Notations

We denote:

- Pressure : $\nu \in [0,1]$
- time : t > 0
- point : defined by position x and y (or $p = (x, y) \in \mathbb{R}^2$)
- Raw inputs are denoted by a tuple $i[k] = (\nu[k], t[k], x[k], y[k])$ with $k \in \mathbb{N}$.

An *input stream* is a sequence of raw inputs $i = (\nu, t, x, y)$

- with time $t[k] \nearrow \nearrow$ strictly increasing
- · starts with a kDown event
- contains KMove for $k \ge 1$
- ends either with a kMove or a kUp . If it is a kUp we say the input stream is complete

We addition define

- v_x, v_y, a_x, a_y as the velocity and acceleration.

With the vector shorthand $v=\left(v_x,v_y\right)\in\mathbb{R}^2$ and $a=\left(a_x,a_y\right)\in\mathbb{R}^2$

Wobble smoothing

To reduce high frequency noise.

Algorithm 1: Wobble smoothing

 $\begin{aligned} &\textbf{Input}: \big\{(x[k],y[k],t[k]) \in \mathbb{R}^2 \times \mathbb{R}_+, 0 \leq k \leq n\big\}, \Delta T > 0 \text{ (from wobble_smoother_timeout)}, v_{\min} \text{ (from wobble_smoother_speed_floor) and } v_{\max} \end{aligned}$ (from wobble_smoother_speed_ceiling)

Compute a weighted moving average of the positions $\overline{p}[j] = (\overline{x}[j], \overline{y}[j])$

$${}_{2}\,\forall j\in\llbracket 0,n\rrbracket,\overline{p}[j]=\left\{ \begin{aligned} &\sum_{\underline{k=1}}^{n}p[k](t[k]-t[k-1])\mathbb{1}_{\left[t[j]-\Delta T,t[j]\right]}(t[k])\\ &\sum_{\underline{k=1}}^{n}\mathbb{1}_{\left[t[j]-\Delta T,t[j]\right]}(t[k])\\ &p[j] \end{aligned} \right. \text{ otherwise}$$

Calculate a moving average velocity $\overline{v}[j]$

$$\forall j \in [\![0,n]\!], \overline{v}[j] = \begin{cases} 0 & j = 0 \\ \sum_{k=1}^n \lVert p[k] - p[k-1] \rVert \mathbb{1}_{\left[t[j] - \Delta T, t[j]\right]}(t[k]) \\ \sum_{k=1}^n (t[k] - t[k-1]) \mathbb{1}_{\left[t[j] - \Delta T, t[j]\right]}(t[k]) \end{cases} \quad \text{otherwise}$$

Interpolate between the average position and the raw ones based on the average speed

$$\begin{split} \forall j \in [\![0,n]\!], p'[j] &= \min \biggl(\frac{\overline{v}[j] - v_{\min}}{v_{\max} - v_{\min}} \mathbb{1}_{[v_{\min}, \infty[}(\overline{v}[j]), 1 \biggr) \overline{p}[j] \\ &+ \biggl(1 - \min \biggl(\frac{\overline{v}[j] - v_{\min}}{v_{\max} - v_{\min}} \mathbb{1}_{[v_{\min}, \infty[}(\overline{v}[j]) \biggr) \biggr) p[j] \end{split}$$

where p'[j] = (x'[j], y'[j])

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Output: $\{(x'[k], y'[k]) \in \mathbb{R}^2, 0 \le k \le n\}$ the filtered positions

Hence for low local speeds, the smoothing is maximum (we take exactly the average position over the time ΔT) and for high speed there is no smoothing. We also note that the first position is thus never filtered.

Resampling

Algorithm 2: Upsamling

 ${\bf Input}: \{v[k] = (x[k], y[k], t[k]), 0 \le k \le n\}$ and a target rate sampling_min_output_rate

Interpolate time and position between each k by linearly adding interpolated values. This is done by adding linearly interpolated values so that the output stream is

$$\left\{v[0], \underbrace{u_0[1], ..., u_0[n_0-1]}_{\text{interpolated}}, v[1], \underbrace{u_1[1], ..., u_1[n_1-1]}_{\text{interpolated}}, v[2], ...\right\}$$

Each n_i is the minimum integer such that

$$\begin{split} \frac{t[i] - t[i-1]}{n_i} < \Delta_{\text{target}} \\ \Leftrightarrow n_i = \left\lceil \frac{t[i+1] - t[i]}{\Delta_{\text{target}}} \right\rceil \end{split}$$

and the linear interpolation means

$$u_j[k] = \left(1 - \frac{k}{n_j}\right) v[j] + \frac{k}{n_i} v[j+1]$$

Output : $\{(x'[k'], y'[k'], t'[k']), 0 \le k; \le n'\}$ the upsampled position and times. This verifies

$$\forall k', t'[k'] - t'[k'-1] < \Delta_{\text{target}} = \frac{1}{\text{sampling_min_output_rate}}$$

Remark: As this is a streaming algorithm, we only calculate this interpolation with respect to the latest stroke position.

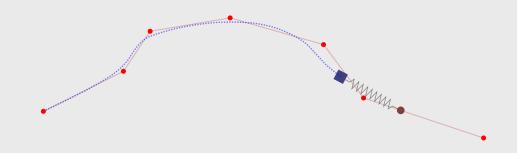
Position modeling

The raw input is processed as follow

raw input \rightarrow wobble smoother \rightarrow upsampled \rightarrow position modeling

The position of the pen is modeled as a weight connected by a spring to an anchor.

The anchor moves along the *resampled dewobbled inputs*, pulling the weight along with it accross a surface, with some amount of friction. Euler integration is used to solve for hte position of the pen.



