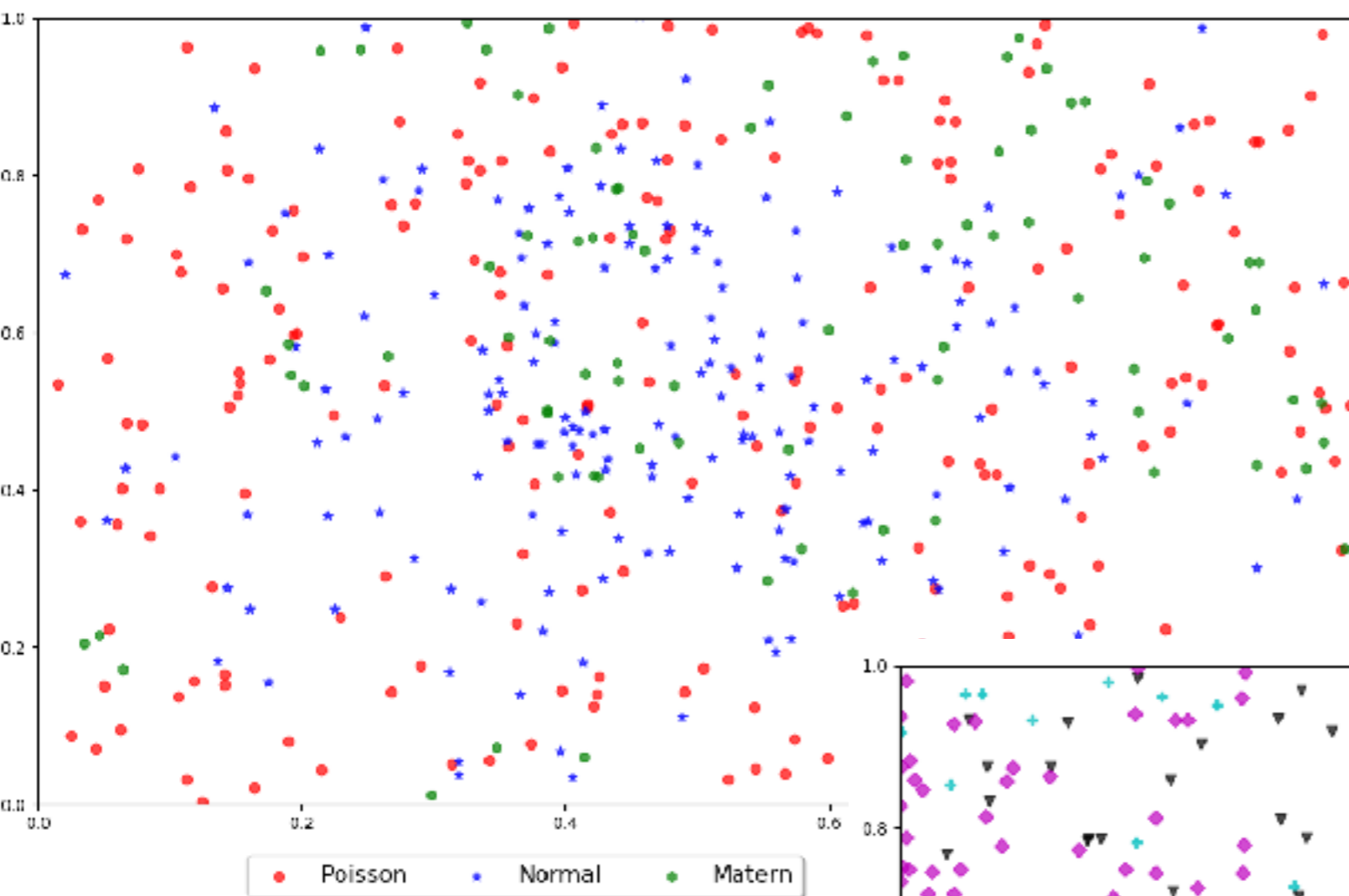
points on average



6 point processes on the unit square, containing 200 points on average

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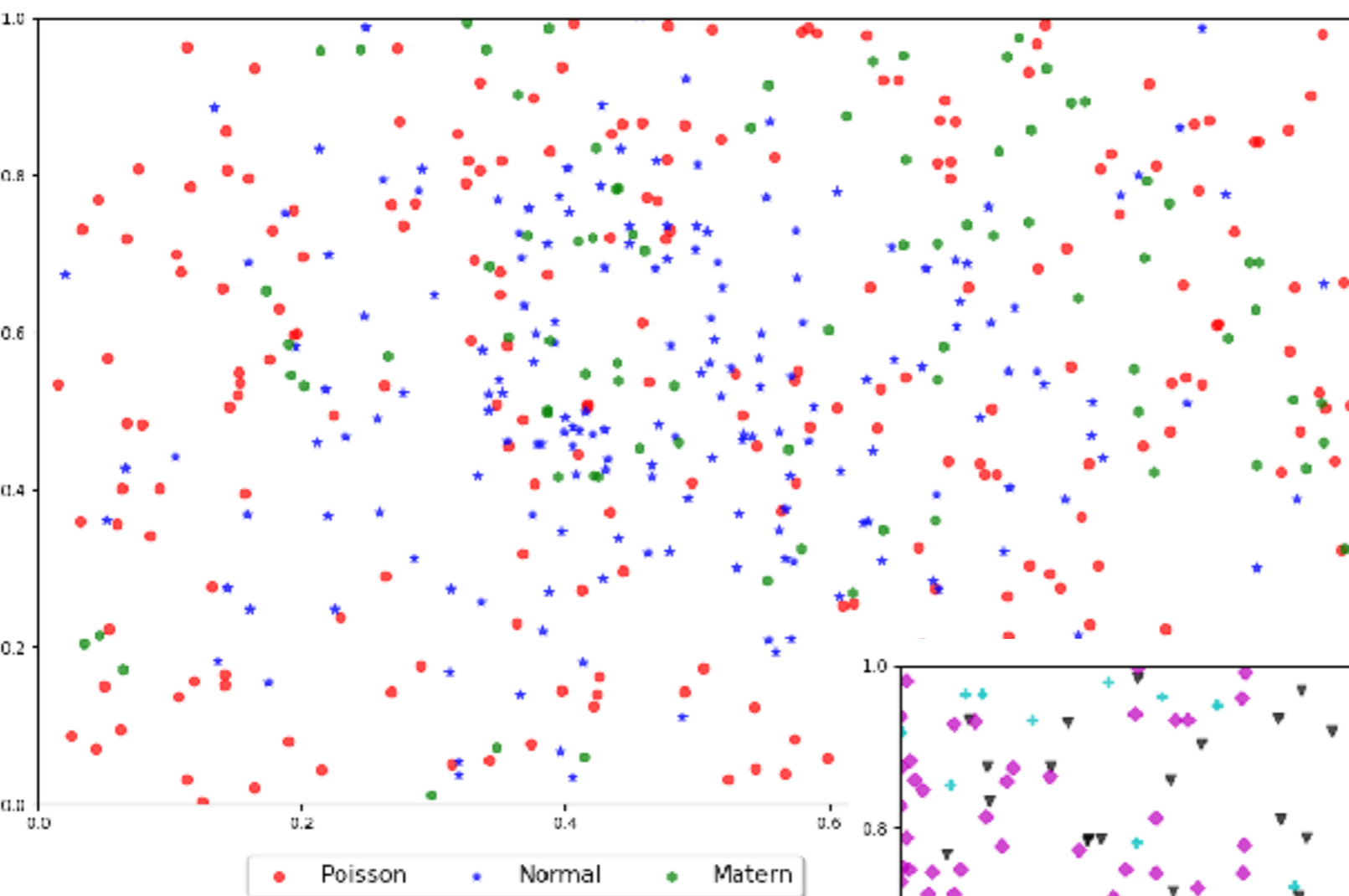
500 instances of each process was generated



