

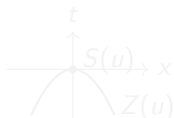
Structure theory of parabolic nodal and singular sets

Robert Koirala (UCSD)
(Joint work with Max Hallgren and Zilu Ma)

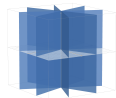
Geometry and volume of $Z(u)$ and $S(u)$: polynomial case

For a nonzero function $u : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}$, define

$$Z(u) := \{u = 0\}, \quad S(u) := \{u = 0, |\nabla u| = 0\}.$$



$$u = x^2 + 2t$$



$$u = x^4 - 6x^2y^2 + y^4$$

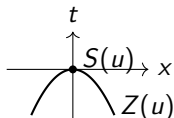
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- Wongkew ('93): $\text{Vol}(P_r(Z(u) \cap P_1)) \lesssim_d r$, $\text{Vol}(P_r(S(u) \cap P_1)) \lesssim_d r^2$, d is degree of u

Which geometric and measure-theoretic features persist beyond the polynomial setting?

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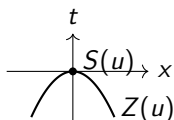
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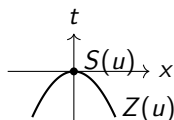
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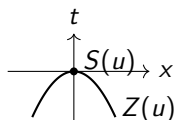
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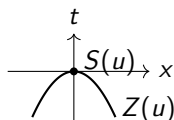
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Main results when u solves a parabolic inequality

Let u satisfy

$$|\partial_t u - \operatorname{div}(a(x, t)\nabla u)| \leq M(|u| + |\nabla u|) \quad (1)$$

on an open subset of $\mathbb{R}^n \times \mathbb{R}$, where $a = (a_{ij})$ is uniformly elliptic and

$$|a_{ij}(x, t) - a_{ij}(y, s)| \leq M \max\{|x - y|, \sqrt{|t - s|}\}.$$

Model case. When $M = 0$ and $a_{ij} = \delta_{ij}$, this is the heat equation $\partial_t u = \Delta u$, and u is **caloric**.

Main results (Halgren–K.–Ma, 2025)

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Main results (Halgren–K.–Ma, 2025)

- ① Up to negligible sets, $Z(u)$ and $S(u)$ are covered by a countable union of **regular parabolic Lipschitz graphs** of codimension 1 and 2, respectively.
- ② Space-time Minkowski estimates:

$$\operatorname{Vol}(P_r(Z(u) \cap P(\mathbf{0}, 1))) \lesssim r, \quad \operatorname{Vol}(P_r(S(u) \cap P(\mathbf{0}, 1))) \lesssim r^2,$$

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Preview of the proof: zooming in and symmetry

- After **parabolic rescaling**, a solution u is well approximated at **infinitesimal scales** by a **caloric polynomial**.
- The **volume** of $Z(u)$ and $S(u)$ is a **dimensional** information, and dimension is reflected by the **number of symmetries**.



- If a nonzero caloric function is **symmetric (invariant)** in all space-time directions, then

$$\partial_t u = 0, \quad |\nabla u| = 0,$$

so $u \equiv c \neq 0$ and hence $Z(u) = \emptyset$.

- If a caloric function is **symmetric** in all but one direction, then after rotation

$$\partial_t u = 0, \quad \partial_i u = 0 \quad (1 \leq i \leq n-1).$$

Thus $u = u(x_n)$ and $\Delta u = 0$, so u is linear and therefore $S(u) = \emptyset$.

Maximal symmetry rules out nodal and singular sets.

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Roadmap: detect polynomial behavior and symmetry

Parabolic analogue of the quantitative stratification of Cheeger, Jiang, Naber, and Valtorta.

- 1 Introduction: Structure of nodal and singular sets
- 2 History
- 3 Frequency detects polynomial behavior and symmetry
- 4 Quantitative strata and ε -regularity
- 5 Neck regions organize multiscale polynomial behavior and symmetry
- 6 General setting

Parabolic obstruction: Regularity in time is subtler. Controlling the half-time derivative $\partial_t^{1/2}$ is the main new difficulty. Also, errors must be summable across scales.

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Where does our work fit in the literature?

- In the 1990s, Han and Lin proved Euclidean Hausdorff estimates for $Z(u)$ and obtained non-optimal dimensional bounds for $S(u)$.
- There is an extensive work in both the elliptic and parabolic settings: Donnelly and Fefferman 90s, Logunov '18, Logunov and Malinnikova 20s, and many others.
- Cheeger, Naber, Valtorta (2015) proved optimal-dimensional Minkowski estimates and rectifiability for critical sets in the elliptic setting.
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- Related ideas from **quantitative stratification** and **neck decomposition** have been used in parabolic problems such as mean curvature flow, harmonic map heat flow, and Ricci flow, as well as in elliptic problems such as harmonic maps, minimal surfaces, and spaces with lower Ricci curvature bounds.
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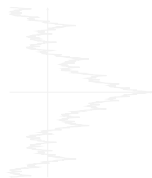
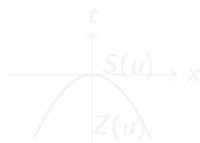
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The parabolic setting is different due to anisotropy in time

Natural space-time geometry: The parabolic metric $|(x, t)|_{\mathcal{P}}^2 = \max\{|x|^2, |t|\}$ is adapted to the scaling $(x, t) \mapsto (\lambda x, \lambda^2 t)$. **In this geometry, time counts like two spatial dimensions.**

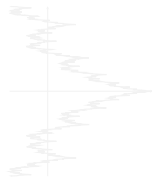
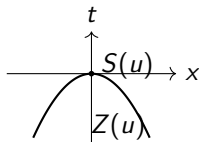