The physical model that is used to model the stroke is the following

$$\frac{\mathrm{d}^2 s}{\mathrm{d}t^2} = \frac{\Phi(t) - s(t)}{k_{\mathrm{spring}}} - k_{\mathrm{drag}} \frac{\mathrm{d}s}{\mathrm{d}t}$$

where

- ullet t is time
- s(t) is the position of the pen
- $\Phi(t)$ is the position of the anchor
- + $k_{
 m spring}$ and $k_{
 m drag}$ are constants that sets how the spring and drag occurs

 $k_{\rm spring}$ is given by position_modeler_spring_mass_constant and $k_{\rm drag}$ by position_modeler_drag_constant

We will thus have as input the *upsampled dewobbled inputs* taking the role of discretized $\Phi(t)$ and s(t) will be the output

Modeling a stroke.

- Input : input stream $\{(p[k],t[k]), 0 \leq k \leq n\}$
- Output : smoothed stream $\left\{\left(p_f[k], v_f[k], a_f[k]\right), 0 \leq k \leq n\right\}$

We define $\Phi[k]=p[k]$. An euler scheme integration scheme is used with the initial conditions being v[0]=0 and $p_f[0]=p[0]$ (same initial conditions)

Update rule is simply

$$\begin{split} a_f[j] &= \frac{p[j] - p_f[j-1]}{k_{\text{spring}}} - k_{\text{drag}} v_f[j-1] \\ v_f[j] &= v_f[j-1] + (t[j] - t[j-1]) a_f[j] \\ p_f[j] &= p_f[j-1] + (t[j] - t[j-1]) v_f[j] \end{split}$$

The position s[j] is the main thing to export but we can also export speed and acceleration if needed. We denote

$$q[j] = \left(p_f[j], v_f[j], a_f[j], t[j]\right)$$

and this will be our output with $0 \le j \le n$

Stroke end

The position modeling algorithm will lag behind the raw input by some distance. This algorithm iterates the previous dynamical system a few additional time using the raw input position as the anchor to allow a catch up of the stroke (though this prediction is only given by predict, so is not part of the results and becomes obsolete on the next input).

Algorithm 3: Stroke end

Input:

```
• Final anchor position p[end] = (x[end], y[end])
                                                                                                  ▶ From the original input stream
     • final tip state q_f[\mathrm{end}] = (p_f[\mathrm{end}] = (x_f[\mathrm{end}], y_f[\mathrm{end}]), v_f[\mathrm{end}], a_f[\mathrm{end}]) preturned
          from the physical modeling from the last section, \cdot_f signifies that we are looking at the filtered output
       K_{\rm max} max number of iterations
                                                                                    ⊳ sampling_end_of_stroke_max_iterations
     - \Delta_{\mathrm{target}} the target time delay between stroke
                                                                                                  ▷ 1/sampling_min_output_rate
     • d_{\rm stop} stopping distance
                                                                                ⊳ sampling_end_of_stroke_stopping_distance
     • k_{
m spring} and k_{
m drag} the modeling coefficients
     initialize \Delta t = \Delta_{\mathrm{target}}
     for 1 \le k \le K_{\text{max}}
         calculate the next candidate
       a_c = \frac{p[\mathrm{end}] - p_o[\mathrm{end}]}{k_{\mathrm{spring}}} - k_{\mathrm{drag}} v_0[\mathrm{end}]
        v_c = v_o[0] + \Delta t a_c
        p_c = p_o[0] + \Delta v_c
        	ext{ if } \|p_c - p[	ext{end}]\| < d_{	ext{stop}} 	ext{ further iterations won't be able to catch up and won't move closer to the anchor,
         return q_0
 6
 7
         \textbf{if } \langle p_c - p_o[\text{end}], p[\text{end}] - p_o[\text{end}] \rangle < \|p_c - p_o[\text{end}]\| \quad \text{$\,$} \text{ we've overshot the anchor, we retry}
 8
          with a smaller step
         \Delta t \leftarrow \frac{\Delta t}{2}
10
         q_0[\text{end } +1] = (p_c, v_c, q_c)
11
                                                                               \,{\scriptscriptstyle\,\,\triangleright}\,\, We append the result to the end of the q_0 vector
12
         \mathbf{if} \| p_c - p[\text{end}] \| < d_{\text{stop}}
13
         return
14
                                                                       ▶ We are within tolerance of the anchor, we stop iterating
15
         endif
         \textbf{Output}: \{q_o[k] = (s_o[k], v_o[k], a_o[k]), 0 \leq k \leq n (\leq K_{\max} - 1)\}
```

Stylus state modeler

Up till now we have only used the raw input stream to create a new smoothed stream of positions, leaving behind the pressure attribute. This is what's done here, to model the state of the stylus for these new position based on the pressure data of the raw input strokes.

Algorithm 4: Stylus state modeler

```
Input:
```

- input stream with pressure information $\{(p[k] = (x[k], y[k]), \nu[k]), 0 \le k \le n\}$
- query position q = (x, y)
- initialize $d = \infty$, index = None, interp = None
- 2 **for** i = 0 to n 1 **do**
- Find q_i the position that's closest to q on the segment [p[i], p[i+1]] and denote
- $r \in [0,1]$ the value such that $q_i = (1-r)p[i] + rp[i+1]$
- $\begin{array}{ccc} \mathbf{if} \ \|q q_i\| < d \\ d \leftarrow \|q q_i\| < d \end{array}$
- index = i
- interp = r
- 6 endif
- 7 endfor
- calculate

o carcurat

 $\nu = (1 - r)\nu[\text{index}] + r\nu[\text{index} + 1]$

Output : interpolated pressure ν