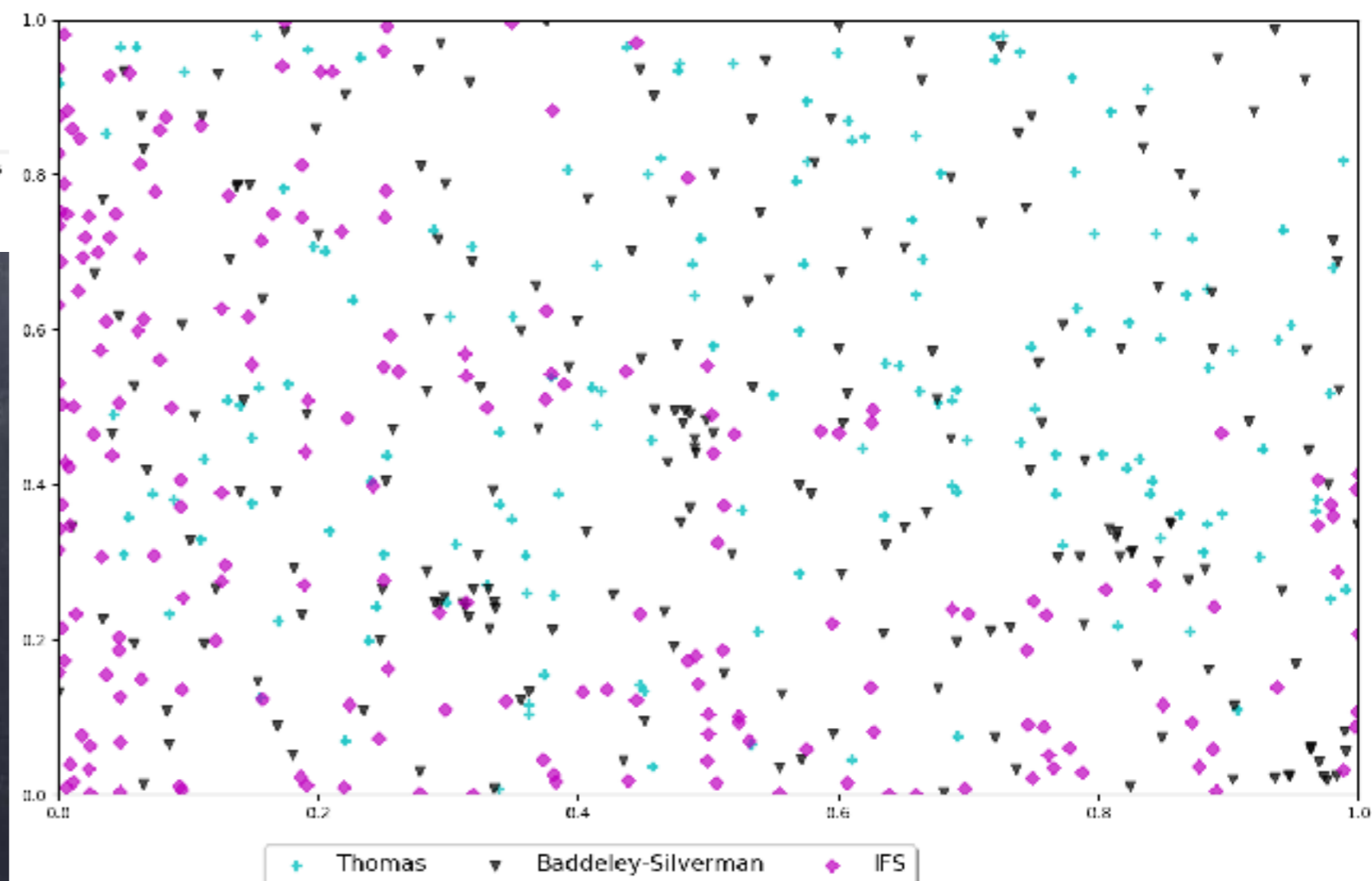
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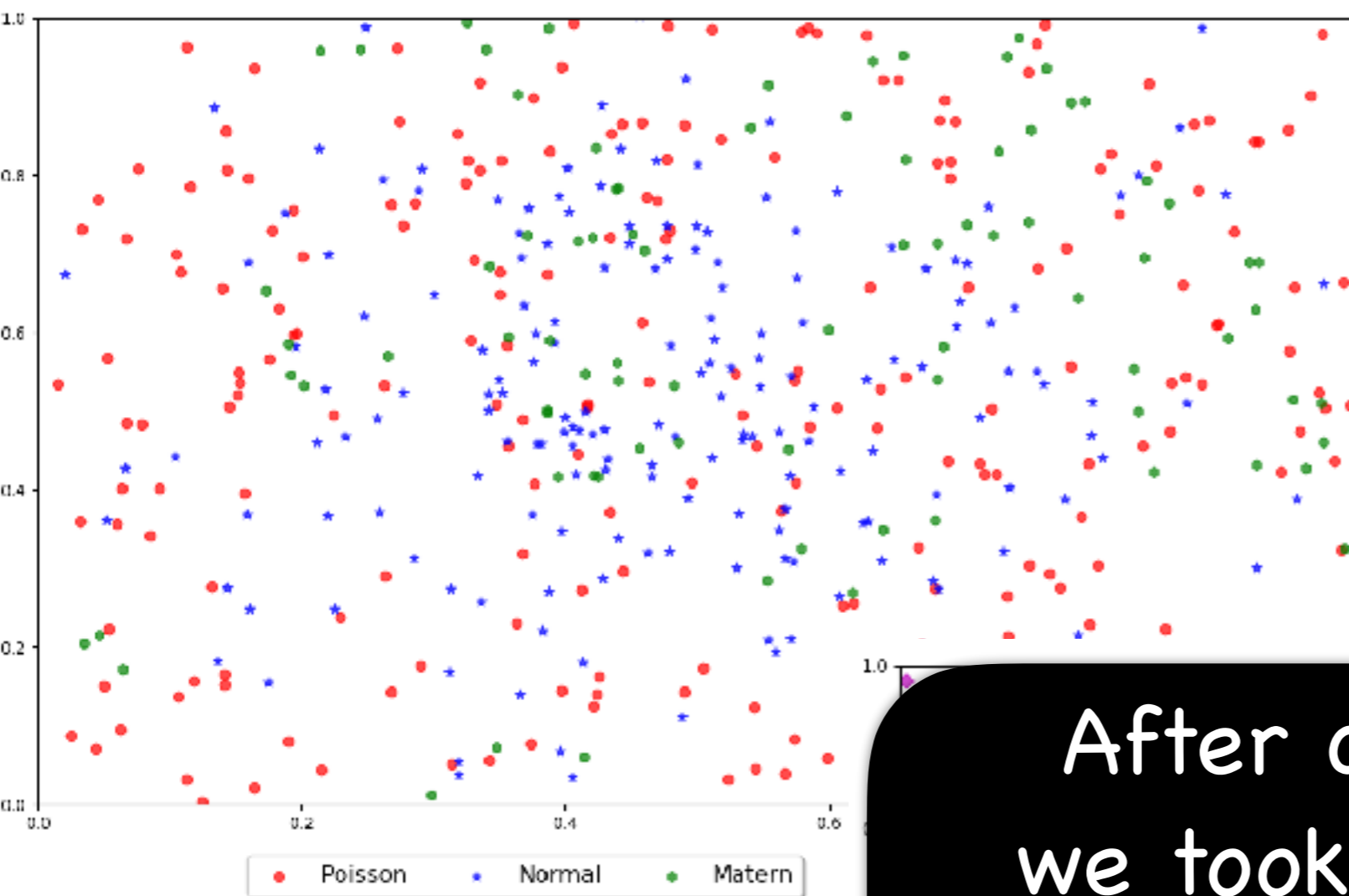
200 used for training
300 used for testing



6 point processes on the unit square, containing 200

points on average

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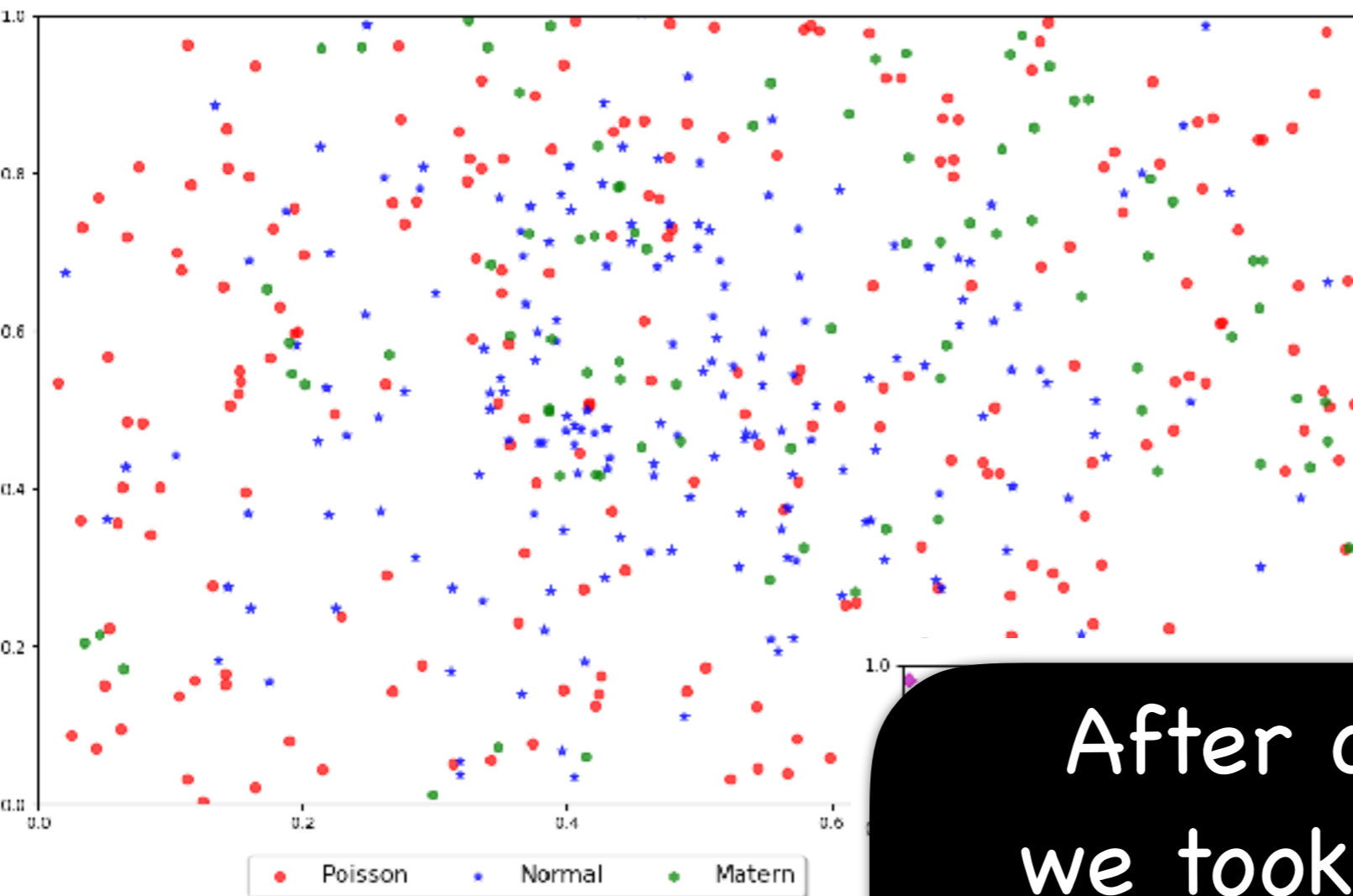
200 used for
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After choosing a contour
we took averages of stable
ranks of the 200 processes in
the training set

