Mathematics and Implementation of Computer Graphics Techniques 2015

Boundary Aligned Smooth 3D Cross-Frame Field

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Background: 2D Frame Field & Quad Meshing



- 2D Frame Field
 Auto-computed
- UV Parameterization
 Auto-computed



Background: 3D Frame Field & Hex Meshing



- 3D Frame Field
 Heuristic
- UVW Parameterization
 Auto-computed

"Meta-Mesh" to define 3D Frame Field



UVW Parameterization = Hex Mesh

CubeCover - Parameterization of 3D Volumes [Nieser, Reitebuch, Polthier, SGP11]

Definition of 3D Frame

• Don't care about orientation / ordering of axes

$$R_{1} \qquad R_{2} \qquad R_{3} \qquad R_{4} \qquad \cdots$$

$$\cdot [R_{1}] = [R_{2}] = \cdots = \{R_{1}, R_{2}, \dots, R_{24}\}$$

- Question: How distant is $[R_a]$ from $[R_b]$?
- Key insight: $h(s) \coloneqq s_x^2 s_y^2 + s_y^2 s_z^2 + s_z^2 s_x^2$, $s \in S^2$
 - Invariant under sign flip / axis reordering!





Integral over an entire sphere → Spherical Harmonics!

Basics of Spherical Harmonics

• Something like Fourier series on sphere -3 -2 -1 0 1 2

$$f(\theta,\phi) = \sum_{l=0}^{\infty} \sum_{m=-l}^{l} \hat{f}_{l}^{m} Y_{l}^{m}(\theta,\phi)$$

• Orthonormality:

$$\int_{s \in S^2} Y_l^{m_1}(s) Y_l^{m_2}(s) ds = \begin{cases} 1 & \text{if } m_1 = m_2 \\ 0 & \text{otherwise} \end{cases}$$



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Frame represented by SH

h(s) "Frequency" unaffected by rotation:

$$h(s) = -\frac{2\sqrt{\pi}}{15} \left(Y_4^0(s) + \sqrt{\frac{5}{7}} Y_4^4(s) + 16\sqrt{\pi} Y_0^0(s) \right) \stackrel{\text{simplify}}{\longrightarrow} h(s) \coloneqq \sqrt{7} Y_4^0(s) + \sqrt{5} Y_4^4(s)$$

 Frame represented as SH coeffs for band l = 4 (i.e. 9-vector) :

- $f_{[R]} \coloneqq (\lambda_{-4}, \lambda_{-3}, \lambda_{-2}, \lambda_{-1}, \lambda_0, \lambda_1, \lambda_2, \lambda_3, \lambda_4)$ $\hat{h} = \hat{f}_{[I]} = (0,0,0,0,\sqrt{7},0,0,0,\sqrt{5})$
- Coeffs mapped by some 9x9 matrix \hat{R} :
- Distance between $R_a \& R_b$:

 $h(R^{T}s) =: f_{[R]}(s) = \sum \lambda_{m} Y_{4}^{m}(s)$

 $\hat{f}_{[R]} = \hat{R} \ \hat{h}$





Computing $9x9 \hat{R}$ from 3D rotation R

- Not immediately obvious
- Insight: Obvious for certain cases: $R_Z^{\theta} \& R_X^{\pi/2}$
 - (Not sure...)

Rotation about Z axis by θ

- → Represent rotation by ZYZ Euler angle
 - $R(\alpha,\beta,\gamma) \coloneqq R_Z^{\gamma} \left(R_Y^{\beta} \right) R_Z^{\alpha} = R_Z^{\gamma} \left(R_X^{-\frac{\pi}{2}} R_Z^{\beta} R_X^{\frac{\pi}{2}} \right) R_Z^{\alpha}$

•
$$\hat{R}(\alpha,\beta,\gamma) = \hat{R}_{Z}^{\gamma} \left(\hat{R}_{X}^{-\frac{\pi}{2}} \ \hat{R}_{Z}^{\beta} \ \hat{R}_{X}^{\frac{\pi}{2}} \right) \ \hat{R}_{Z}^{\alpha}$$