A new regularity issue. Parabolic Lipschitz regularity gives only 1/2-Hölder control in time. The stronger notion of a **regular parabolic Lipschitz graph** requires $\partial_t^{1/2} u \in \text{BMO}$, which is the **natural class in parabolic boundary value theory**. *This stronger class has no elliptic analogue and does not appear in earlier parabolic treatments of nodal and singular sets that we are aware of.*

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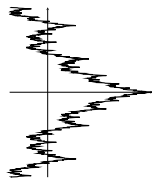
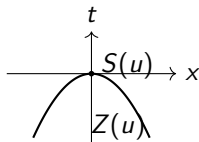


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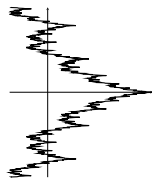
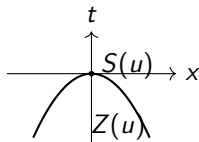


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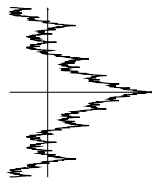
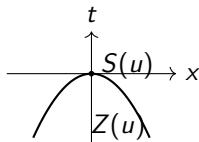


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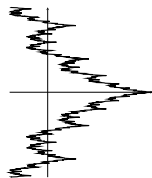
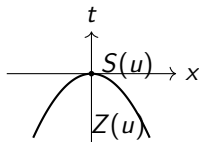


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The parabolic setting is different due to anisotropy in time

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Frequency, a parabolic analog of degree, is monotone

Fix $\mathbf{x}_0 := (x_0, t_0)$ and write $\tau = t_0 - t > 0$. Define

$$d\nu_{\mathbf{x}_0;t}(x) = \frac{1}{(4\pi\tau)^{n/2}} e^{-f_{\mathbf{x}_0}} dx, \quad f_{\mathbf{x}_0}(x, t) := \frac{|x - x_0|^2}{4\tau}.$$

Poon's frequency is

$$N_{\mathbf{x}_0}(\tau) = \frac{2\tau \int_{\mathbb{R}^n} |\nabla u|^2 d\nu_{\mathbf{x}_0;t}}{\int_{\mathbb{R}^n} u^2 d\nu_{\mathbf{x}_0;t}} =: \frac{E_{\mathbf{x}_0}(\tau)}{H_{\mathbf{x}_0}(\tau)}.$$

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where $\Delta_{\phi} := \Delta - \nabla\phi \cdot \nabla$ and $\square u := \partial_t u - \Delta u$.

In the caloric case, $\square u = 0$, so $N_{\mathbf{x}_0}(\tau)$ is monotone.

Frequency detects relative scale.

For the rest of the talk, we zoom in on the caloric setting.

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Almost constant N implies almost homogeneity

N_{x_0} is constant if and only if $2\tau\Delta_{f_{x_0}}u + N_{x_0}(\tau)u = 0$, which forces u to be an m -homogeneous caloric polynomial p with $N_{x_0}(\tau) \equiv m \in \mathbb{Z}_{\geq 0}$, i.e.,

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Theorem (Quantitative uniqueness of tangent flow in the caloric setting)

For any $0 < \tau < \tau_1/2 < \tau_2/4$, if $N_{x_0}(\tau_2) - N_{x_0}(\tau_1) < 1/10$, then there is $m \in \mathbb{Z}_{\geq 0}$ such that

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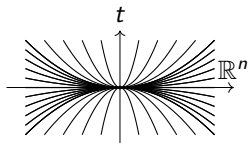
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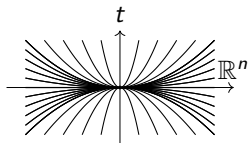
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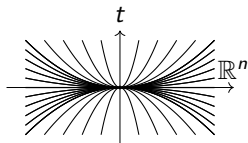
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If N is constant at $\mathbf{x}_1 = (x_1, t_1)$ and $\mathbf{x}_2 = (x_2, t_2)$, then $2\tau_i \Delta_{f_i} u + mu = 0$ for $i \in \{1, 2\}$, for some $m \in \mathbb{Z}_{\geq 0}$, where $\tau_i = t_i - t$ and $f_i = f_{x_i}$. We compute

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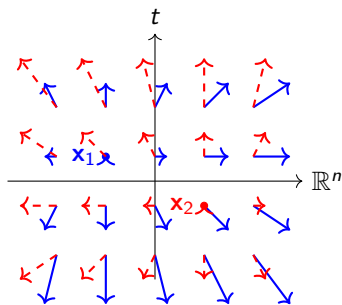


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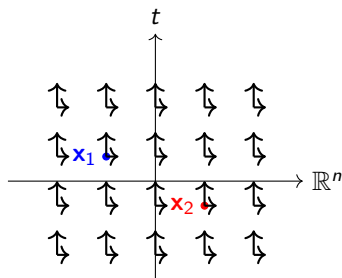
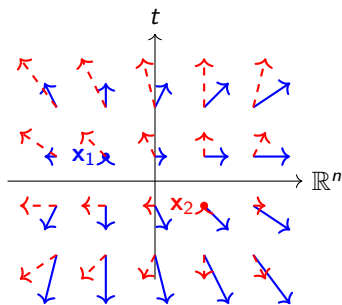
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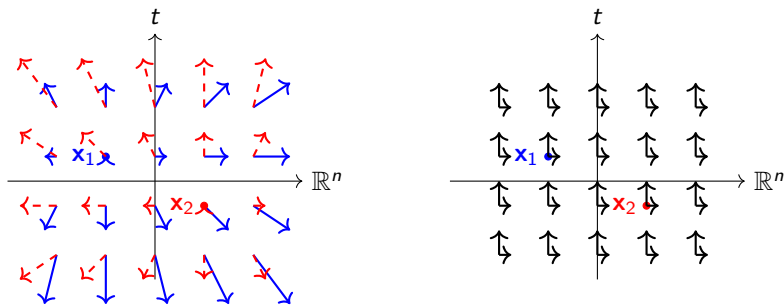


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Characteristic vector fields based at two points generate space–time directions.

