

Focus-Glue-Context Fisheye Transformations for Spatial Visualization

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Abstract Fisheye views magnify local detail while preserving context, yet projection-aware, scriptable tools for R spatial analysis remain limited. `mapycusmaximus` introduces a Focus-Glue-Context (FGC) fisheye transform for numeric coordinates and `sf` geometries. Acting radially around a chosen center, the transform defines a magnified focus (`r_in`), a smooth transitional glue zone (`r_out`), and a fixed exterior. Distances expand or compress via a zoom factor and a power-law squeeze, with an optional angular twist that enhances continuity. The method is projection-conscious: lon/lat inputs are reprojected to suitable CRSs (e.g., GDA2020/MGA55), normalized for stable parameter control, and restored afterward. A geometry-safe engine (`st_transform_custom`) supports all feature types, maintaining ring closure and metadata. The high-level `sf_fisheye()` integrates with tidyverse, ggplot2, and Shiny, with built-in datasets and tests ensuring reproducibility. By coupling coherent radial warps with tidy, CRS-aware workflows, `mapycusmaximus` enables spatial exploration that emphasizes local structure without losing global context.

1 Introduction

Maps that reveal fine local structure without losing broader context face a persistent challenge: zooming in hides regional patterns, while small-scale views suppress local detail. Traditional solutions—insets, multi-panel displays, aggressive generalization—break spatial continuity and increase cognitive load (Cockburn et al., 2008). What if we could smoothly magnify a metropolitan core *while keeping it embedded* in its state-level context?

This package implements a Focus-Glue-Context (FGC) fisheye transformation that continuously warps geographic space. The transformation magnifies a chosen focus region, compresses surrounding areas into a transitional glue zone, and maintains stability in the outer context. The approach operates directly on vector geometry coordinates, preserves topology, and supports reproducible, pipeline-oriented cartography within the R `sf` and `ggplot2` ecosystem. An optional glue-zone twist (the `revolution` parameter) can gently rotate features to aid continuity; in this paper’s figures we set `revolution = 0`.

The intellectual lineage of focus+context visualization traces back to Furnas (1986)’s *degree-of-interest* function, which introduced a formal method to rank information elements by combining intrinsic importance with distance from the user’s focus. In this model, items with low DOI are deemphasized or hidden, enabling emphasis on salient regions without losing global structure. Sarkar and Brown (1992) and Sarkar and Brown (1994) extended this to geometric distortion, demonstrating smooth magnification transitions for graph visualization. Subsequent innovations explored diverse lenses: hyperbolic geometry for hierarchies (Lamping et al., 1995), distortion-view frameworks (Carpendale and Montagnese, 2001), and “magic lens” overlays (Bier et al., 1993). By 2008, Cockburn et al. (2008)’s comprehensive review synthesized two decades of research across overview+detail, zooming, and focus+context paradigms.

In cartography, the need for nonlinear magnification emerged independently. Snyder (1987) developed “magnifying-glass” azimuthal projections with variable radial scales—mathematical foundations. Harrie et al. (2002) created variable-scale functions for mobile devices where user position appears large-scale against small-scale surroundings. An influential contribution came from Yamamoto et al. (2009) and Yamamoto et al. (2012): their **Focus+Glue+Context model** introduced an intermediate “glue” region that absorbs distortion, preventing the excessively warped roads and boundaries that plagued earlier fisheye maps. This three-zone architecture proved particularly effective for pedestrian navigation and mobile web services.