6 point processes on the unit square, containing 200

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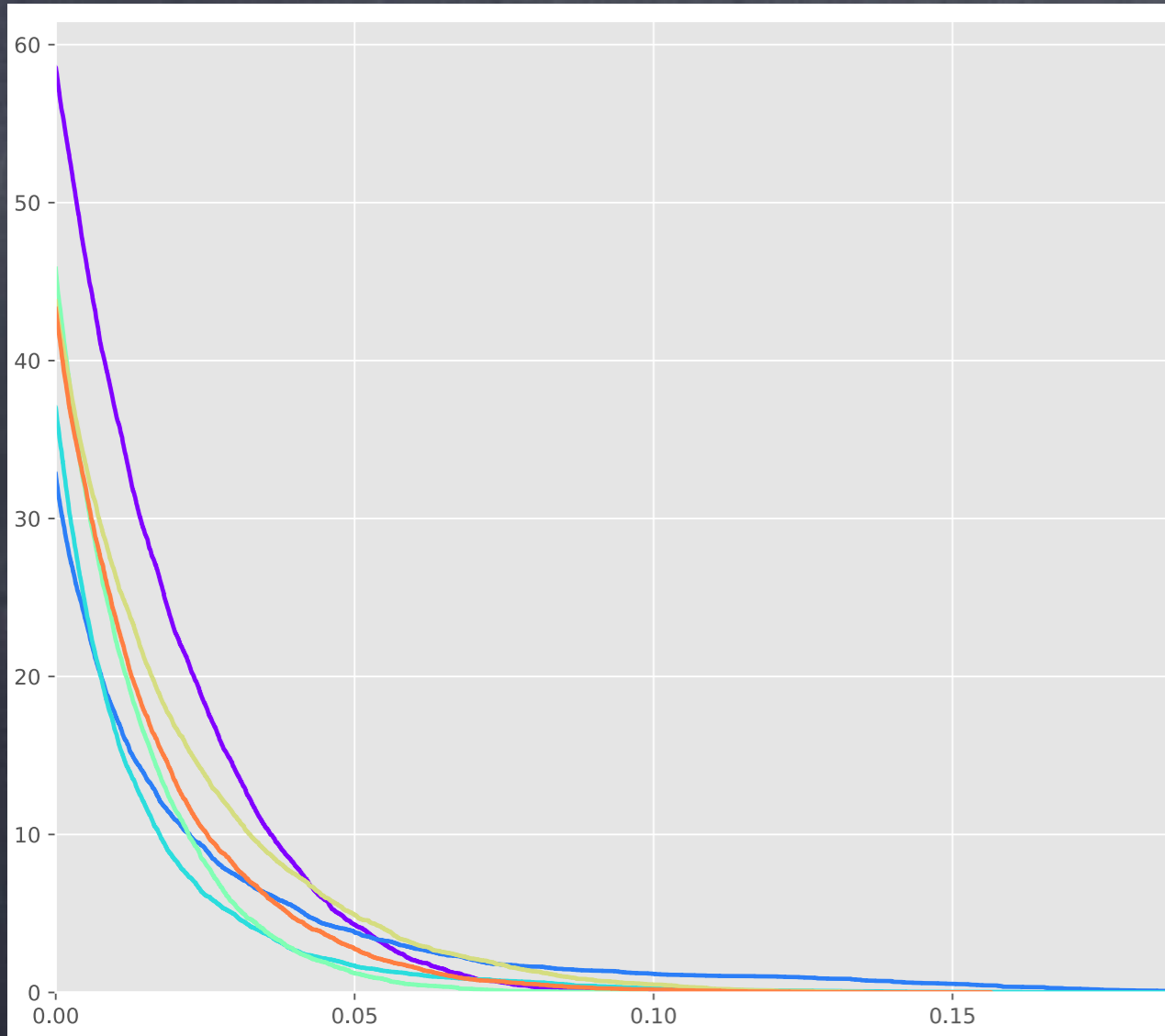
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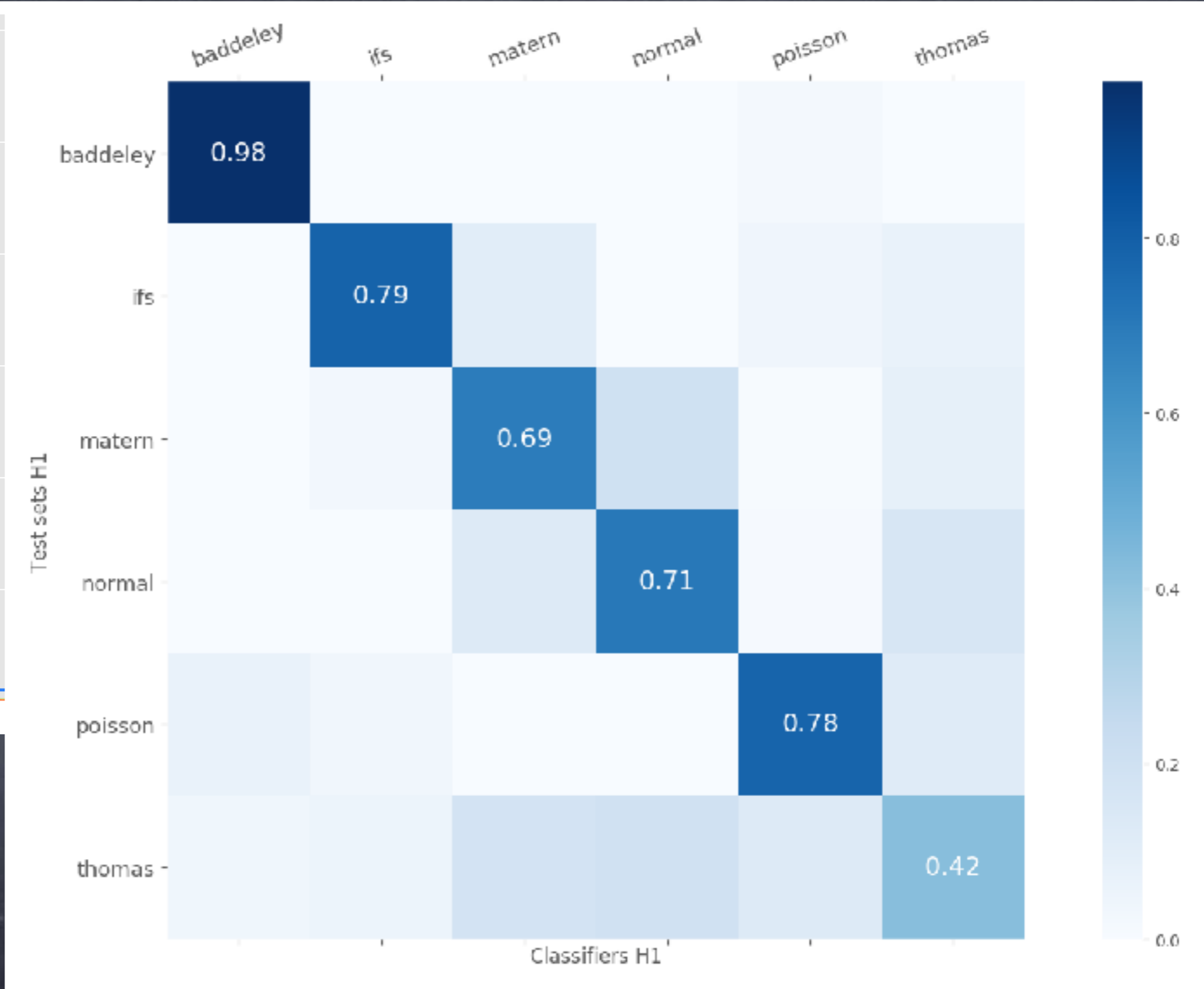
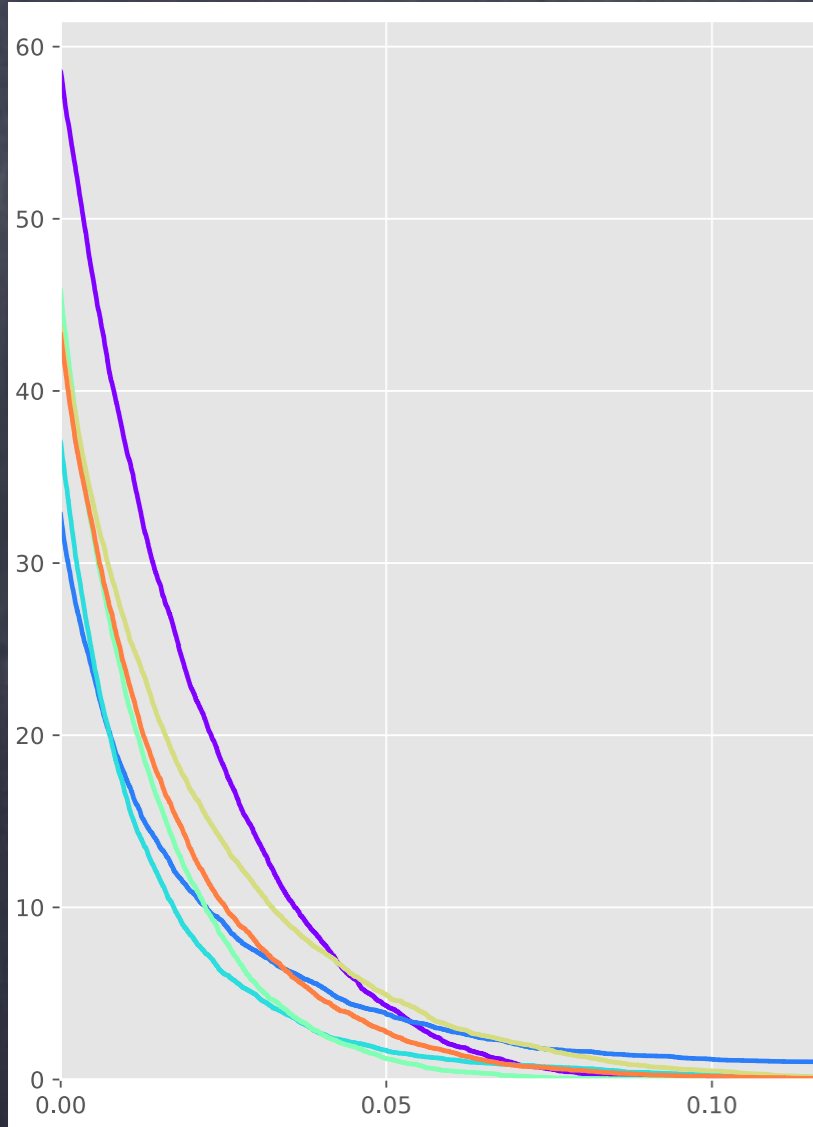
These averages are then used
as classifiers

Standard contour

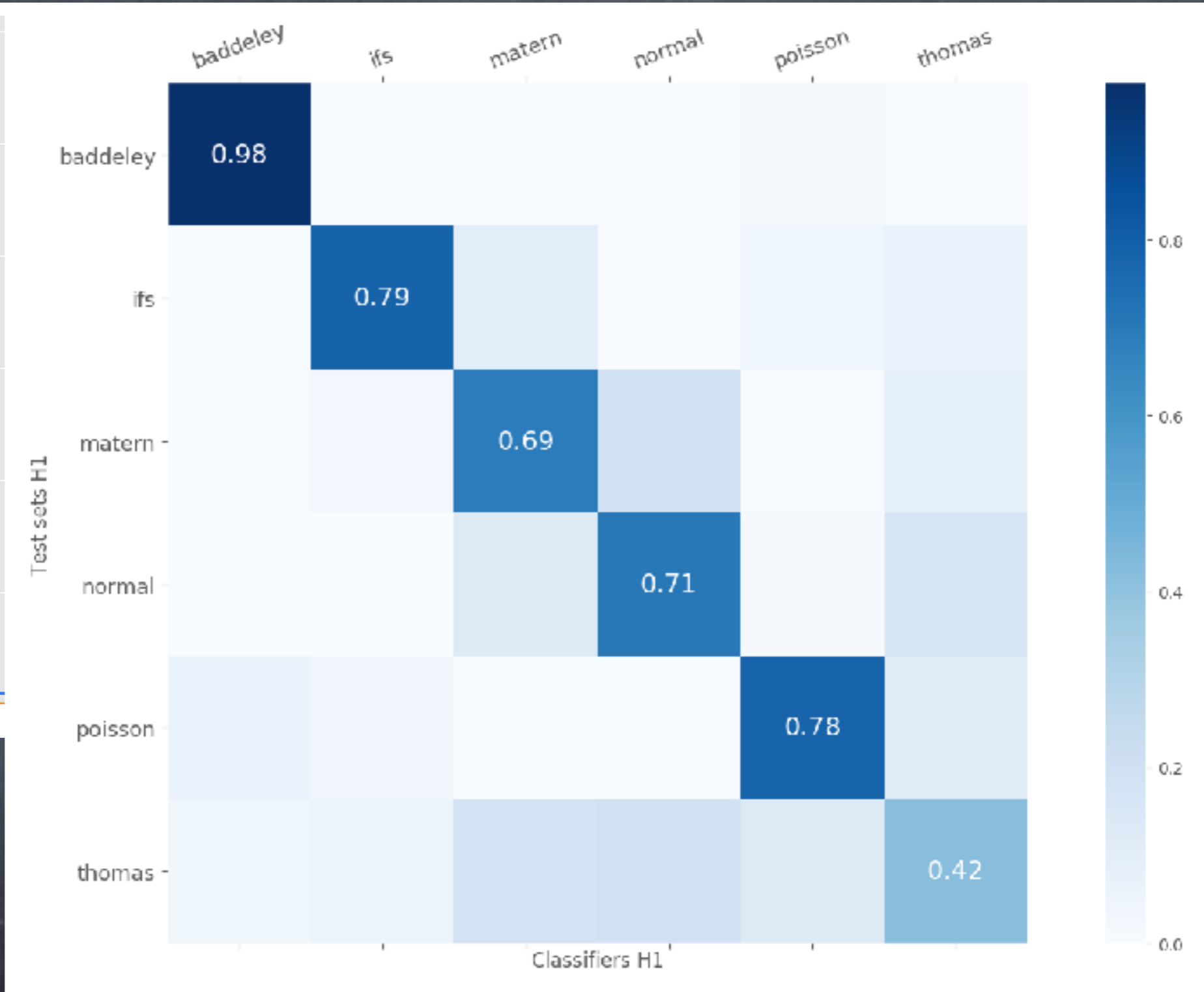
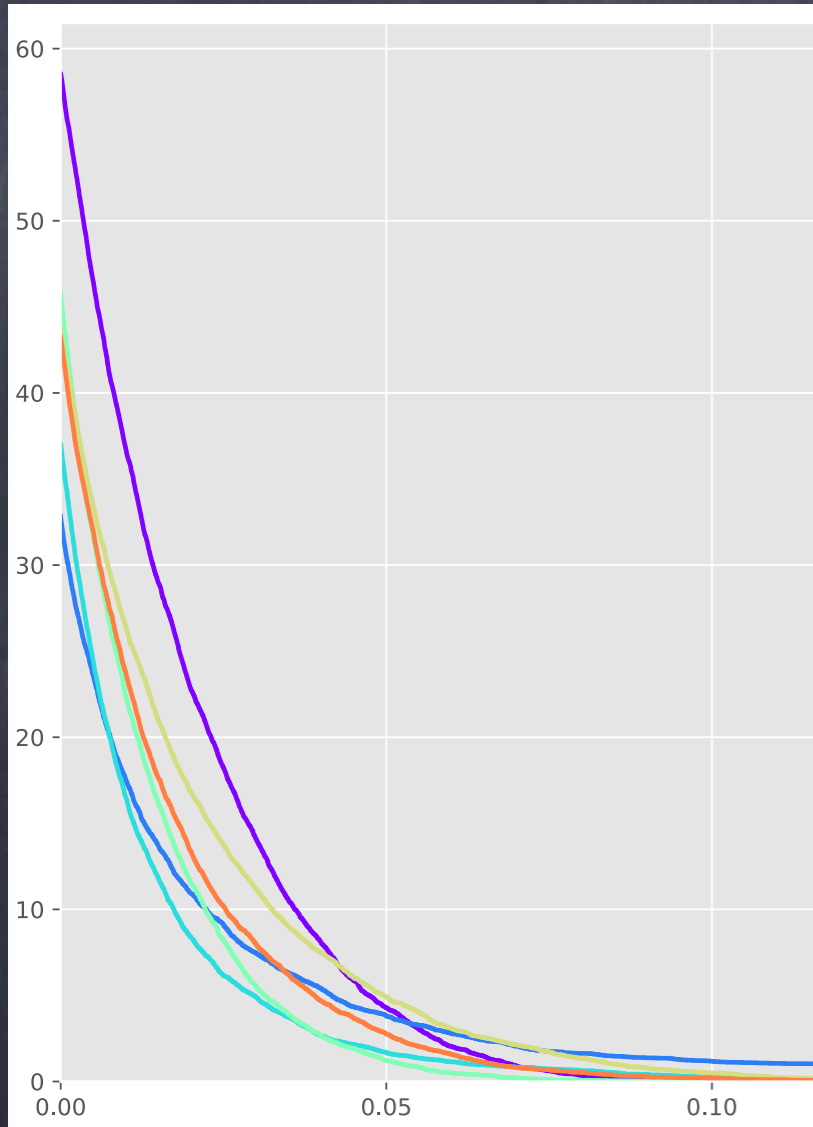
Standard contur