Homogeneity at many points forces space-time symmetry

An L^2 version of the previous argument yields the following.

Theorem (Cone splitting inequality)

Let u be a caloric function with $N_{\mathbf{x}_0}(10^5 r^2) \leq \Lambda$. If $\{\mathbf{x}_i\}_{i=0}^K \subset P(\mathbf{x}_0, \frac{1}{10}r)$ is $(k, \alpha r)$ -independent and $K = k$, then

$$r^2 \int_{\mathbb{R}^n} |\pi_L \nabla u|^2 d\nu_{\mathbf{x}_0; t_0 - r^2} \leq C(\alpha, \Lambda) \max_i \{N_{\mathbf{x}_i}(8r^2) - N_{\mathbf{x}_i}(r^2/8)\} \int_{\mathbb{R}^n} u^2 d\nu_{\mathbf{x}_0; t_0 - r^2},$$

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Homogeneity at many well-separated points forces higher space-time symmetry.

Frequency detects both homogeneous polynomial behavior and space-time symmetry.

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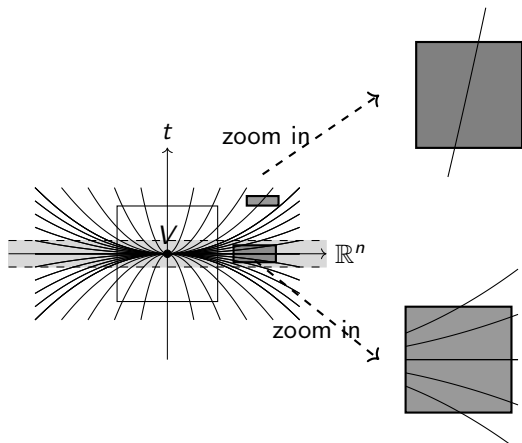
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Extra symmetry emerges away from a plane of symmetry

Theorem (Quantitative dimension reduction)

If u is almost symmetric with respect to a k -plane V , then away from V it is almost symmetric in $k + 1$ directions, but only at a smaller scale.



Quantitative strata and effective nodal/singular sets

All points are created symmetric, but some are more symmetric than others.

Definition (Quantitative strata)

A point lies in $\mathcal{S}_{\varepsilon,r}^k$ if it has *at most* k effective symmetries at every intermediate scale $s \in [r, 1]$.

Definition (Effective nodal and singular sets)

For $r > 0$, define

$$Z_r(u) := \left\{ \mathbf{x} \in P(\mathbf{0}, 1) : \inf_{P(\mathbf{x}, \frac{s}{16})} |u|^2 \leq \frac{1}{8} \int_{\mathbb{R}^n} |u|^2 d\nu_{\mathbf{x}; t-s^2}, \forall s \in [r, 1], \right.$$

$$\left. S_r(u) := \left\{ \mathbf{x} \in P(\mathbf{0}, 1) : \inf_{P(\mathbf{x}, \frac{s}{16})} (|u|^2 + 2s^2 |\nabla u|^2) \leq \frac{1}{16} \int_{\mathbb{R}^n} |u|^2 d\nu_{\mathbf{x}; t-s^2}, \forall s \in [r, 1]. \right. \right.$$

$$Z(u) = \bigcap_{r>0} Z_r(u), \quad S(u) = \bigcap_{r>0} S_r(u).$$

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Quantitative strata and effective nodal/singular sets

All points are created symmetric, but some are more symmetric than others.

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Theorem (ε -regularity)

$$Z_r(u) \subset \mathcal{S}_{\varepsilon,r}^{n+1} \quad \text{and} \quad S_r(u) \subset \mathcal{S}_{\varepsilon,r}^n.$$

The proof is an L^2 version of the rigidity argument from earlier.

- If a nonzero caloric function is symmetric in all space-time directions, then $\partial_t u = 0$, $\nabla u = 0$, so $u \equiv c \neq 0$ and hence $Z(u) = \emptyset$.
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A telescoping estimate guarantees homogeneity on most microscopic scales

Where does symmetry come from?

Assuming $N_x(10^5) < \Lambda$, the telescoping bound

$$\sum_i \left(N_x(8^{-i}) - N_x(8^{-i-1}) \right) = N_x(1) - N_x(0) < \Lambda$$

and monotonicity of N imply that the frequency is almost constant at every point on all but a bounded number of scales.

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Neck regions organize multiscale polynomial symmetry

A **k -neck region** is, informally, a region where across *many scales* $\mathcal{S}_{\epsilon,r}^k \subset \mathcal{C}$ is quantitatively trapped near a **single k -plane**.

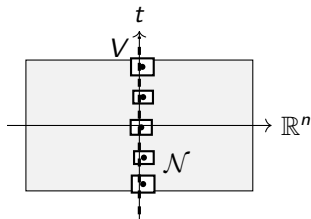
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Strong neck = neck + two upgrades: V is covered by balls centered on \mathcal{C} , and r_\bullet is parabolically δ -Lipschitz.

