

Modeling a functional curve in a digital image

Image

Curve image in gray scale: `M := Mean (image2rgb ("img.gif"))`

Validation of the data computing the average value of each row and column, then compares this value with kline which has represents the gray value selected to define the maximum "black" pixel.

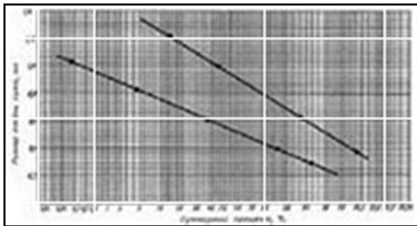
```
clean (M#, k11, k12) := [ M := M# Mc := M R := 0 C := 0
                        [ m := [ 1..rows (M) ] n := [ 1..cols (M) ]
                        [ Rm := Mean (row (M, m)) Cn := Mean (col (M, n)) ]
                        [ R C ] := eval ( [  $\overrightarrow{k_{11} \leq R \leq k_{12}}$   $\overrightarrow{k_{11} \leq C \leq k_{12}}$  ] )
                        for m ∈ [ 1..rows (M) ]
                          for n ∈ [ 1..cols (M) ]
                            if Rm ∧ Cn
                              continue
                            else
                              Mcm n := 255
                        Mc
```

`k1 := 130`

`Mc := clean (M, k1, 255)`

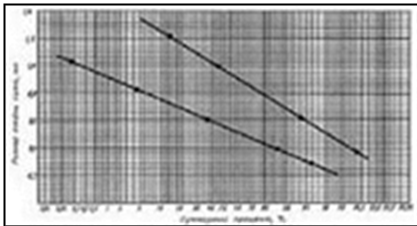
Try for different k1 values until find a good one.

Cleaned data



Mc

Original data



M

With the matrix cleaned we can look for pixels from the function curve. With the following function we can scan the cleaned image matrix in the vertical and horizontal directions to find pixels that belong to the curve; the maximum opaque color is contained in the variable kcurve.

```
scan (M, kc1, kc2) := [ nc := cols (M) n := [ 1..length (M) ] A := 0
                        v := col ( findrows ( eval ( augment (  $\overrightarrow{k_{c1} \leq M \leq k_{c2}}$ , n ) ), 1, 1 ), 2 )
                        [ r c ] := eval ( [  $\overrightarrow{1 + \text{trunc} \left( \frac{v-1}{nc} \right) 1 + \text{mod} (v-1, nc)}$  ] )
                        eval ( csort ( augment ( c, rows (M) - r + 1 ), 1 ) )
```

`kc := 80`

`Ms := scan (Mc, 0, kc)`

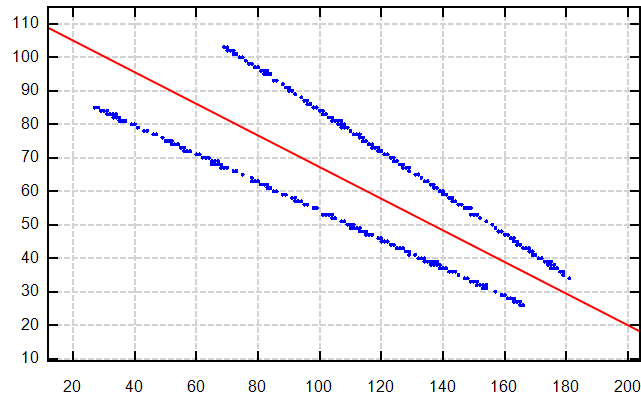
Try for different kc values until find a good one.