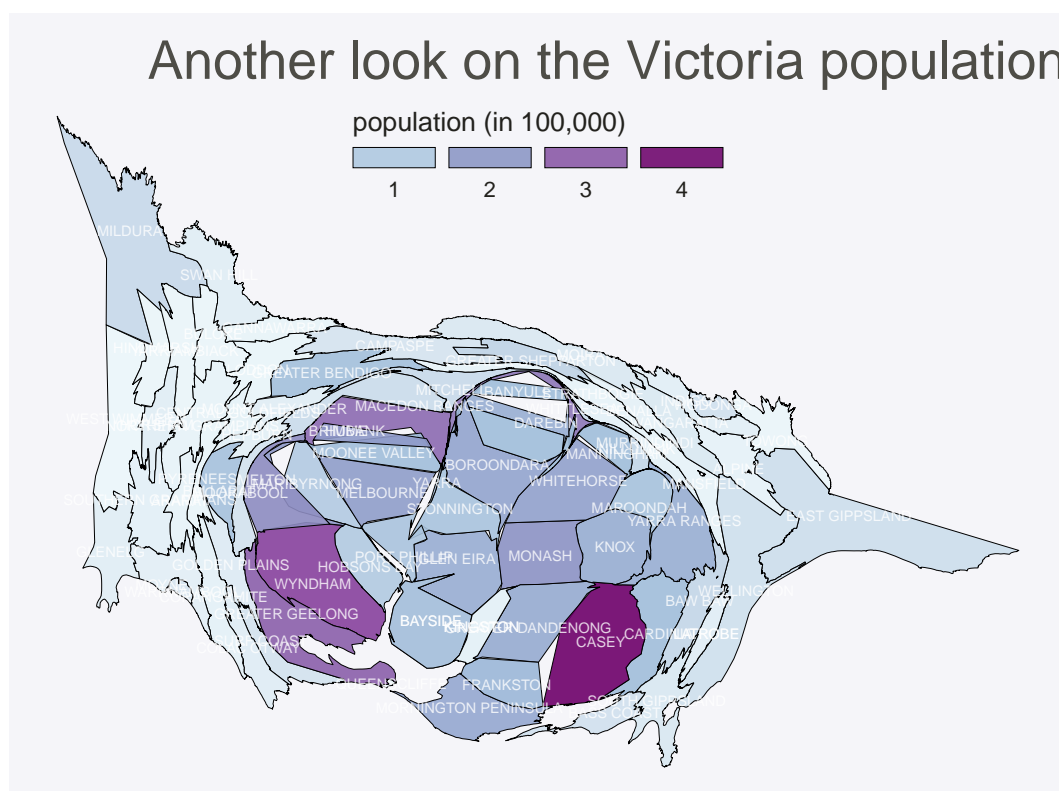


Figure 1: Cartogram example: thematic distortion changes shapes and sizes to encode population.

Within R’s spatial ecosystem, *sf* (Pebesma, 2018) provides robust vector handling and CRS transformations, while *ggplot2* (Wickham, 2016) offers declarative visualization grammar. Yet a gap remained: existing tools addressed *related* distortion needs but not continuous geometric fisheye lenses. This package fills that niche by formalizing an *sf*-native FGC radial model with controllable zone parameters, optional angular effects, automatic normalization, and safe geometry handling across points, lines, and polygons.

2 Background

Before examining the mechanics of fisheye transformations, it is important to review how R’s spatial ecosystem currently addresses the detail-versus-context tradeoff. This context clarifies why existing solutions, though valuable, do not fully address the need for continuous lens-based warping.

Cartograms: Thematic distortion

The cartogram family (Gastner and Newman, 2004) intentionally distorts geographic areas to encode variables—population density reshapes regions so area becomes proportional to demographic weight (see Figure 1).

This approach fundamentally differs from focus+context methods. Cartograms substitute spatial accuracy for data encoding, often severely disrupting shapes and adjacencies. For example, a population cartogram enlarges Monash while shrinking La Trobe, prioritizing thematic insight over geographic fidelity. The reader can still somehow find the shape of Victoria familiars, as the cartogram preserves relative positions and topology, however, its shape and size are distorted. In contrast, the FGC fisheye transformation preserves relative positions and topology while magnifying a user-selected spatial region rather than a data-driven variable. The use cases are distinct: cartograms address the dominance of a variable in space, whereas fisheye lenses facilitate exploration of local detail within a broader