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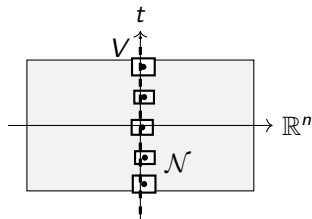
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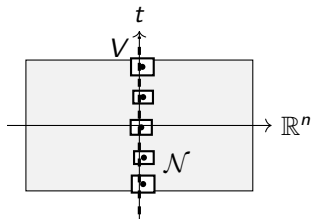
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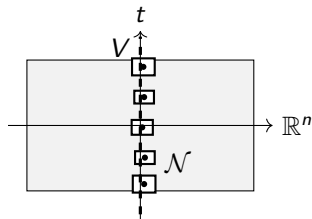
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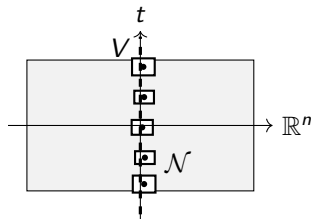
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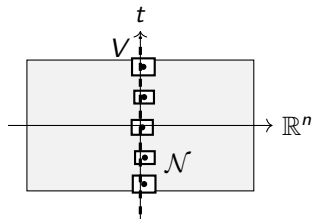
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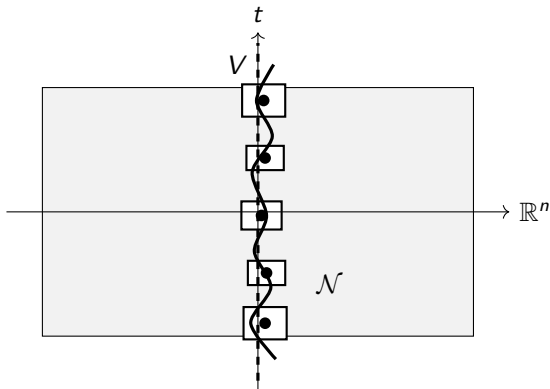
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Neck structure theorem I

Neck structure theorem: multiscale trapping near the plane V implies packing content estimate for \mathcal{C} and for $\mathcal{S}_{\epsilon, r}^k$.



$$P(x_0, 2r) = B(x_0, 2r) \times (t_0 \pm (2r)^2)$$

Regular parabolic Lipschitz graphs

Fix a vertical k -plane V and write a graph $\Gamma = \{y + G(y) : y \in V\}$ with $G : V \rightarrow V^\perp$.

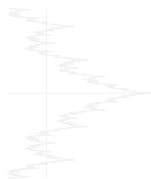
Parabolic Lipschitz: $|G(y) - G(z)| \leq \delta |y - z|_p$ ($y, z \in V$).

Half-time derivative: define for compactly supported φ

$$\partial_t^{1/2} \varphi(x, t) := c \int_{\mathbb{R}} \frac{\varphi(x, s) - \varphi(x, t)}{|s - t|^{3/2}} ds,$$

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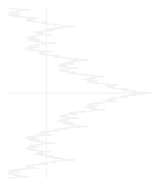
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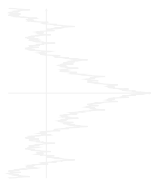
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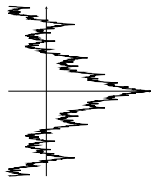
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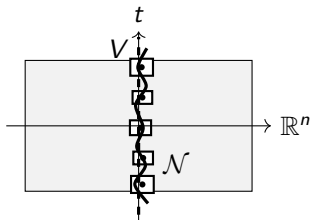
Strong multiscale trapping forces regular graphicality

Theorem (Strong neck structure theorem)

Assume $k \in \{n, n+1\}$ and $\mathcal{N} = P(\mathbf{x}_0, 2r) \setminus P(\mathcal{C}, r_\bullet)$ is a **strong** (m, k, δ, η) -neck region modeled on a vertical plane V . Then there exists a compactly supported $C \delta^{1/(k+2)}$ -regular parabolic Lipschitz map $G : V \rightarrow V^\perp$ such that

$$d_P(\mathbf{x}, \text{graph}(G)) \leq C(\eta, \Lambda) \delta r_x \quad \text{for all } \mathbf{x} \in \mathcal{C} \cap P(\mathbf{x}_0, \frac{4}{3}r).$$

Multiscale trapping near the *vertical* plane V upgrades to a single regular parabolic $\text{graph}(G)$ approximating \mathcal{C} .



Carleson estimate for κ -number implies BMO

$$\beta_{\mathcal{P},k}^2(\mathbf{x}, r) := \inf_{V \in \text{Aff}_{\mathcal{P}(k)}} \frac{1}{r^k} \int_{P(\mathbf{x}, r)} \left(\frac{d_{\mathcal{P}}(\mathbf{y}, V)}{r} \right)^2 d\mu(\mathbf{y}).$$

In elliptic Reifenberg theory, Carleson control of β -numbers with respect to the packing measure μ of \mathcal{C}

$$\int_{B(\mathbf{x}, 10r)} \int_{\frac{1}{8}r_y}^r \beta_k^2(y, 10^5 s) \frac{ds}{s} d\mu(y) \leq C\delta r^k.$$

yields a **Euclidean Lipschitz parameterization** of \mathcal{C} .

We need a *parabolic* Lipschitz graph with $\partial_t^{1/2}$ -BMO control. Upgrade β -control to a **Carleson estimate for a stronger “ κ -number”** that implies BMO for $\partial_t^{1/2}$.

$$\kappa^2(\mathbf{x}, r) := \kappa_k^2(\mathbf{x}, r; G) := \inf_{\ell} \frac{1}{r^k} \int_{P(\mathbf{x}, r) \cap V} \left(\frac{|G(\mathbf{y}) - \ell(\mathbf{y})|}{r} \right)^2 d\mathcal{H}_P^k(\mathbf{y}),$$

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where inf is over all affine functions $\ell: V \rightarrow \mathbb{R}^{n+2-k}$ satisfying $\partial_t \ell \equiv 0$.