Standard contour

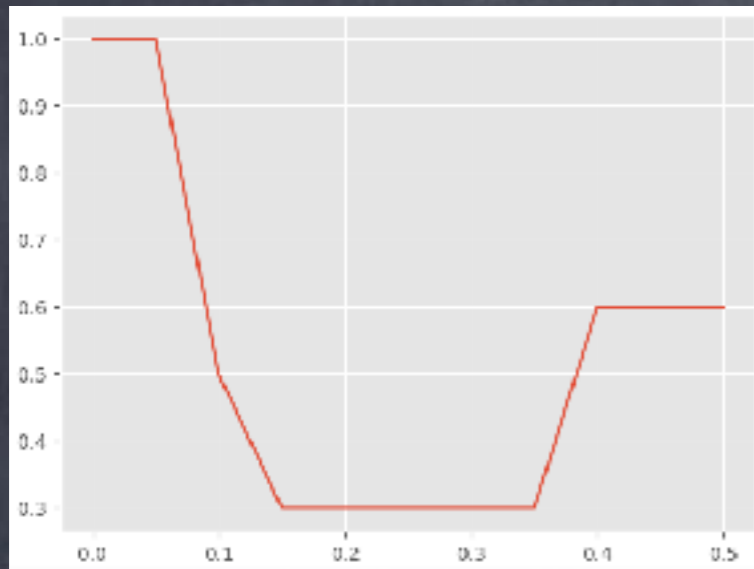


Standard contour

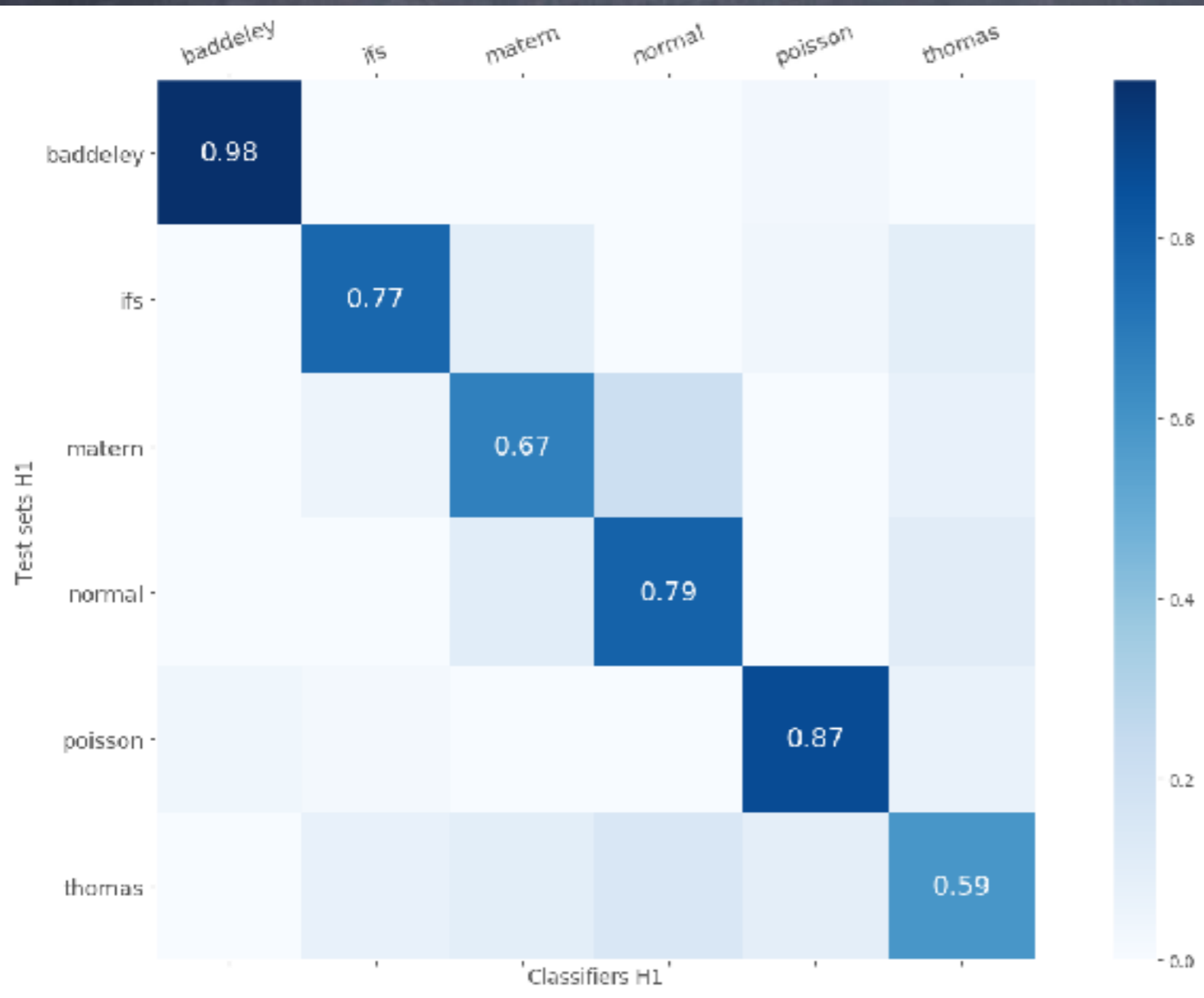
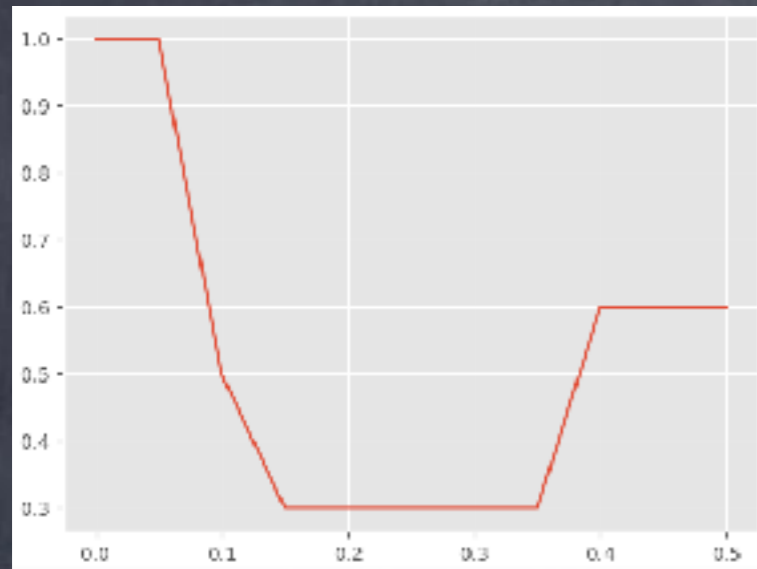


Averages after 20 X cross validation

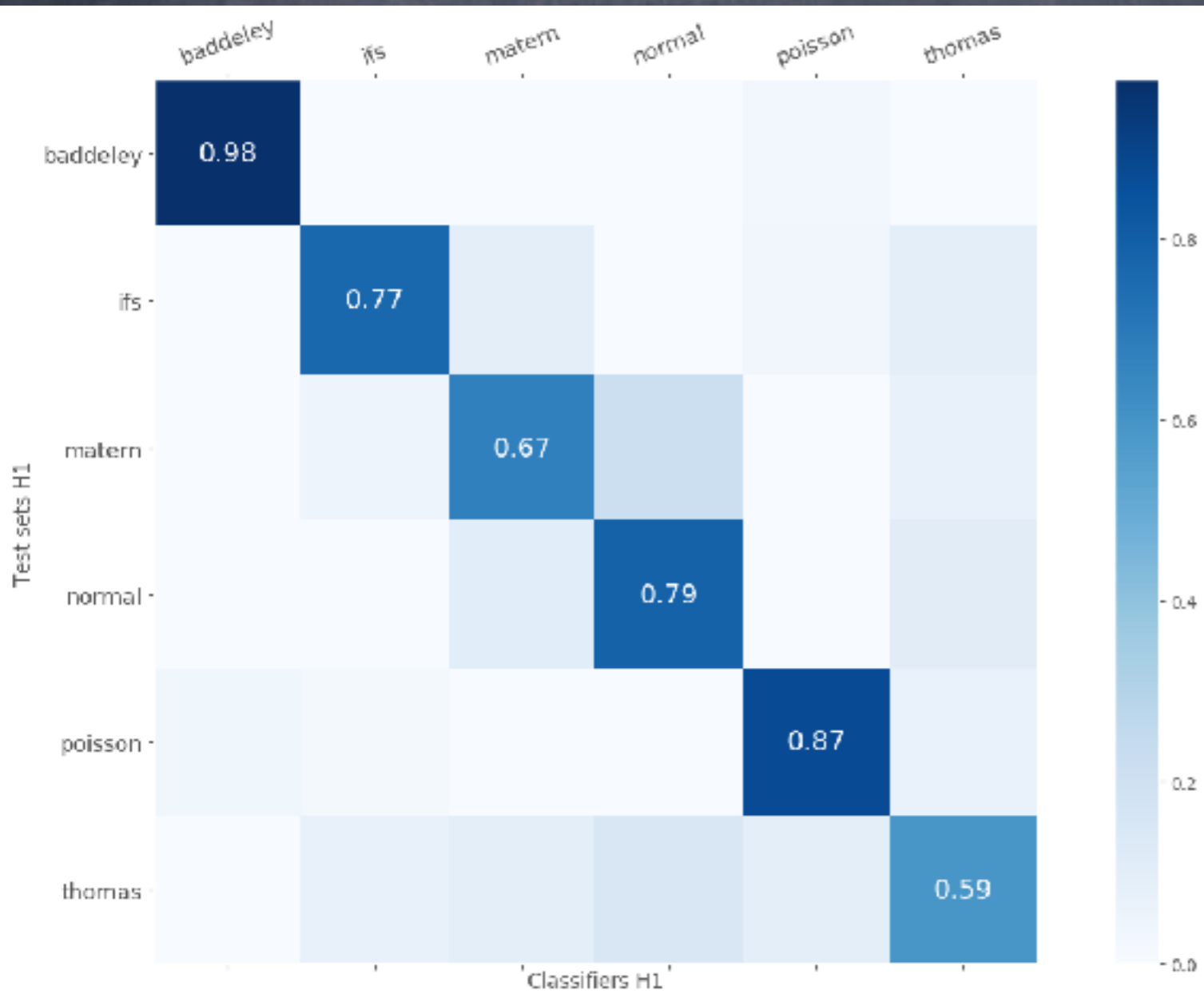
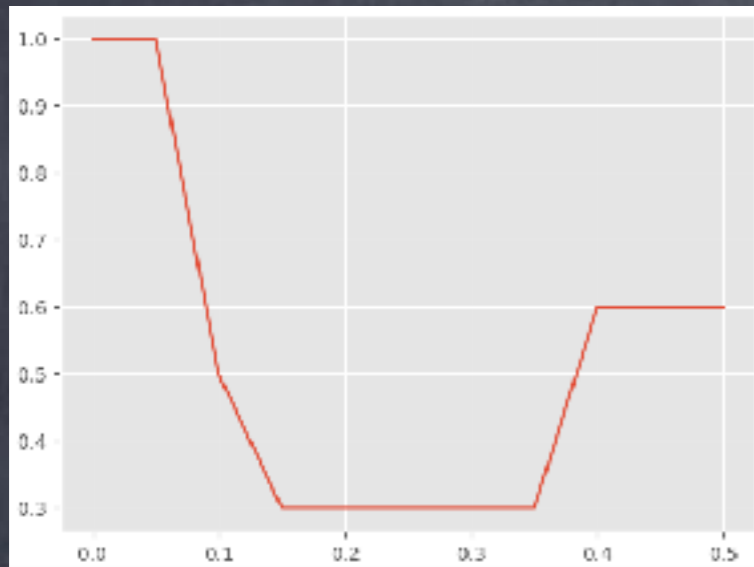
Density:



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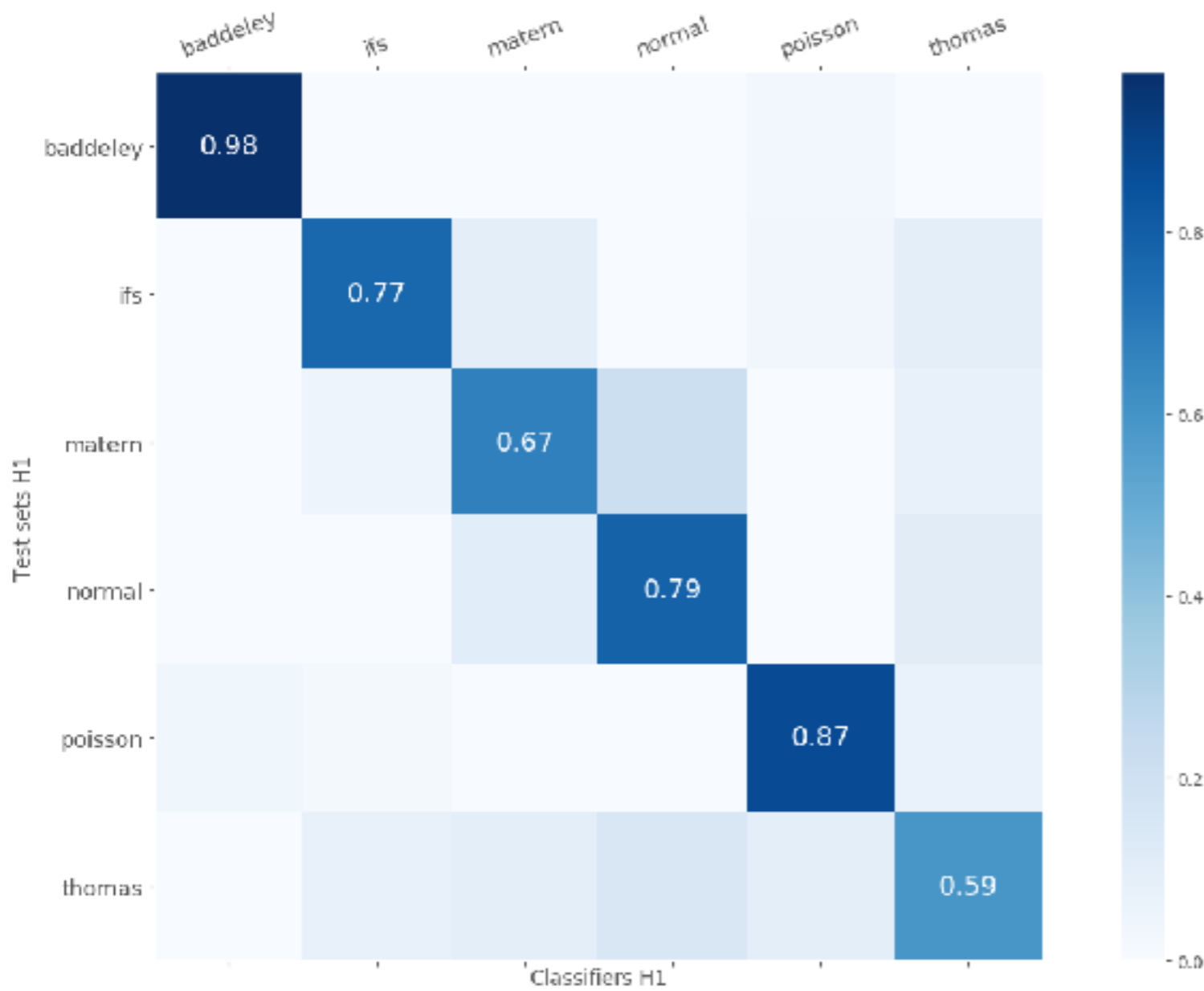
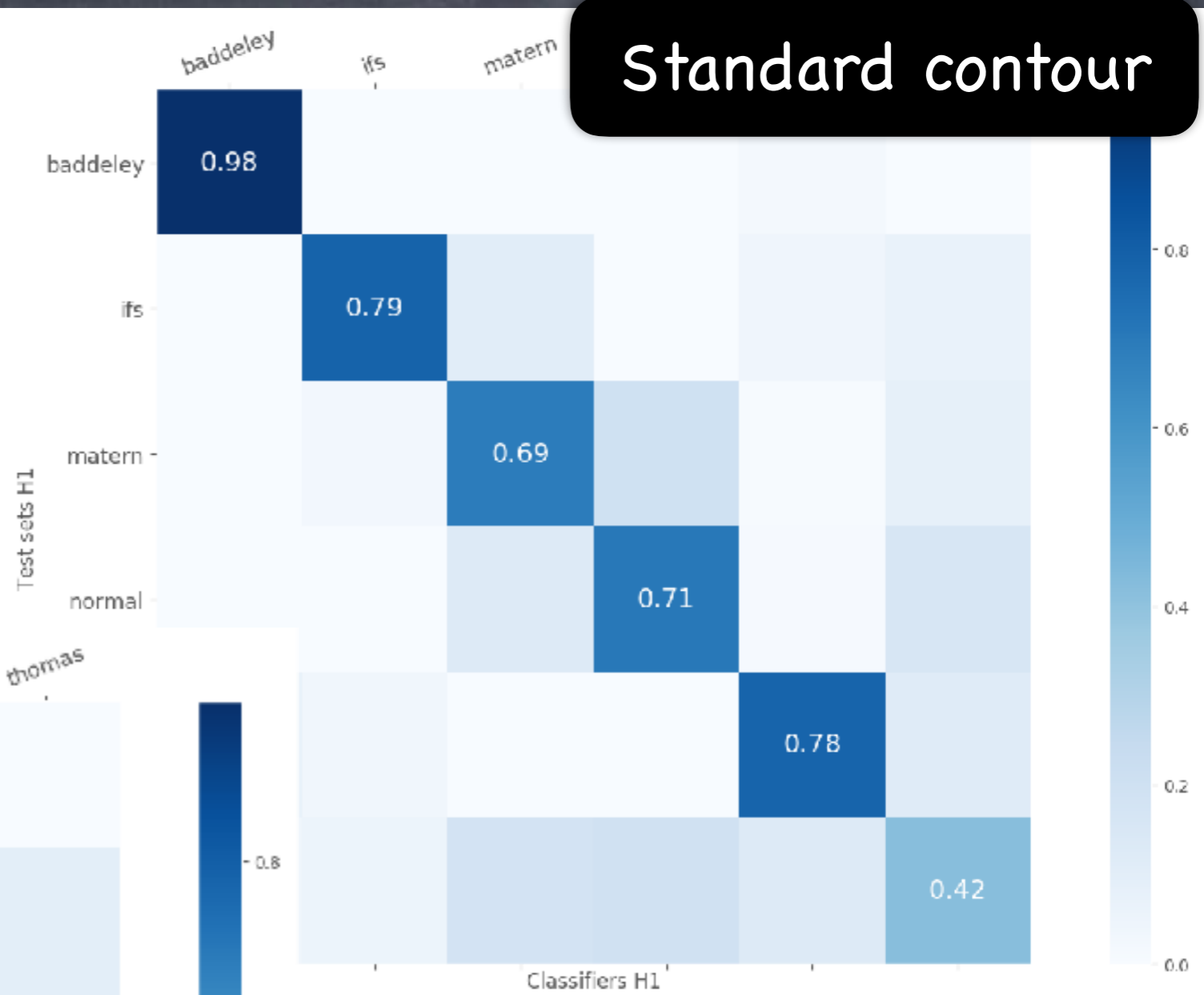
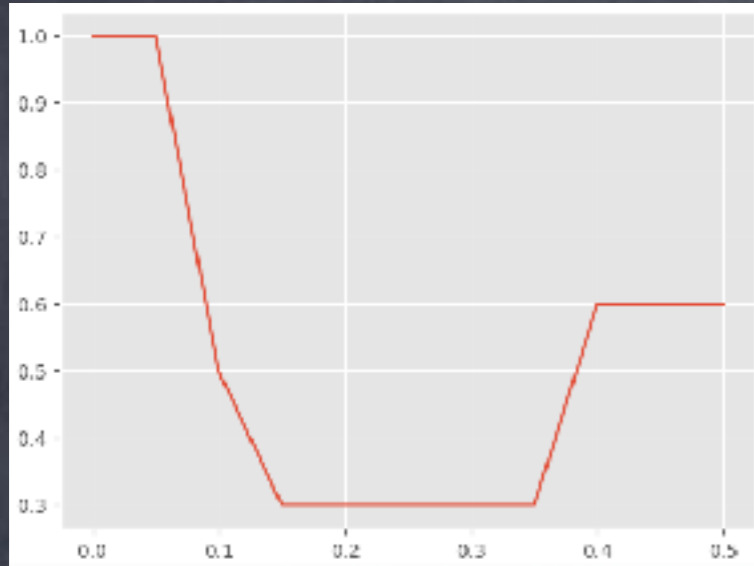


Density:



Averages
after 20 X cross
validation

Density:

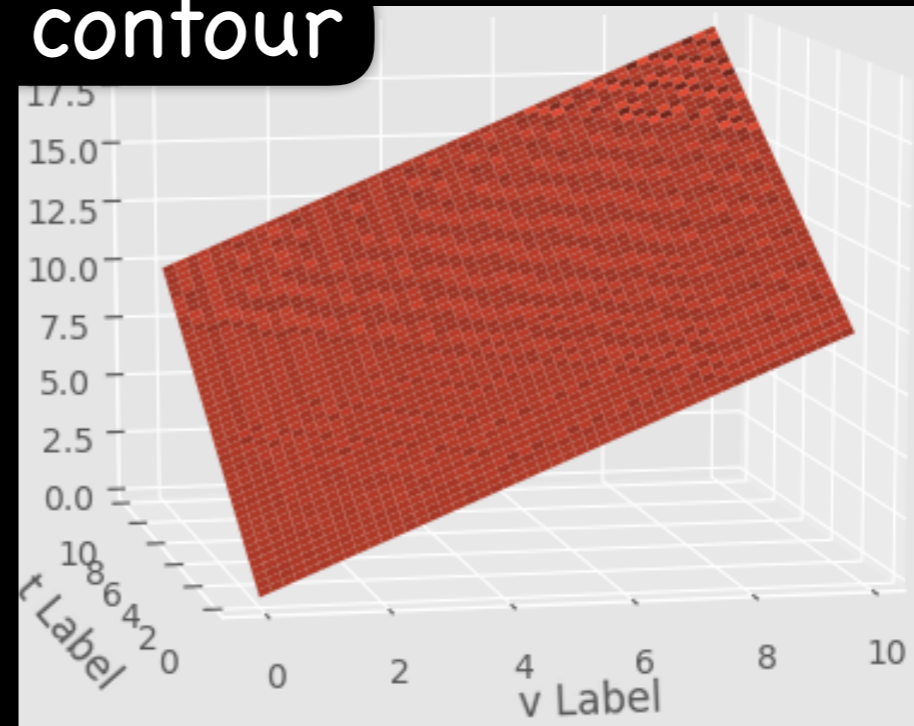
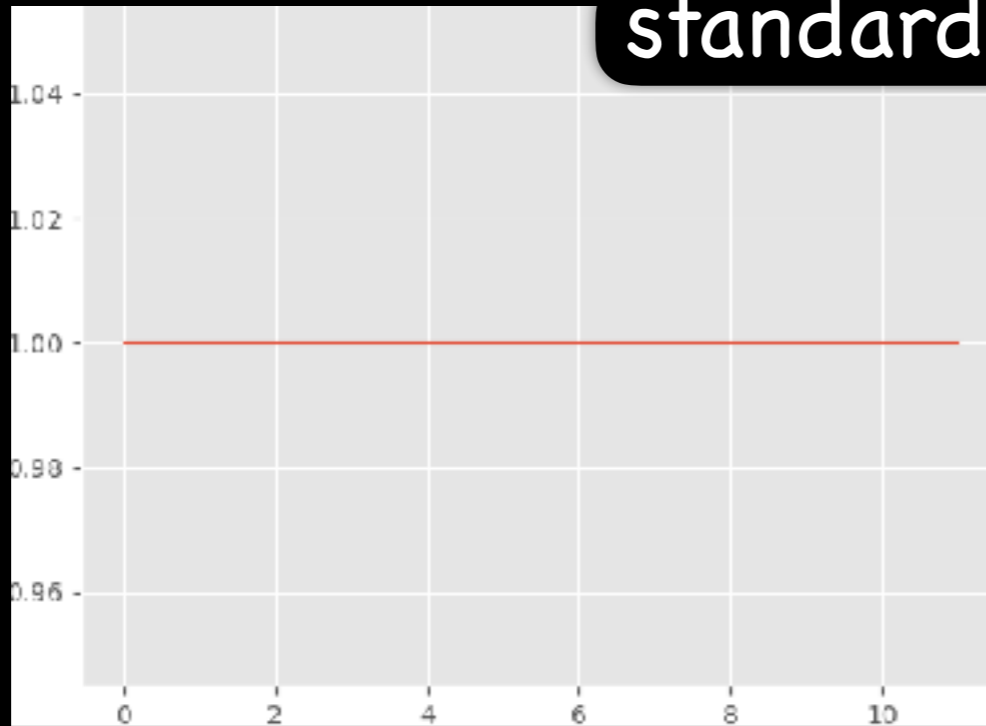


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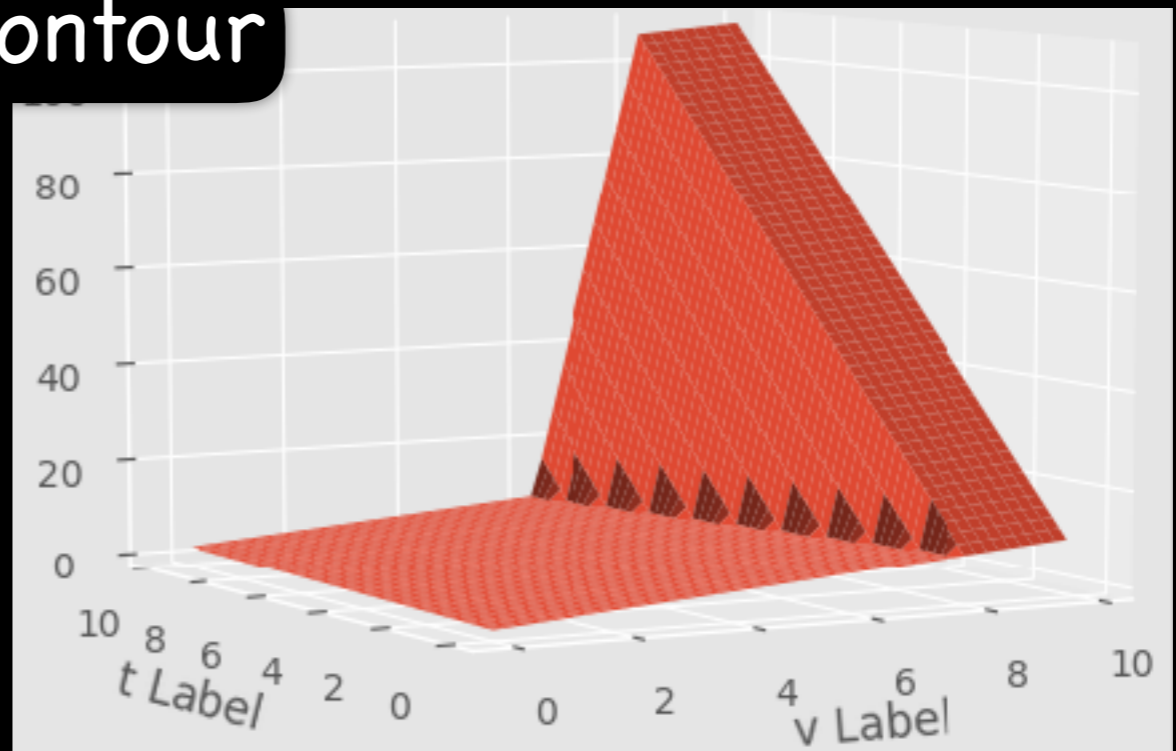
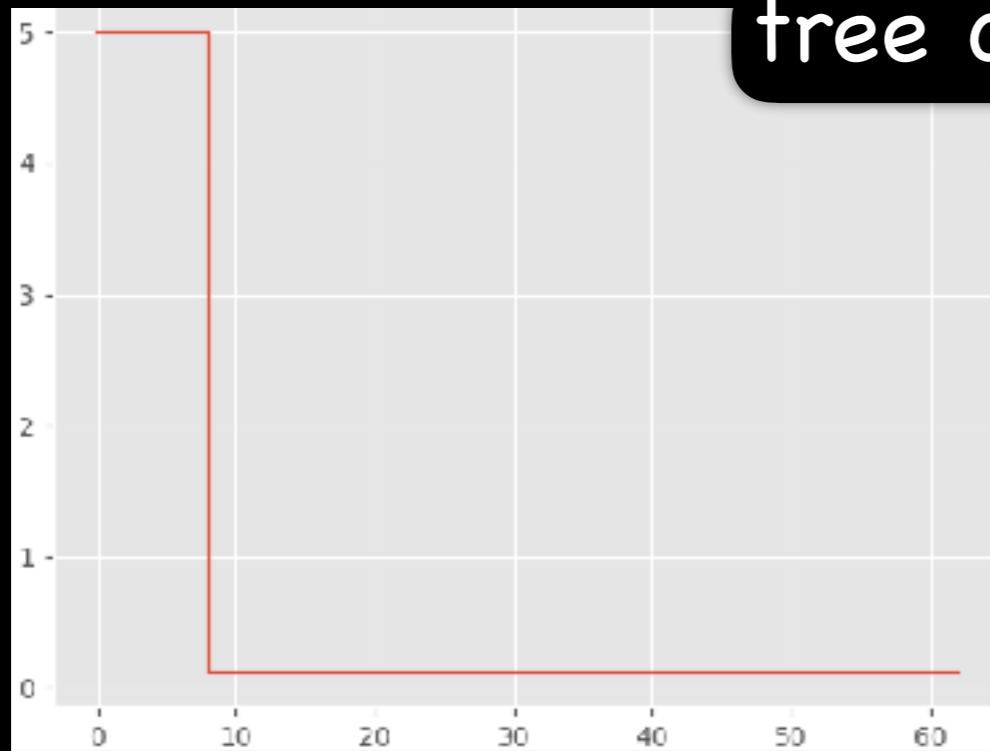
Henri Riihimäki, forest tree data:

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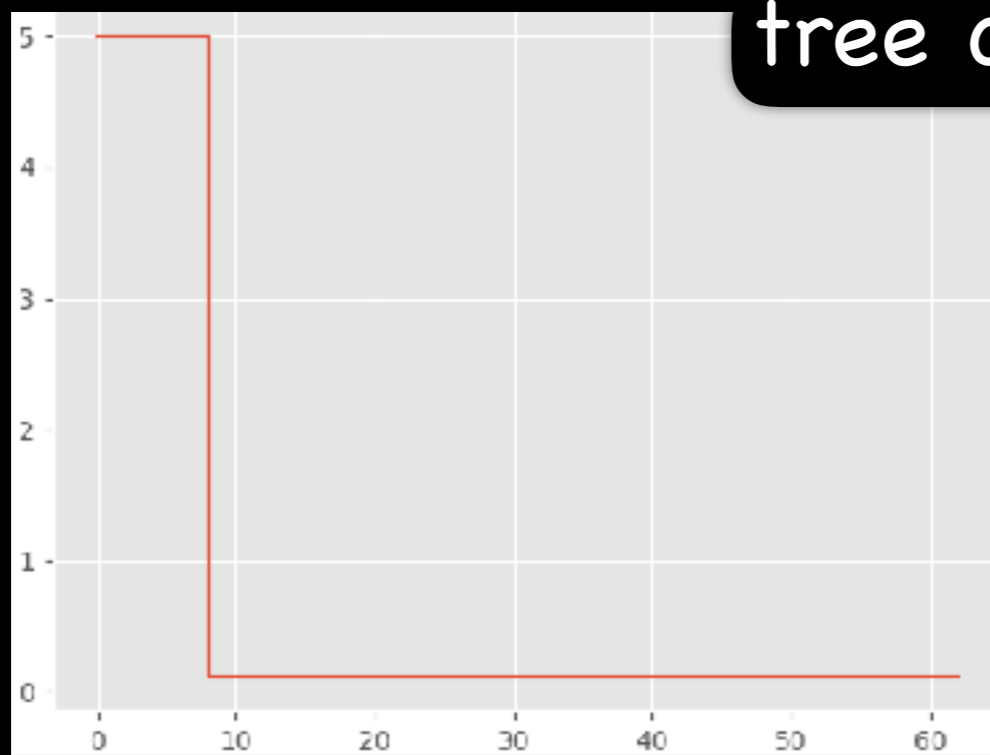
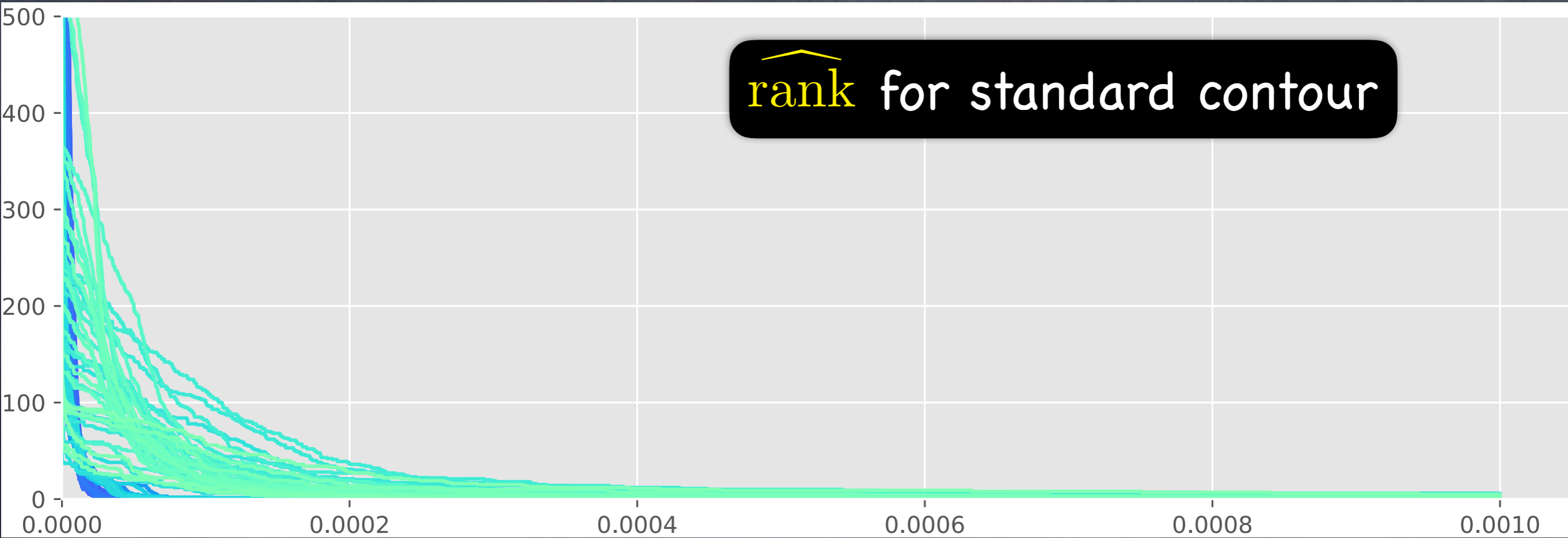
standard contour



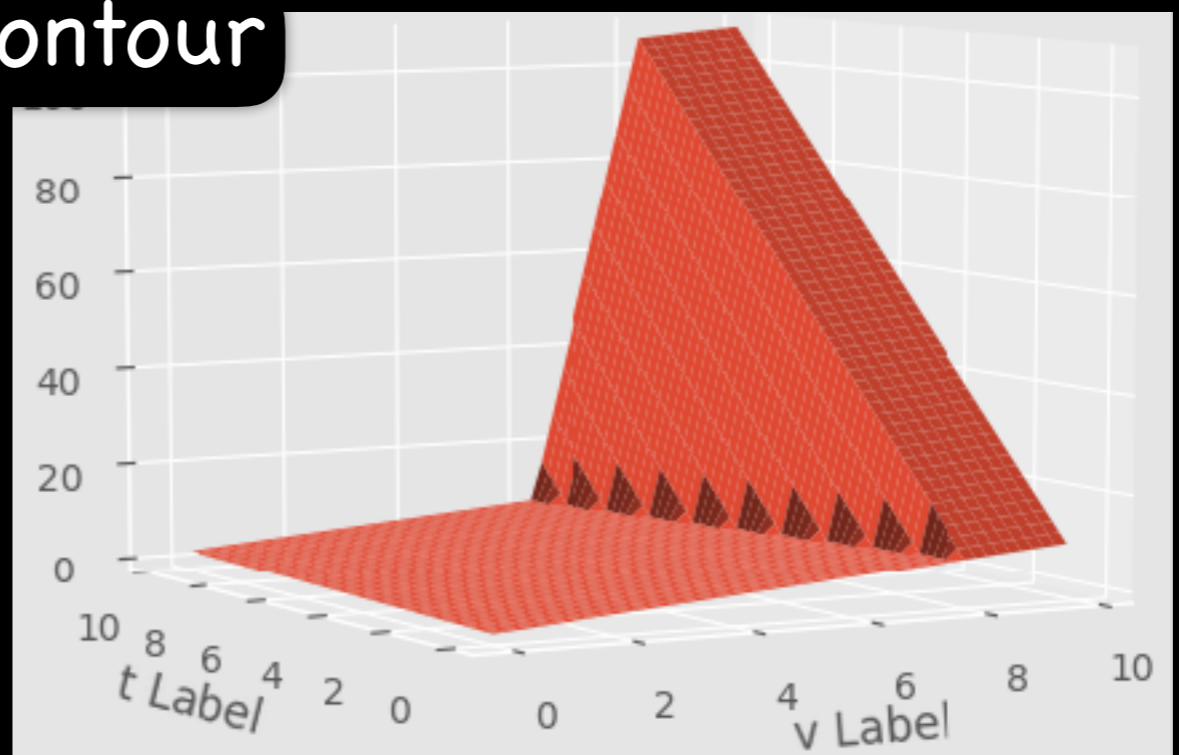
tree contour



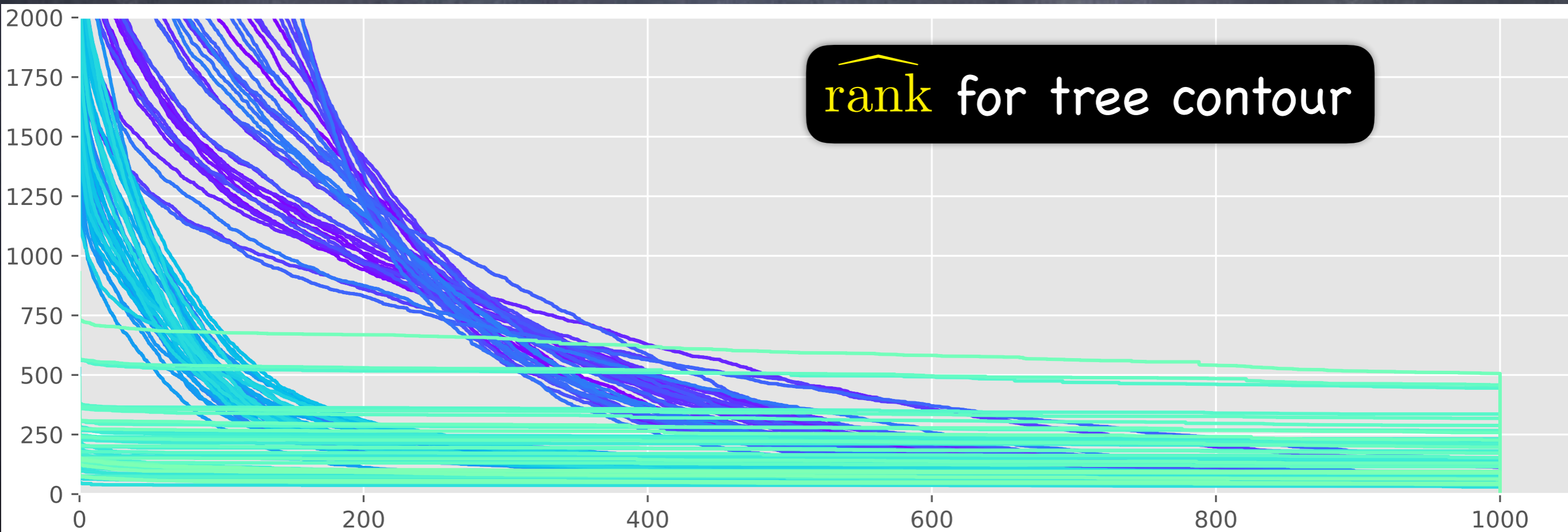
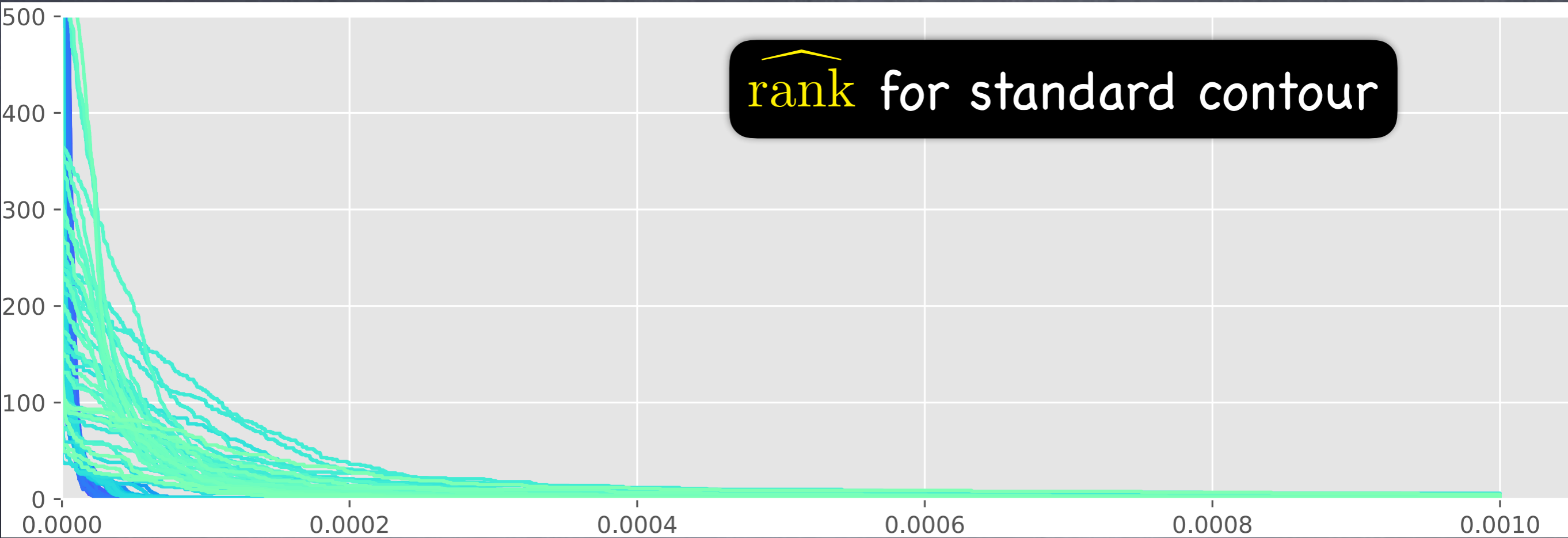
Henri Riihimäki, forest tree data:



tree contour



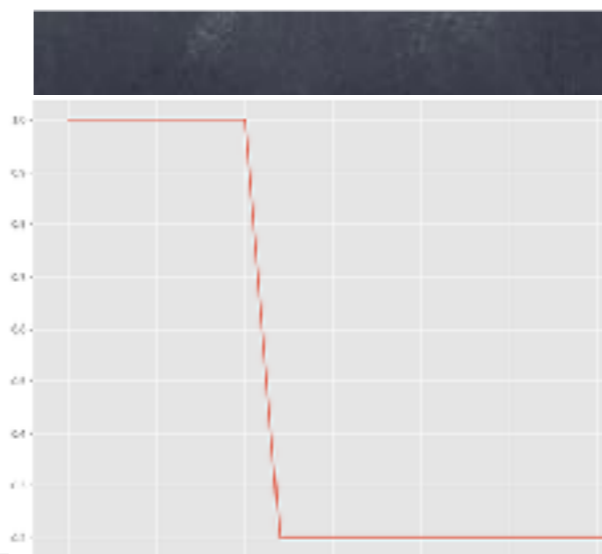
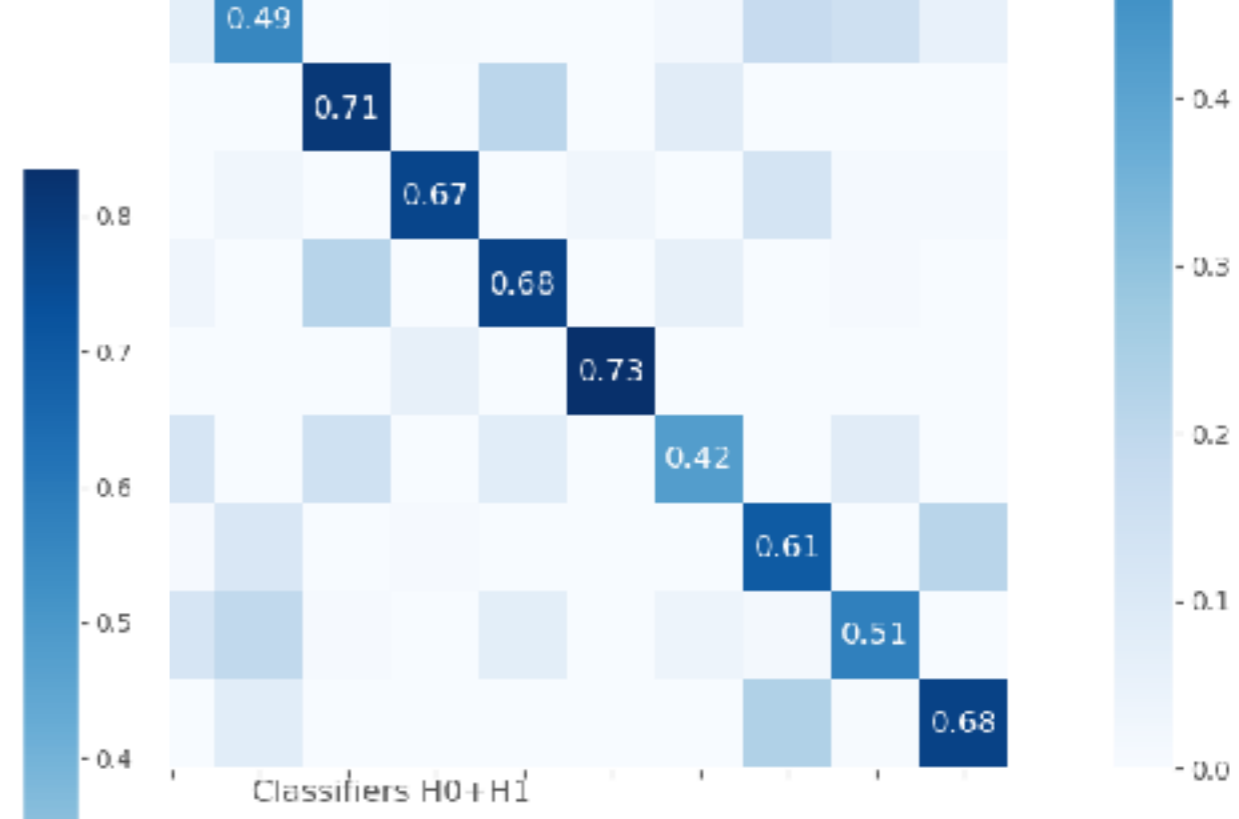
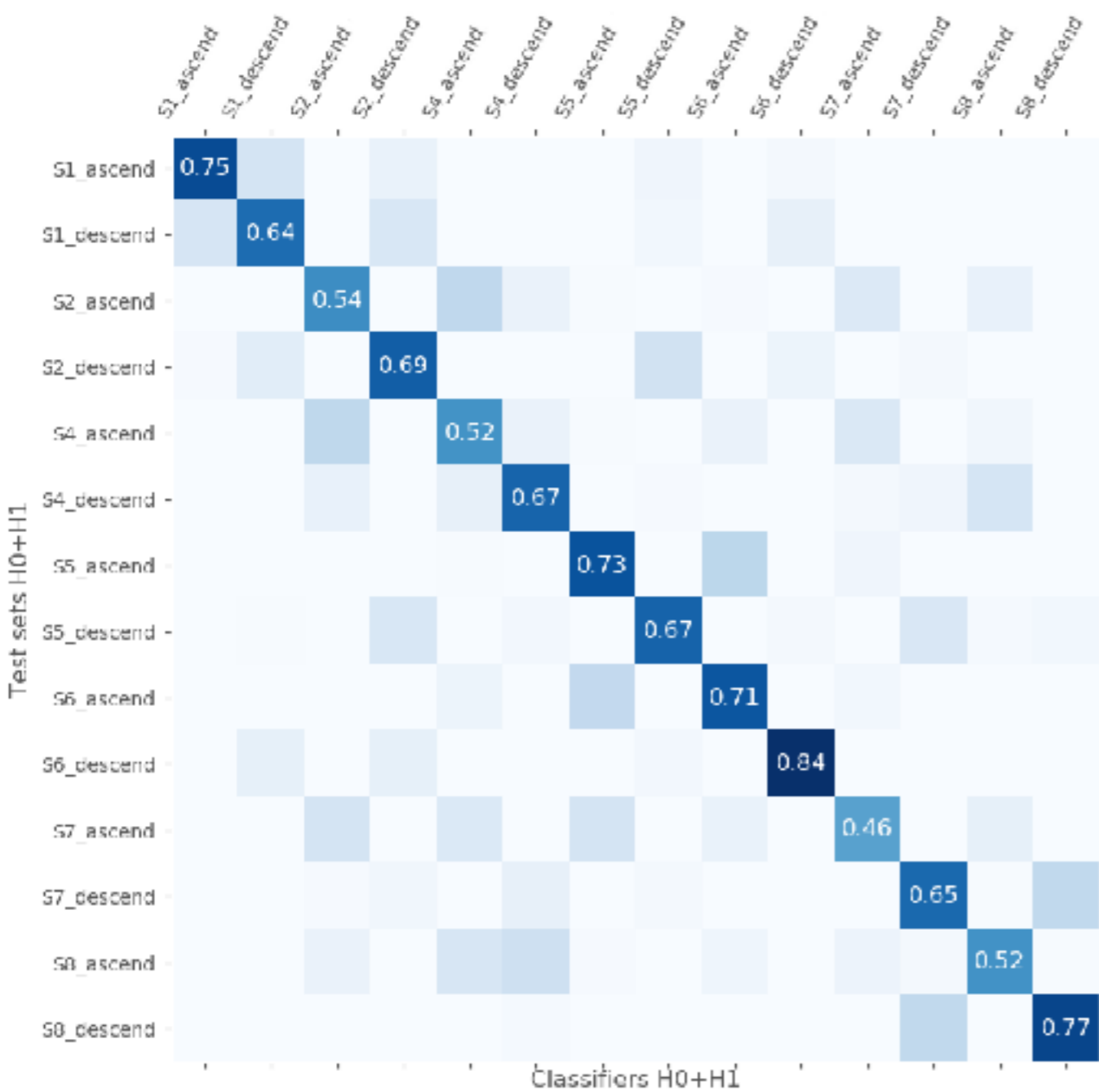
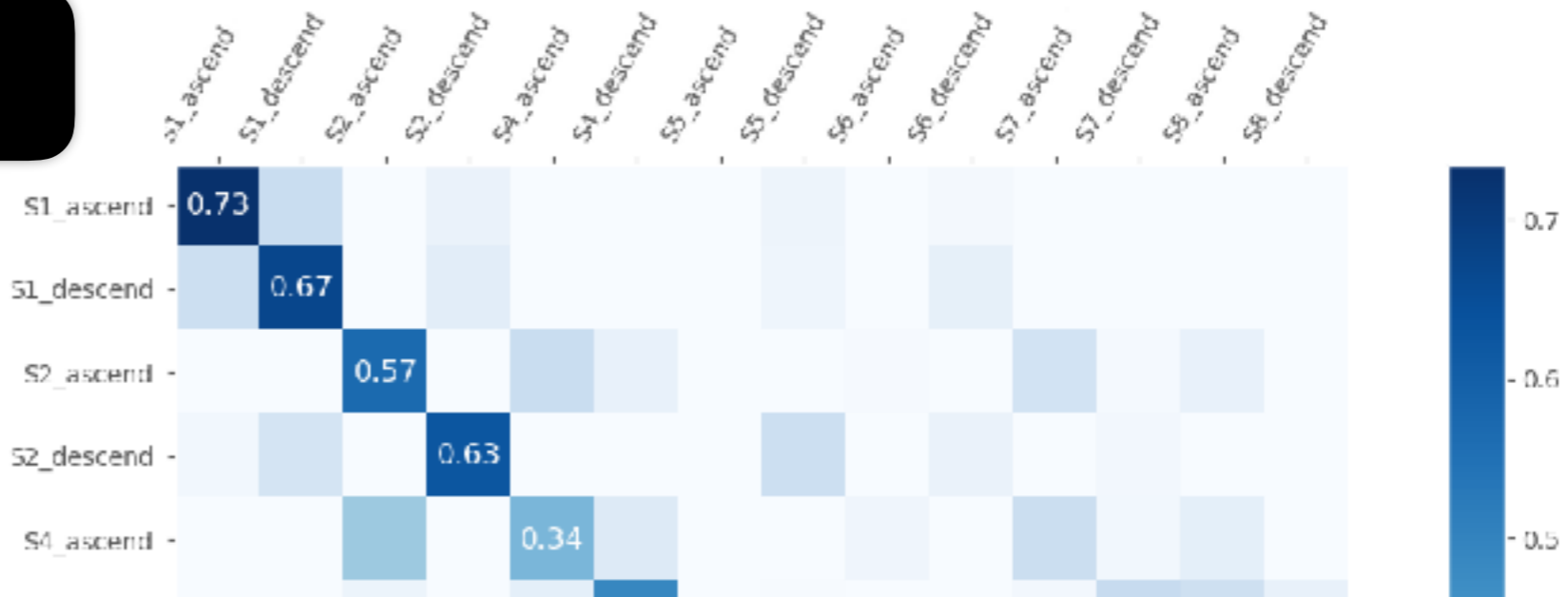
Henri Riihimäki, forest tree data:



PAMAP2 DATA (www.pamap.org):

- 7 subjects were equipped with 3 sensors (IMU)
- producing 28 measurements every 0.1 second
- in the following analysis we considered two activities of ascending and descending stairs.
- for every subject we have two time series in 28 dim space
- for each series we sample 100 time slots without replacement
- we took 100 such samplings
- 40 are used as training and 60 for testing
- after choosing a contour, the average of the 40 train data are used as classifiers

Standard contour



Thank you!