





# Array Model / Point Cloud Matching Cost ("Sample")

$$J(p) = \underbrace{R^T C}_{\text{pts match cost}} + \underbrace{P_e^T Q P_e}_{\text{regularization cost}} \quad P_e = P_{\text{norm}} P$$
$$= \sum_{i=1}^m R_i C_i \quad + \quad R_i = \frac{1}{m} \rightarrow \text{average cost}$$

$\forall i$

## Cost per measurement

$$C_i = \log_{\text{base}} \left( 1 + \frac{[b d_i]^{\text{power}}}{l} \right)$$

## error/distance to pts

$$d_i = \min_{j \in K} \| \underline{r}_i^w - \underline{r}_{jk}^w \| = \min_K \left[ \underbrace{\min_j \| \underline{r}_i^w - \underline{r}_{jk}^w \|}_{\text{closest distance to each } K \text{ cable}} \right]$$

$\underbrace{\hspace{10em}}_{\text{closest of all cable}}$

~~if wrong~~

## Model points

$$\underline{r}_{jk}^w = \begin{bmatrix} c_4 & -s_4 & 0 \\ s_4 & c_4 & 0 \\ 0 & 0 & z_j \end{bmatrix} \begin{pmatrix} x_j + x_k \\ y_k \\ \text{cat}(x_j) + z_k \end{pmatrix} + \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix}$$

cat

$$z_j = \text{cat}(x_j) = a \left[ \cosh\left(\frac{x_j}{a}\right) - 1 \right]$$



# Array Model / Point Cloud Matching Cost

$$J(p) = \underline{R}^T \underline{C} + \underline{p}^T \underline{Q} \underline{p} \quad \underline{p}_e = \underline{p}_{nom} - \underline{p}$$

$$\boxed{\frac{\partial J}{\partial \underline{p}}} = \underline{R}^T \frac{\partial \underline{C}}{\partial \underline{p}} + 2 \underline{p}^T \underline{Q} \frac{\partial \underline{p}}{\partial \underline{p}}^{-1}$$

$$= \sum_i^n R_i \left( \frac{\partial C_i}{\partial \underline{p}} \right) + 2 [\underline{p}_{nom} - \underline{p}]^T \underline{Q}$$

$$\boxed{\frac{\partial C_i}{\partial \underline{p}}} = \frac{\partial C_i}{\partial d_i} \frac{\partial d_i}{\partial e_i} \frac{\partial e_i}{\partial r_{ijk}} \frac{\partial r_{ijk}}{\partial \underline{p}}$$

closest point  $j^*$  on closest cable  $k^*$

$$C_i = \log_{base} \left( 1 + \frac{[b d_i]^{power}}{l} \right) \quad \left| \quad \frac{\partial C_i}{\partial d_i} = \frac{1}{\ln(base)} \left( \frac{l}{l + [b d_i]^{power}} \right) \frac{b^{power} [b d_i]^{power-1}}{l} \right.$$

$$= \frac{b^{power} [b d_i]^{power-1}}{\ln(base) [l + [b d_i]^{power}]}$$

$$d_i = \sqrt{\underline{e}_i^T \underline{e}_i} = (\underline{e}_i^T \underline{e}_i)^{1/2} \quad \left| \quad \frac{\partial d_i}{\partial \underline{e}_i} = \frac{2 \underline{e}_i^T}{2 (\underline{e}_i^T \underline{e}_i)^{1/2}} = \frac{\underline{e}_i^T}{d_i} \right.$$

$$\underline{e}_i = \begin{bmatrix} x_i \\ y_i \\ z_i \end{bmatrix}^w - \begin{bmatrix} c_4 & -s_4 & 0 \\ s_4 & c_4 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{pmatrix} x_j + x_k \\ y_k \\ \cosh(x_j) + z_k \end{pmatrix}^c - \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix}^w$$

$$\frac{\partial \underline{e}_i}{\partial \underline{p}} = \begin{bmatrix} -1 & 0 & 0 & [x_j + x_k] s_4 + y_k c_4 \\ 0 & -1 & 0 & -[x_j + x_k] c_4 + y_k s_4 \\ 0 & 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 + \frac{x_j}{a} \sinh\left(\frac{x_j}{a}\right) - \cosh\left(\frac{x_j}{a}\right) \end{bmatrix} \begin{bmatrix} -c_4 & +s_4 & 0 \\ -s_4 & -c_4 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