Carleson estimate for κ -number implies BMO

$$\beta_{\mathcal{P},k}^2(\mathbf{x}, r) := \inf_{V \in \text{Aff}_{\mathcal{P}(k)}} \frac{1}{r^k} \int_{P(\mathbf{x}, r)} \left(\frac{d_{\mathcal{P}}(\mathbf{y}, V)}{r} \right)^2 d\mu(\mathbf{y}).$$

In **elliptic Reifenberg theory**, **Carleson control of β -numbers** with respect to the packing measure μ of \mathcal{C}

$$\int_{B(\mathbf{x}, 10r)} \int_{\frac{1}{8}r_{\mathbf{y}}}^r \beta_k^2(\mathbf{y}, 10^5 s) \frac{ds}{s} d\mu(\mathbf{y}) \leq C\delta r^k.$$

yields a **Euclidean Lipschitz parameterization of \mathcal{C}** .

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Local tangents quantitatively align along V

On a **strong vertical neck**, every $\mathbf{x} \in \mathcal{C}$ has a **best vertical k -plane** $V_{\mathbf{x}}(s)$ at each scale $s \in [r_{\mathbf{x}}, \gamma^{-3}r]$ (from (k, δ, s) -symmetry).

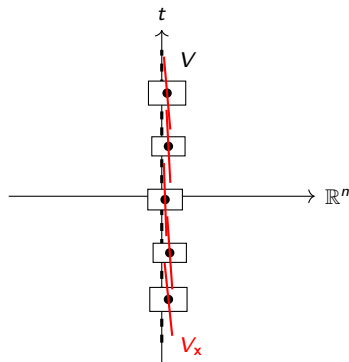
Step 1: $V_{\mathbf{x}}(s)$ is a good tangent approximation: $\mathcal{C} \cap P(\mathbf{x}, s) \subset P(\mathbf{x} + V_{\mathbf{x}}(s), \delta s)$ and $(\mathbf{x} + V_{\mathbf{x}}(s)) \cap P(\mathbf{x}, s) \subset P(\mathcal{C}, \gamma s)$.

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$$\angle(V_{\mathbf{x}}(s), V_{\mathbf{x}}(2s)) \lesssim \beta(\mathbf{x}, s),$$

Step 3: All $V_{\mathbf{x}}(s)$ stay close to a single vertical plane V on \mathcal{N} .

Local tangents align along one global plane V .



Warning for fig: All $V_{\mathbf{x}}$ are vertical. Tilting is happening in space.

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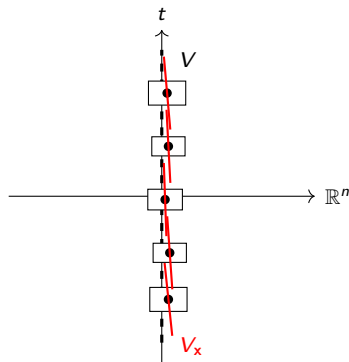
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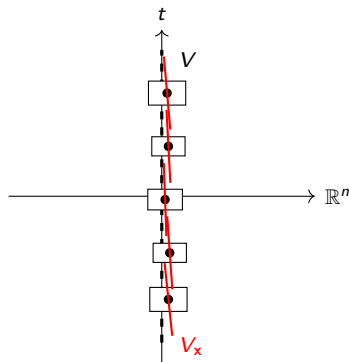
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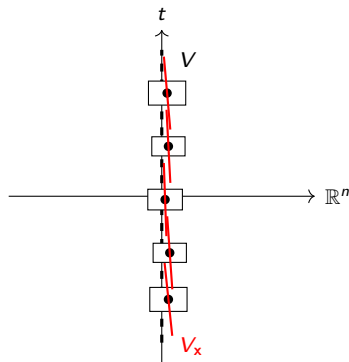
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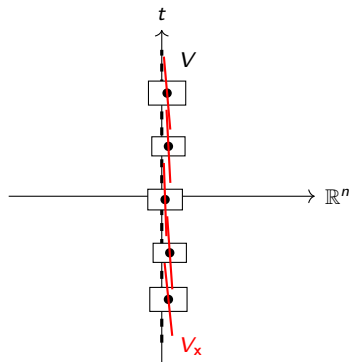
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Local tangents glue to form a graph with Carleson estimate

Whitney decomposition. Cover $V \cap P(\mathbf{0}, 1)$ by parabolic balls $\{P(\mathbf{y}, r_{\mathbf{y}})\}$ where \mathbf{y} is \mathcal{C}_+ projected to V .

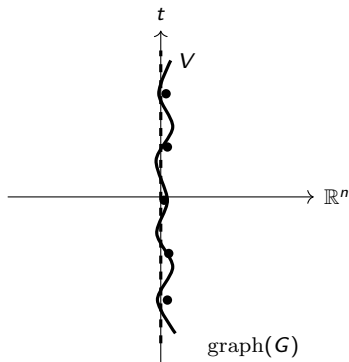
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Glue with a partition of unity. With $\{\phi_{\mathbf{y}}\}$ subordinate to $\{P(\mathbf{y}, r_{\mathbf{y}})\}$.

$$G := \sum_{\mathbf{y}} \phi_{\mathbf{y}} G_{\mathbf{y}}.$$

Parabolic refinement. A Carleson estimate for κ yields $\|\partial_t^{1/2} G\|_{\text{BMO}_P} \lesssim (\text{small})$, so G is regular parabolic Lipschitz.

For $k \in \{n, n+1\}$, $S_{\epsilon, r}^k$ on a strong neck region is contained in a regular parabolic Lipschitz graph.