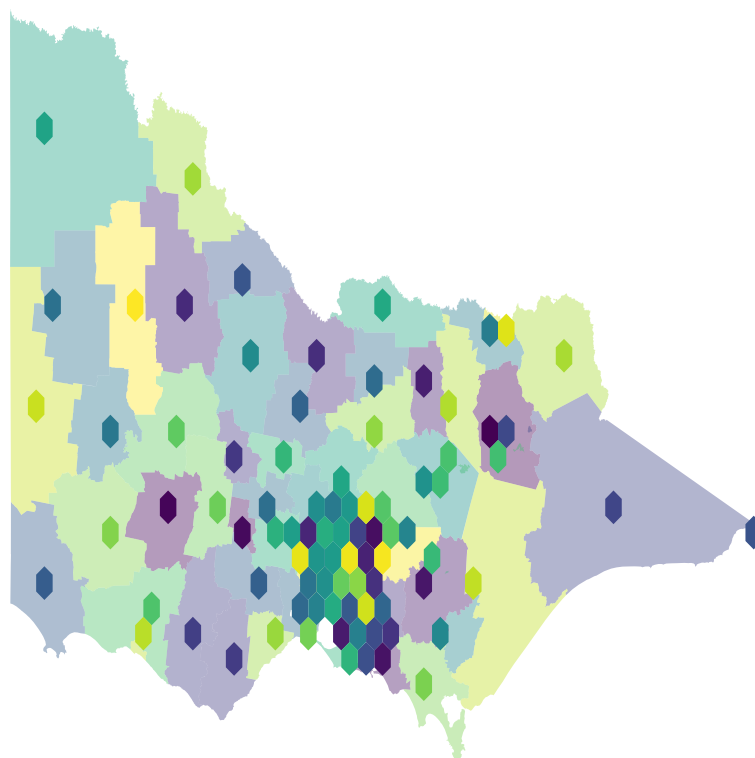


Figure 2: Sugarbag example: thematic distortion changes shapes and sizes to encode population.

geographic context.

Hexagon tile maps: Discrete abstraction

Packages like `sugarbag` replace irregular polygons with regular hexagonal or square tiles, each representing an administrative unit.

As seen in Figure 2, tile maps *abstract away* precise geography entirely, treating space as a topology-preserving tessellation where “neighbors touch” matters more than accurate boundaries. Tile maps excel at avoiding size bias (Mildura gets equal visual weight to Yarra) and creating aesthetic, clutter-free layouts. However, they abandon continuous spatial relationships: you cannot identify precise locations, measure distances, or overlay point data meaningfully. In the example above, the `sugarbag` approach was overlay on top of the original geography. Hexbin aggregation for point data (via `ggplot2::geom_hex()`) serves a different purpose—density estimation—rather than focus+context navigation.

Multi-panel approaches: Spatial separation

Tools like `cowplot::ggdraw()` (Wilke, 2025) create side-by-side views: one panel shows overview, another shows zoomed detail (Figure 3).

These are effective for static reports but require viewers to mentally integrate separate views, and they don’t preserve the *embedded* relationship between focus and context within a single continuous geography. Furthermore, if you introduce one or more elements into the plot like filling value equal to a variable, the audience will have a hard time identify the zoomed detail.

Why FGC fisheye offers something distinct

None of these approaches provide *continuous geometric magnification within a single, topology-preserving map*. Cartograms distort for data, not user-chosen focus. Tile maps abstract away