$\underline{p}^w - \underline{R}^c$

$x_k \quad y_0 \quad z_0$



# Gradient Cost "Sample"

$$e_i = \begin{bmatrix} x_i \\ y_i \\ z_i \end{bmatrix}^w - \begin{bmatrix} c4 & -s4 & 0 \\ s4 & c4 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{pmatrix} x_j + x_k \\ y_k \\ \text{cat}(x_j) + z_k \end{pmatrix}^c - \underbrace{\begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix}^w}_{\text{internal offsets}}$$

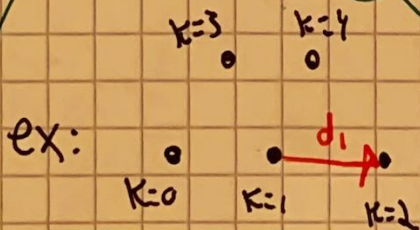
$$\underline{p}^T = [x_0; y_0; z_0; 4 \quad ; \quad a \quad ; \quad d_1; d_2 \dots]$$

$$\frac{\partial e_i}{\partial \underline{p}} = \begin{bmatrix} -1 & 0 & 0 & \vdots & [x_j + x_k]s4 + [y_k]c4 & \vdots & 0 \\ 0 & -1 & 0 & \vdots & -[x_j + x_k]c4 + [y_k]s4 & \vdots & 0 \\ 0 & 0 & -1 & \vdots & 0 & \vdots & 1 + \frac{x_j}{a} \sinh\left(\frac{x_j}{a}\right) - \cosh\left(\frac{x_j}{a}\right) \end{bmatrix}$$

offsets

$$\underline{r}_k = \begin{bmatrix} x_k \\ y_k \\ z_k \end{bmatrix} = f(\overbrace{d_1, \dots, d_e}^{\text{offsets}})$$

$$\frac{\partial e_i}{\partial d_e} = \frac{\partial e_i}{\partial \underline{r}_k} \cdot \frac{\partial \underline{r}_k}{\partial d_e} = \underbrace{\begin{bmatrix} -c4 + s4 & 0 \\ -s4 & -c4 & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{\substack{w \\ -R(\psi)}} \underbrace{\begin{bmatrix} \frac{\partial \underline{r}_k}{\partial d_e} \end{bmatrix}}_{\substack{c \\ \text{supplied by model}}}$$



$$\underline{r}_1 = \begin{bmatrix} 0 \\ d_1 \\ 0 \end{bmatrix} \quad \frac{\partial \underline{r}_1}{\partial d_1} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$



# Array Model / Point Cloud Matching cost ("X")

$$J(p) = \underbrace{R^T C}_{\text{pts match cost}} + \underbrace{P_e^T Q P_e}_{\text{regulation term}} \quad \rightarrow \quad P_e = P_{\text{nom}} - P$$

cost per measurement

$$C_i = \log_{\text{base}} \left( 1 + \frac{[b d_i]^{\text{power}}}{l} \right)$$



distance to closest catenary

$$d_i = \sqrt{e_i^T e_i} = \|e_i\|$$

error in local cat frame

$$d_i = \min_K \|e_{iK}\| = \min_K \|r_i^{c*} - r_{jK}^{c*}\|$$

measurement i  
closest point j on curve K

(closest point on catenary curve K

$$x_j^c + x_K^c = x_i^c$$

$$r_{jK}^c = \begin{bmatrix} x_j \\ 0 \\ \text{cat}(x_j) \end{bmatrix} + \begin{bmatrix} x_K \\ y_K \\ z_K \end{bmatrix} = \begin{bmatrix} x_i^c \\ y_K \\ \text{cat}(x_i^c - x_K^c) + z_K \end{bmatrix}$$

$$r_i^c = \begin{bmatrix} c_4 & s_4 & 0 \\ -s_4 & c_4 & 0 \\ 0 & 0 & 1 \end{bmatrix} \left( \underbrace{\begin{bmatrix} x_i^w \\ y_i^w \\ z_i^w \end{bmatrix}}_{r_i^w} - \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix} \right)$$