Also isolate the curves selectring adequate ranges

Read range values here `xr := [ 20..200 ]` `yr := [ 20..105 ]`

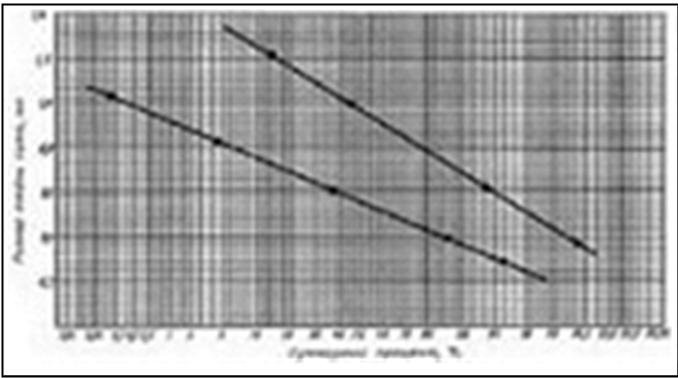
$$separe(x) := \frac{\max(yr) - \min(yr)}{\max(xr) - \min(xr)} \cdot (\max(xr) - x) + \min(yr)$$

Scanned data



```
{
augment(Ms, ".")
separe(x)
}
```

Original data



M

Now we can fit the data according to some model

$$\lambda(x, m, n) := \overrightarrow{m \cdot x + n}$$

```
A := eval(matrix(0, 2))    B := A

for r ∈ [1..rows(Ms)]
  if separe(Ms_r 1) < Ms_r 2
    A := eval(stack(A, row(Ms, r)))
  else
    B := eval(stack(B, row(Ms, r)))
```

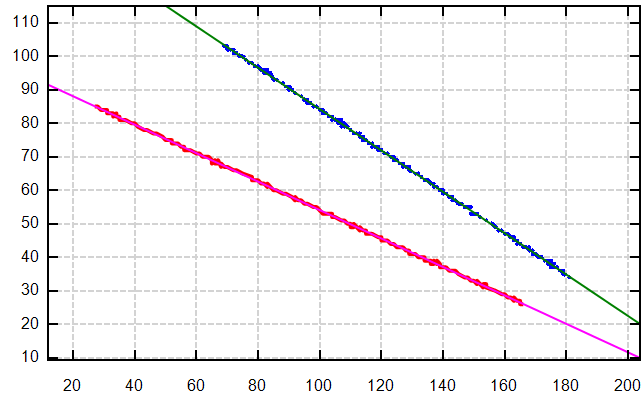
$$\lambda(u, M) := \lambda\left(\text{col}(M, 1), u_1, u_2\right) - \text{col}(M, 2)$$

$$\begin{bmatrix} mA \\ nA \end{bmatrix} := \text{al_nleqsolve}\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \lambda(u) := \lambda(u, A)\right)$$

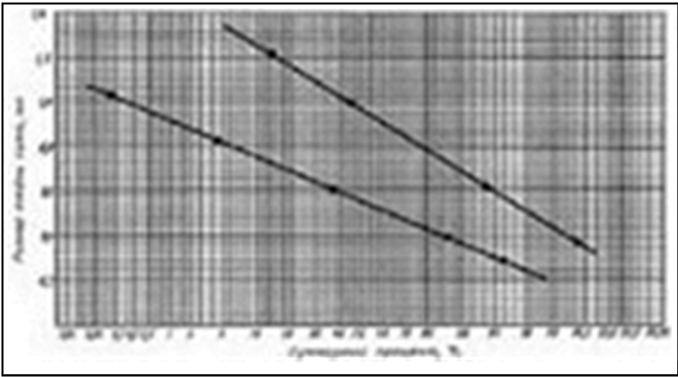
$$\begin{bmatrix} mB \\ nB \end{bmatrix} := \text{al_nleqsolve}\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \lambda(u) := \lambda(u, B)\right)$$

Results ToDo: scale the plot

$$\begin{bmatrix} mA \\ nA \end{bmatrix} = \begin{bmatrix} -0.6174 \\ 145.9792 \end{bmatrix} \qquad \begin{bmatrix} mB \\ nB \end{bmatrix} = \begin{bmatrix} -0.4247 \\ 96.5775 \end{bmatrix}$$



```
{
A
B
λ(x, mA, nA)
λ(x, mB, nB)
}
```



M