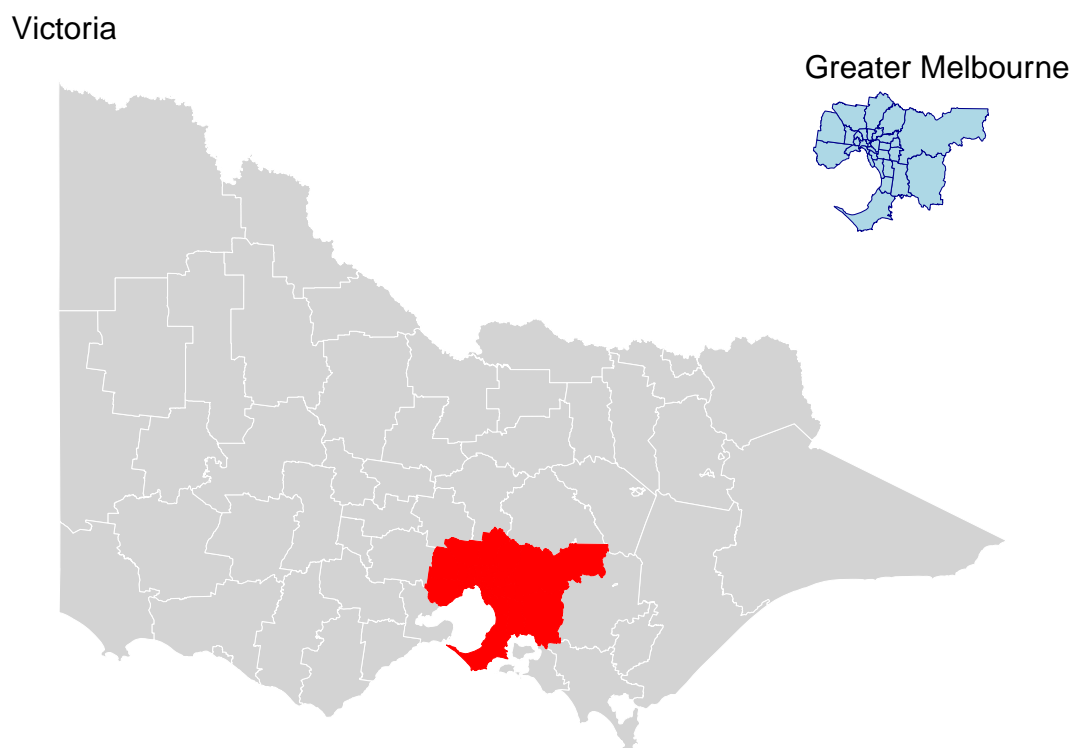


Figure 3: Overview with inset: separates focus from context into distinct panels.

geography. Multi-panel tools spatially separate context. The fisheye lens keeps everything in one frame—roads bend smoothly, metropolitan detail enlarges, but you still see how the city sits within its state. It’s a geometric *warp* rather than a data-driven *substitution* or panel-based *separation*. This matters for use cases like: examining hospital networks in Melbourne while maintaining Victorian context, exploring census tracts in a metro core without losing county boundaries, or analyzing transit lines with their regional hinterland visible.

With this landscape established, we now turn to the technical implementation: how does the FGC transformation actually work, and how does this package make it accessible within R’s spatial workflows?

3 Implementation

```
# Inspect diagnostics returned by fisheye_fgc()
head(transform_df[, c("x_new", "y_new", "zone", "r_orig", "r_new")])

#> # A tibble: 6 x 5
#>   x_new y_new zone    r_orig r_new
#>   <dbl> <dbl> <chr>    <dbl> <dbl>
#> 1  -1    -1    -1 context  1.41  1.41
#> 2 -0.9   -1    -1 context  1.35  1.35
#> 3 -0.8   -1    -1 context  1.28  1.28
#> 4 -0.7   -1    -1 context  1.22  1.22
#> 5 -0.6   -1    -1 context  1.17  1.17
#> 6 -0.5   -1    -1 context  1.12  1.12

table(transform_df$zone)

#>
#> context    focus    glue
```