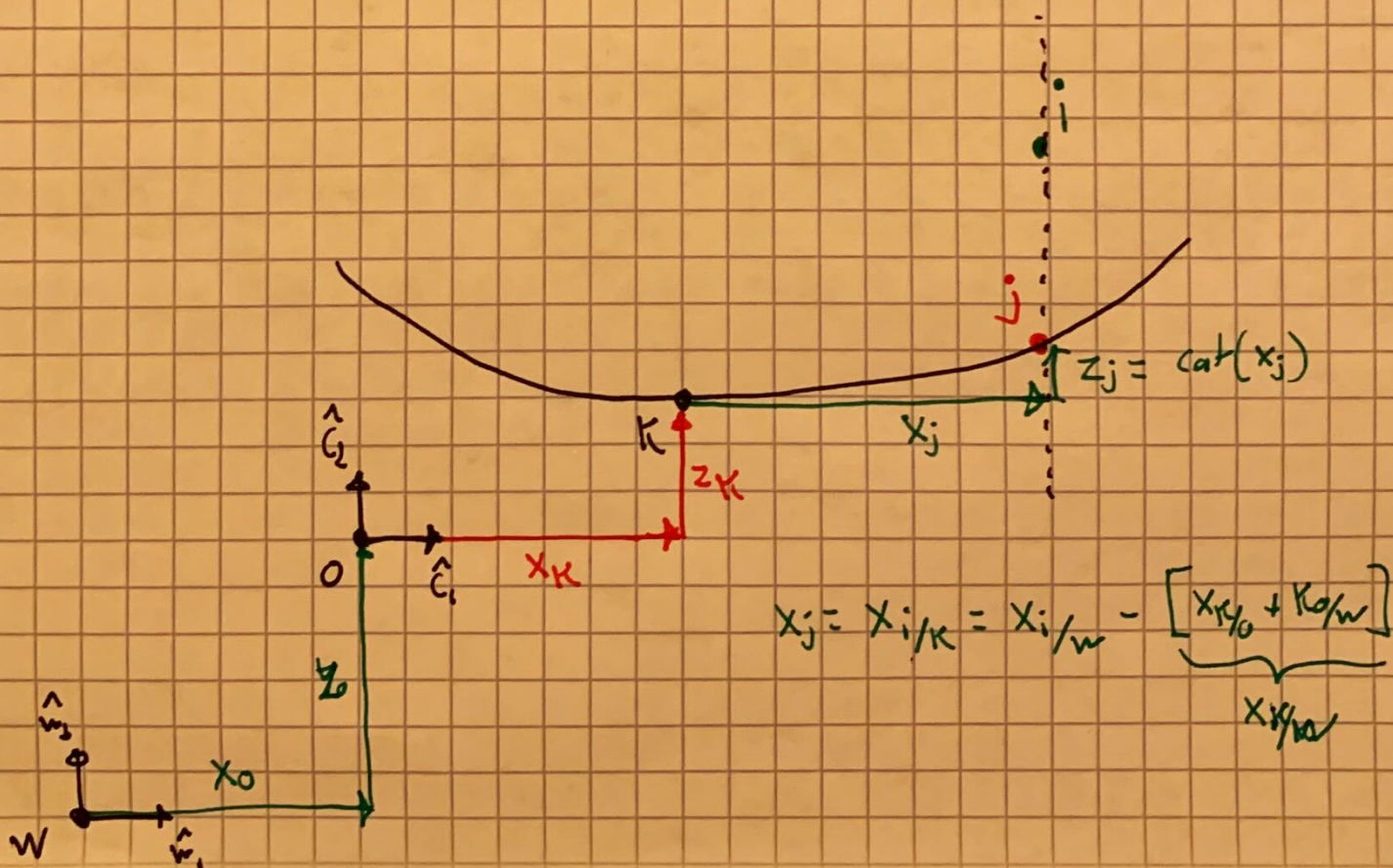


# Array/point cloud Matching Cost (part 2) ("x")

$$e_{ik}^c = \begin{bmatrix} c\psi & s\psi & 0 \\ -s\psi & c\psi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_i^w \\ y_i^w \\ z_i^w \end{bmatrix} - \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix} - \begin{bmatrix} x_k^c \\ y_k^c \\ \underbrace{\text{cat}(x_i^c - x_k^c)}_{z_{j/k}} + z_k \end{bmatrix}$$