Frame that aligns with boundary surface

Coeff for Y_4^0

- Frame [*R*] aligns with:
 - Z axis iff $\hat{f}_{[R]}(0) = \sqrt{7}$

(Proof in Appendix)

• Surface normal *n* iff $(\hat{R}_{n \to Z} \hat{f}_{[R]})(0) = \sqrt{7}$

•
$$R_{n \to Z}$$
: Rotation that brings n to Z axis
 $\alpha = -\operatorname{atan2}(n_y, n_x), \quad \beta = -\operatorname{acos}(n_z), \quad \gamma = 0$

Discretization & Objective

- Tetrahedral mesh over domain $\boldsymbol{\Omega}$
- Frame var \hat{f}_{p_i} at center p_i of every (interior/exterior) triangle TRI_i
- Piecewise-linear frame field \hat{f}
 - Gradient $\nabla \hat{f}_{\text{TET}_j}$ constant within each tetrahedron TET_j
- Objective to be minimized:

$$E_{\text{smooth}} \coloneqq \sum_{\text{TET}_{j}} \text{volume}(\text{TET}_{j}) \sum_{m=-4}^{4} \left\| \nabla \hat{f}_{\text{TET}_{j}}(m) \right\|^{2}$$
$$E_{\text{align}} \coloneqq \sum_{\text{TRI}_{i} \in \partial \Omega} \text{area}(\text{TRI}_{i}) \left\| \left(\hat{R}_{n_{i} \to \text{Z}} \, \hat{f}_{p_{i}} \right)(0) - \sqrt{7} \right\|^{2}$$

$$E_{\text{full}} \coloneqq \frac{E_{\text{smooth}}}{\text{volume}(\Omega)^{1/3}} + w_{\text{align}} \frac{E_{\text{align}}}{\text{area}(\partial\Omega)}$$

Optimization

- Energy quadratic in $\{\hat{f}_i\} \rightarrow$ Simple Laplace-like least squares <Step 1>
- Problem: Arbitrary \hat{f}_i doesn't represent rotation!
 - → <Step 2> *Project* \hat{f}_i to its closest rotation $R(\alpha_i, \beta_i, \gamma_i)$
 - (Not sure how to do it...)
- <Step 3> Using $\Phi_i \coloneqq (\alpha_i, \beta_i, \gamma_i)$ as initial guess, run *nonlinear optimization* over $\{\Phi_i\}$
 - L-BFGS (solver: ALGLIB, dlib, etc)
 - (Not sure about analytic form of derivative...)

$$\begin{split} \hat{f}_{0} \leftarrow \arg\min_{\hat{f}} E_{f}(\hat{f}) \\ \text{for all rotation } \Phi_{i} &= (\alpha_{i}, \beta_{i}, \gamma_{i}) \text{ do} \\ \Phi_{0,i} \leftarrow \arg\min_{\Phi_{i}} \|\hat{f}_{0,i} - \hat{R}(\Phi_{i})\hat{h}\|^{2} \\ \text{end for} \\ \text{repeat} \\ \text{L-BFGS iteration for } \arg\min_{\Phi} E_{f}(\hat{f}_{[R(\Phi)]}) \\ \text{until } -\frac{\Delta E_{f}}{E_{f}} < 10^{-5} \end{split}$$





Questions

- Regarding implementation:
 - Expressions for $\hat{R}_{\rm Z}^{\theta}$ & $\hat{R}_{\rm X}^{\pi/2}$
 - Projection of \hat{f}_i to its closest rotation $R(\alpha_i, \beta_i, \gamma_i)$
 - Analytic derivative of E_{full} w.r.t. $\{\Phi_i\}$
- Possible idea for improvement:
 - Can we sidestep nonlinear optimization by alternating Laplace smoothing and "normalization"?