Fisheye Transformation: Point Movement

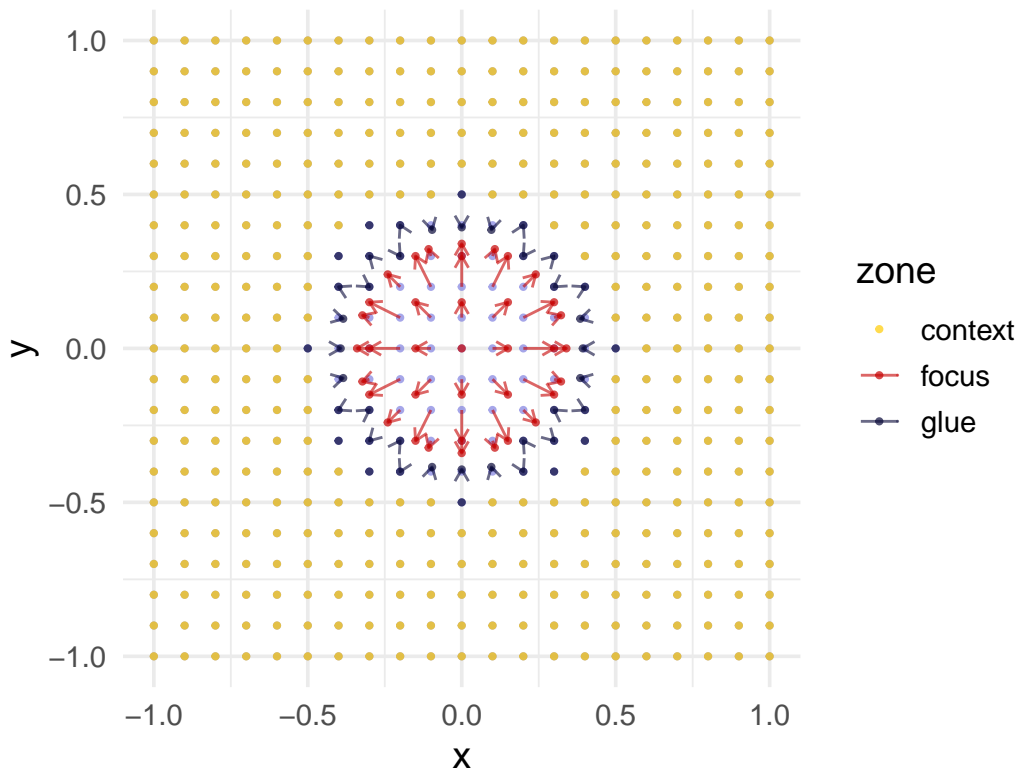


Figure 4: The three zones of an FGC transformation. Points inside the focus (red) expand radially; points in the glue (blue) compress toward the focus boundary; context points (gold) remain fixed.

#> 362 37 42

Consider a point $P = (x, y)$ in a projected coordinate system. The analyst chooses a center $C = (c_x, c_y)$ and two radii: r_{in} delineating the focus region and r_{out} marking the glue boundary. Points inside the focus magnify, points between the radii focus on the center and then compress according to a smooth curve, and points outside remain unchanged. This radial scheme keeps angular coordinates intact, thereby preserving bearings and relative direction.

3.1 Algorithm

Let (r, θ) denote the polar form of point $P = (x, y)$ relative to center $C = (c_x, c_y)$. The transformation defines a new radius r' via a piecewise function:

$$r' = \begin{cases} \min(z \cdot r, r_{\text{in}}) & \text{if } r \leq r_{\text{in}}, \\ r_{\text{in}} + (r_{\text{out}} - r_{\text{in}}) \cdot h(u; s) & \text{if } r_{\text{in}} < r \leq r_{\text{out}}, \\ r & \text{if } r > r_{\text{out}}, \end{cases} \quad (1)$$

where $z > 1$ is the zoom factor within the focus, $s \in (0, 1]$ controls glue compression, and $u = (r - r_{\text{in}}) / (r_{\text{out}} - r_{\text{in}})$ normalises the glue radius to $[0, 1]$. The function $h(u; s)$ is chosen so that $h(0; s) = 0$, $h(1; s) = 1$, and both the first derivatives and the radii match at the boundaries. We adopt a symmetric power curve:

$$h(u; s) = \begin{cases} \frac{1}{2} \cdot u^{1/s} & \text{if } 0 \leq u \leq 0.5, \\ 1 - \frac{1}{2} \cdot (1 - u)^{1/s} & \text{if } 0.5 < u \leq 1, \end{cases} \quad (2)$$

which compresses radii near both boundaries and emphasises the mid-glue region.