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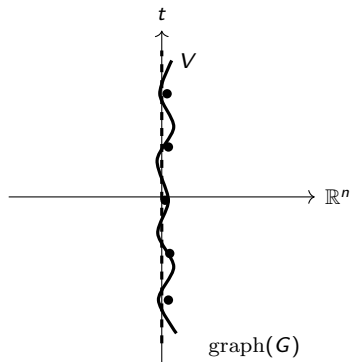
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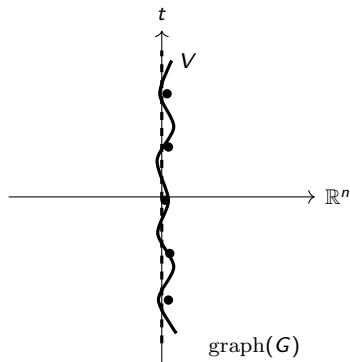
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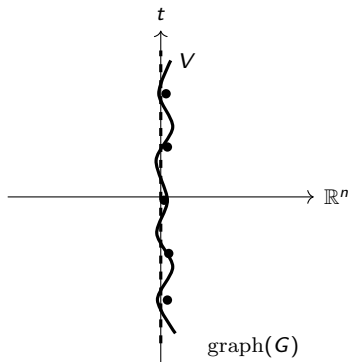
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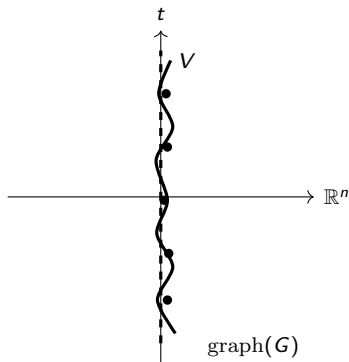
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Organizing points by multiscale polynomial symmetry

We decompose points according to whether, across many scales, they exhibit k symmetries or an additional $(k + 1)$ th symmetry.

Let $k \in \{n, n + 1\}$ and assume $N_0(\gamma^{-100}) \leq \Lambda$. For any $\varepsilon \in (0, 1]$ and $\eta \leq \eta(\Lambda)$, there exists a decomposition

$$P(0, 1) \subseteq \bigcup_a \mathcal{N}_a \cup \bigcup_b P(x_b, r_b) \cup \left(\tilde{\mathcal{C}} \cup \bigcup_a \mathcal{C}_{a,0} \right).$$

Each \mathcal{N}_a is a string neck region.

For each ball $B(x_b, r_b)$, the function u is almost $(k + 1)$ -symmetric at some point $x_{b,0} \in P(x_b, r_b)$.

It follows that

$$u|_{B(x_b, r_b)} \in \mathcal{C}_{k+1}(\varepsilon, \eta)$$

with the error that

$$\|u\|_{C^k(B(x_b, r_b))} \leq \varepsilon$$

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- If $k = n + 1$, then

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Finite-resolution neck decomposition

Fix a resolution $r_* \in (0, 1]$. Given $\epsilon, \eta, r_*, \Lambda$, there is a cover of the unit parabolic ball:

$$P(\mathbf{0}, 1) \subset \bigcup_a \mathcal{N}_a \cup \bigcup_b P(\mathbf{x}_b, r_b) \cup \bigcup_f P(\mathbf{x}_f, r_f).$$

- $\mathcal{N}_a = P(\mathbf{x}_a, 2r_a) \setminus P(\mathcal{C}_a, r_\bullet)$ is a neck region, with $r_\bullet \geq r_*$.
- u is almost $k+1$ -symmetric at some $\mathbf{y}_b \in P(\mathbf{x}_b, r_b)$.
- $r_a, r_b \geq r_*$ and $r_f \in [r_*, C(\Lambda, \epsilon, \eta) r_*]$ so all residual pieces are commensurable to r_* .
- $\sum_a r_a^k + \sum_b r_b^k + \sum_f r_f^k \leq C(\Lambda) (\epsilon \eta)^{-C(\Lambda)}$.

Why “finite resolution”? Unlike earlier neck decompositions, the residual cannot be thrown away by approximating u with a nicer solution: we must keep a leftover region at scale r_* to obtain Minkowski bounds at resolution r_* .

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Main result: regular parabolic Lipschitz graphicality

When $k = n + 1$ set $\mathcal{C}^k(u) = Z(u)$ and when $k = n$ set $\mathcal{C}^k(u) = S(u)$.

Most of $\mathcal{C}^k(u)$ is organized along **regular parabolic Lipschitz graphs** over vertical affine subspaces of dimension k .

Theorem

Let u be a caloric function with $N(10^5) \leq \Lambda$. For any $k \in \{n, n + 1\}$ and $\epsilon > 0$, there exist $\mathbf{x}_i \in P(\mathbf{0}, \frac{1}{4})$, vertical planes $V_i \in \text{Aff}_{\mathcal{P}}(k)$, $r_i \in (0, 1]$, and ϵ -regular parabolic Lipschitz functions $F_i: V_i \rightarrow V_i^\perp$ such that $\sum_i r_i^k \leq C(\epsilon, \Lambda)$, and such that $G_i := \{\mathbf{x} + F_i(\mathbf{x}): \mathbf{x} \in V_i \cap P(\mathbf{x}_i, 10r_i)\}$ satisfy

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Main result: Codimension 1 and 2 minkowski estimate

Theorem

Let $k \in \{n, n+1\}$ and let u be a caloric function with $N(10^5) \leq \Lambda$. Then, for any $r \in (0, 1]$,

$$\mathcal{H}_P^{n+2}(P(\mathcal{C}_r^k(u), r) \cap P(\mathbf{0}, 1)) \leq C\Lambda^4 r^{n+2-k}.$$

Volume say \mathcal{C}^k is k dimensional. We expect the dependence on Λ is not optimal. It is polynomial on Λ in the elliptic setting.