$$e_{ik}^c = \begin{bmatrix} 0 \\ -s\psi [x_i^w - x_0] + c\psi [y_i^w - y_0] - y_k \\ z_i^w - z_k - z_0 - \text{cat}(c\psi [x_i^w - x_0] + s\psi [y_i^w - y_0] - x_k) \end{bmatrix}$$

$$\text{cat}(x) = a \left[ \cosh\left(\frac{x}{a}\right) - 1 \right]$$





# Gradient of "x" array cost

→ Same as "sample" up to

$$\frac{\partial C_i}{\partial p} = \underbrace{\frac{\partial C_i}{\partial d_i}}_{\text{same}} \underbrace{\frac{\partial d_i}{\partial e_i}}_{\text{new}} \frac{\partial e_i}{\partial p}$$

$$z_j = \text{cat}(x_j)$$

$$e_i = \begin{bmatrix} 0 \\ -s\psi[x_i^w - x_0] + c\psi[y_i^w - y_0] - y_k \\ z_i^w - z_k - z_0 - \underbrace{\text{cat}(c\psi[x_i^w - x_0] + s\psi[y_i^w - y_0] - x_k)}_{x_j} \end{bmatrix}$$

$$p^T = [x_0 \quad y_0 \quad z_0 \quad ; \quad \psi \quad ; \quad a \quad ; \quad \dots \quad d_e \quad \dots]$$

$$\frac{\partial e_i}{\partial p} = \begin{bmatrix} 0 & 0 & 0 & ; & 0 & ; & 0 \\ s\psi & -c\psi & 0 & ; & [x_0 - x_i^w]c\psi + [y_0 - y_i^w]s\psi & ; & 0 \\ -\frac{\partial z_j}{\partial x_0} & -\frac{\partial z_j}{\partial y_0} & -1 & ; & -\frac{\partial z_j}{\partial \psi} & ; & -\frac{\partial z_j}{\partial a} \end{bmatrix}$$

$$\dots \frac{\partial e_i}{\partial d_e} = \underbrace{\frac{\partial e_i}{\partial r_k}}_{\text{supplied by model}} \frac{\partial r_k}{\partial d_e} \dots$$

$$\left[ \frac{\partial e_i}{\partial r_k}, \frac{\partial e_i}{\partial y_k}, \frac{\partial e_i}{\partial z_k} \right]$$



$$\frac{\partial e_i}{\partial x_k} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -1 & 0 \\ -\frac{\partial z_j}{\partial x_k} & 0 & -1 \end{bmatrix}$$

$$z_j = \text{cat}(x_j) = a \left[ \cosh\left(\frac{x_j}{a}\right) - 1 \right]$$

$$x_j = c4 [x_i^w - x_0] + s4 [y_i^w - y_0] - x_k$$

$$\frac{\partial z_j}{\partial a} = \frac{\partial [a \cosh(\frac{x_j}{a}) - a]}{\partial a} = \cosh\left(\frac{x_j}{a}\right) + a \frac{\partial [\cosh(\frac{x_j}{a})]}{\partial a} - 1 = \cosh\left(\frac{x_j}{a}\right) - \frac{x_j}{a} \sinh\left(\frac{x_j}{a}\right) - 1$$

$$\frac{\partial z_j}{\partial x_j} = \frac{\partial [a \cosh(\frac{x_j}{a}) - a]}{\partial x_j} = \sinh\left(\frac{x_j}{a}\right)$$

$$\frac{\partial x_j}{\partial x_0} = -c4$$

$$\frac{\partial x_j}{\partial y_0} = -s4$$

$$\frac{\partial x_j}{\partial y} = s4 [x_0 - x_i^w] + c4 [y_i^w - y_0] = -s4 [x_i^w - x_0] + c4 [y_i^w - y_0]$$

$$\frac{\partial x_j}{\partial x_k} = -1$$