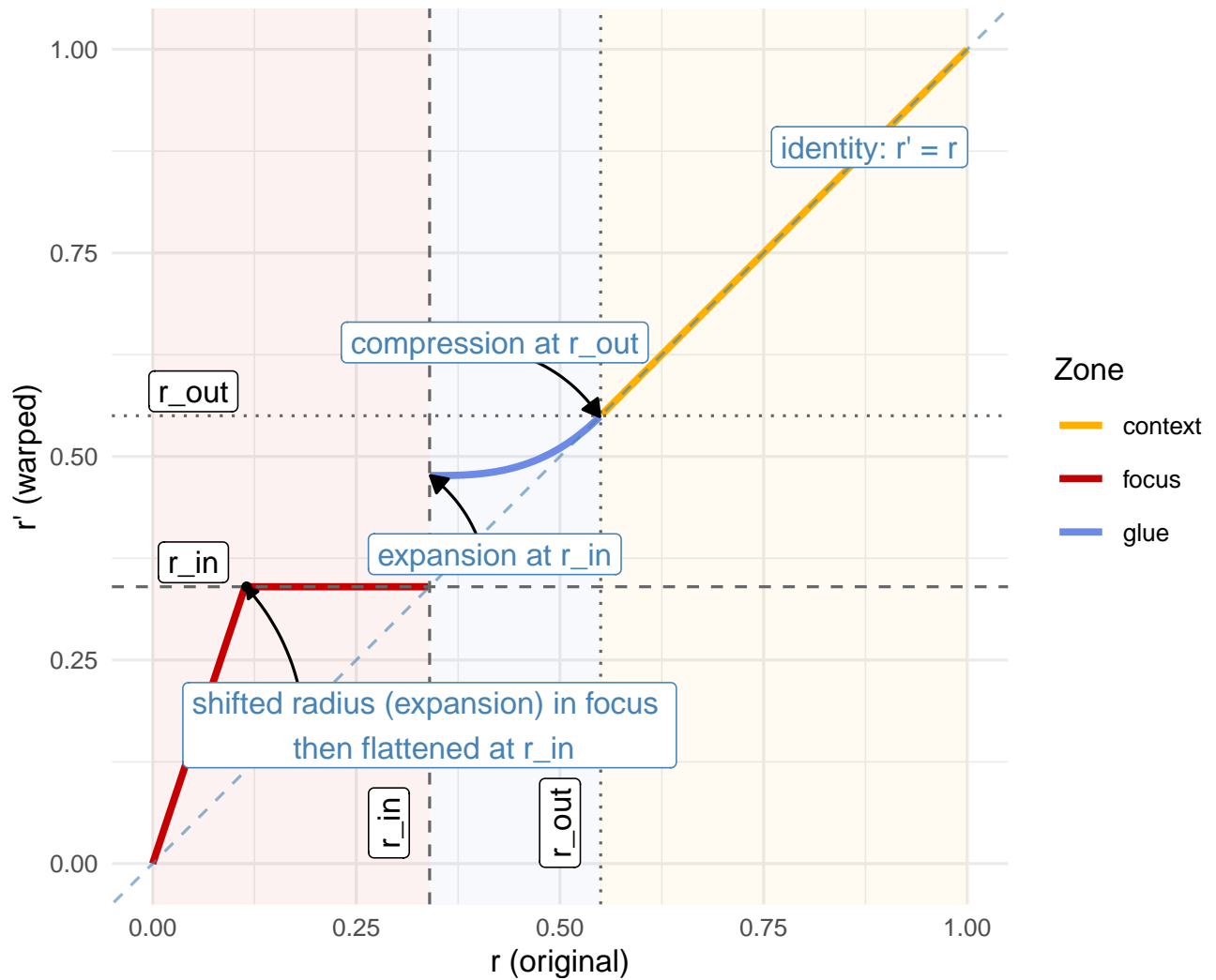


Figure 5: Radial mapping $r \rightarrow r'$ across focus, glue, and context.

Analysts seeking outward compression can choose alternative methods (e.g., the "outward" mode) that bias the curve towards r_{out} . The demonstration on how original and transformed radius can be seen at the Figure 5. The transform optionally introduces rotation within the glue zone to accentuate the flow from detail to context. Let $\phi(u)$ denote the angular adjustment. We employ a bell-shaped profile: $\phi(u) = \rho \cdot 4u(1 - u)$, where ρ is the revolution parameter (in radians). This function peaks at the glue midpoint and vanishes at the boundaries, ensuring continuity.

3.2 Integration with sf

Spatial datasets vary widely in CRS, extent, feature types, and schemas. `mapycusmaximus` follows a disciplined staged workflow where each step is explicit, auditable, and invariant to input type. The architecture separates numeric mapping, spatial orchestration, and geometry reconstruction, allowing the core transform to remain small and testable while `sf`-specific concerns are isolated in thin wrappers.

Workflow and CRS handling

The pipeline proceeds: **sanitize input** \rightarrow **select working CRS** \rightarrow **normalize** \rightarrow **warp** \rightarrow **denormalize** \rightarrow **restore original CRS**. Empty geometries are dropped and `sf::st_zm()` enforces 2D coordinates.