Corollary

- 1 For $\mathcal{H}^{\frac{1}{2}}$ -almost every $t \in (-1, 1)$, $\mathcal{C}_t^k(u) := \mathcal{C}^k(u) \cap (B(0^n, 1) \times \{t\})$ satisfies $\mathcal{H}^{k-1}(\mathcal{C}_t^k(u)) = 0$.
- 2 For \mathcal{H}^1 -almost every $t \in (-1, 1)$, $\mathcal{C}_t^k(u)$ is $(k-2)$ -rectifiable, and $\mathcal{H}^{k-2}(\mathcal{C}_t^k(u)) \leq C\Lambda^4$.

Estimate for S_t can't be improved for all time slices since $S_0(u) = Z_0(u) = \{(0, 0)\}$ when $u = x^2 + 2t$.

Main result: Codimension 1 and 2 minkowski estimate

Theorem

Let $k \in \{n, n+1\}$ and let u be a caloric function with $N(10^5) \leq \Lambda$. Then, for any $r \in (0, 1]$,

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Volume say \mathcal{C}^k is k dimensional. We expect the dependence on Λ is not optimal. It is polynomial on Λ in the elliptic setting.

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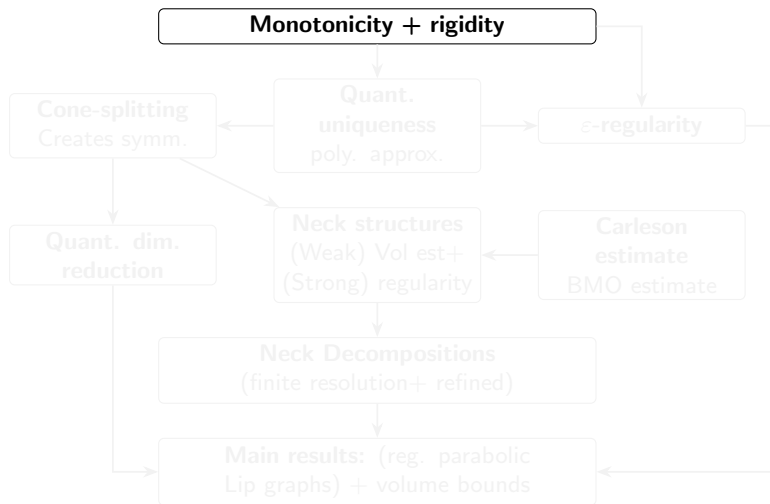
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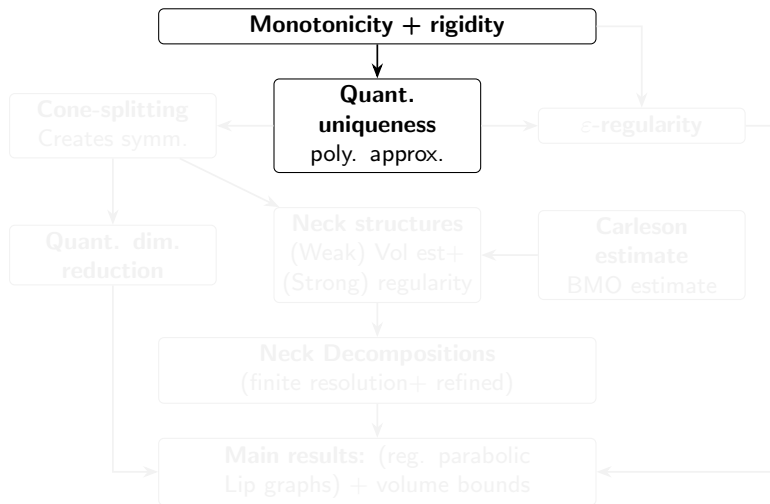
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Recap of the argument in the caloric setting



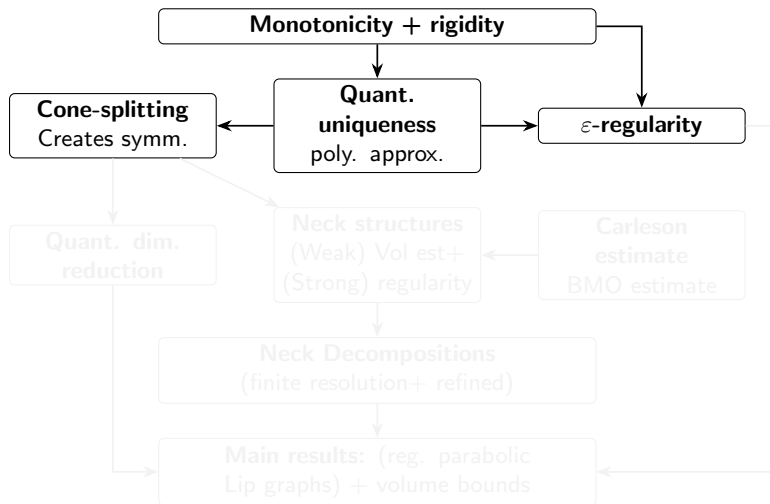
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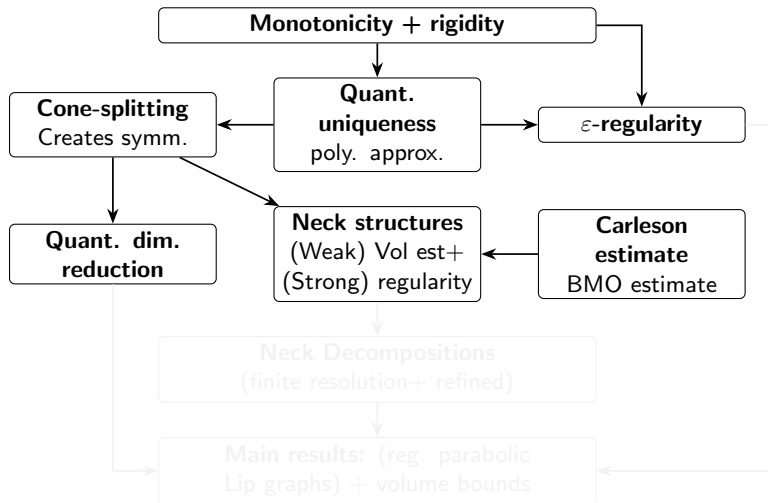
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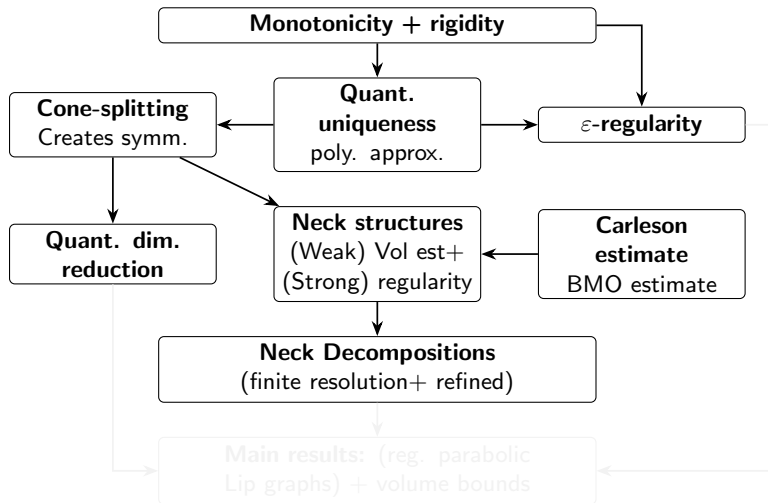
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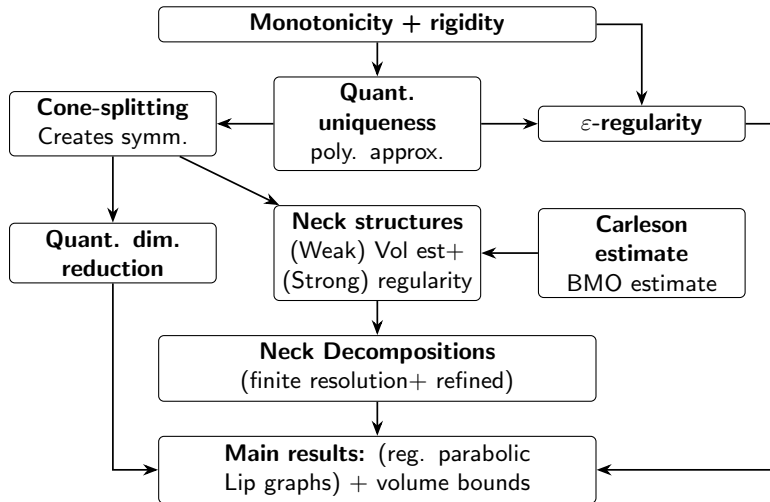
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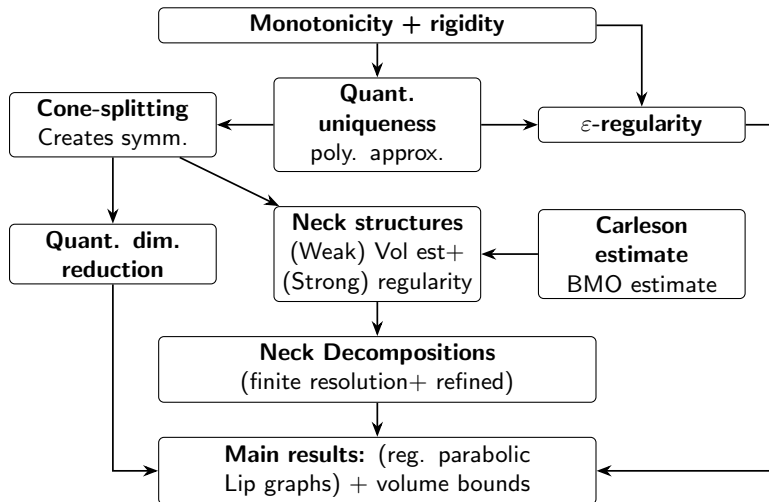
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General setup: errors summable across scales

Recall, general setting of parabolic inequality with Lipschitz coefficients (on a parabolic ball)+ doubling coefficients.

For global functions, we have

$$N'_{x_0}(\tau) = \frac{4\tau}{H_{x_0}(\tau)} \int_{\mathbb{R}^n} \left(\left(\Delta_{f_{x_0}} u + \frac{N_{x_0}(\tau)}{2\tau} u \right)^2 + \left(\Delta_{f_{x_0}} u + \frac{N_{x_0}(\tau)}{2\tau} u \right) \square u \right) dv_{x_0;t}.$$

The second term is a scale/ independent error term so that it is a combination of cubic terms, it is always nonnegative. Another source of error comes from using a cutoff because we need global functions.

For the first term, we will use the doubling property of the measure.

□

For rest of the talk, the geometric side survives, but we will ignore the error terms. This is not a good approximation.

For the rest of the talk, we will assume that the measure is doubling.

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- The **second term is a scale/ M -dependent error** term so that if u is a perturbation of caloric function, N is **almost monotone**. Another source of error comes from using a **cutoff** because u is not global.
- Quantitative uniqueness is not straightforward. Need to do **stability analysis** of dynamical rescaling.
- For rest of the tools, the **geometric step survives**, but we must absorb PDE errors. Caloric approximation saves us.

Fixes Use doubling condition to get **polynomial growth of solutions** u and the **exponential tail of Gaussian** to control error terms so that they are **summable across scales for Carleson estimate**. And it works!!!

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Thank you!