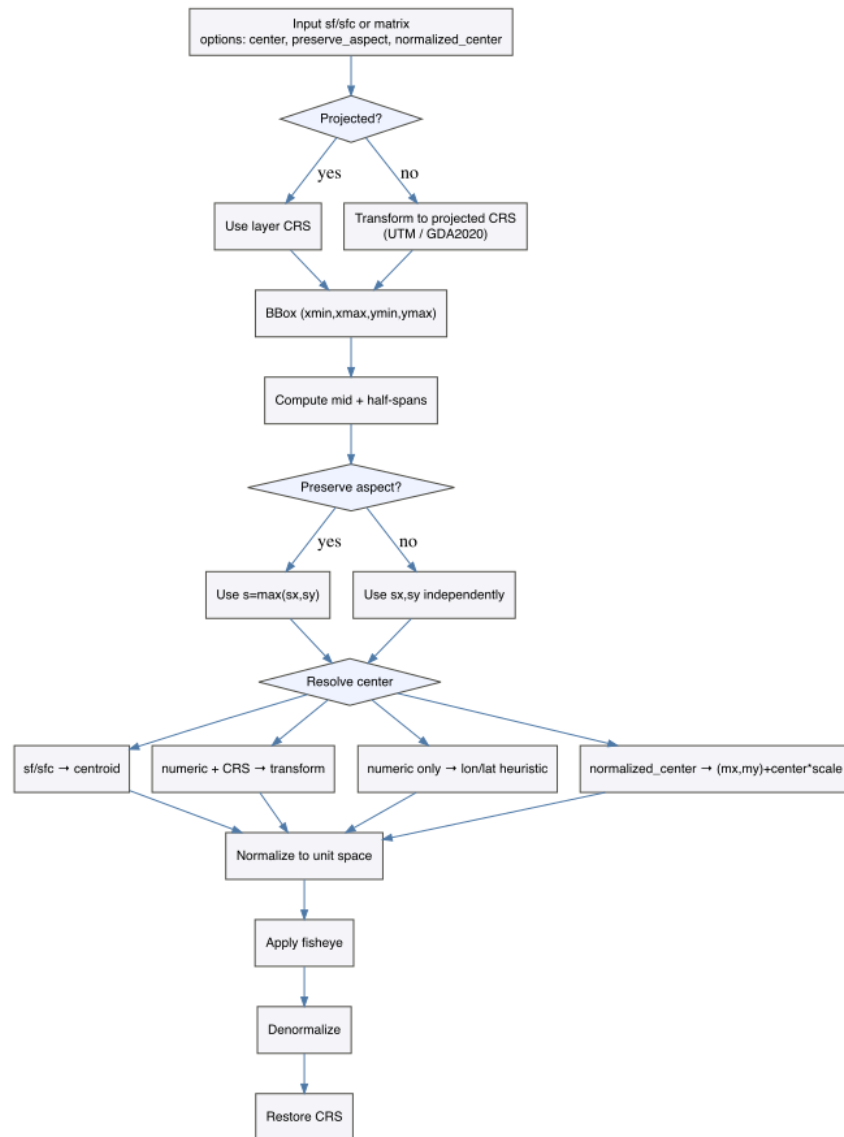


Figure 6: Diagram of the normalization step

CRS selection If the layer is already in a projected CRS, that CRS is used. If it is geographic (lon/lat), the data are transformed to a sensible local projected CRS (e.g., UTM inferred from the centroid; for Victoria, GDA2020/MGA55 is typical). Distances are then in metres and parameters behave consistently. The original CRS is restored on return.

Normalization A bounding box defines the normalizing scale. With `preserve_aspect = TRUE`, a uniform scale $s = \max(s_x, s_y)$ is applied; otherwise axes scale independently. Center resolution happens before normalization: `sf/sfc` centres reduce to a centroid then transform to the working CRS; numeric pairs with `center_crs` are transformed; numeric pairs without CRS are interpreted heuristically; with `normalized_center = TRUE`, pairs live in $[-1, 1]$ relative to the bbox midpoint. If no center is given, the bbox midpoint is used.

Core transformation

At the heart of the package is `fisheye_fgfc()`, a vectorized function mapping an $n \times 2$ coordinate matrix to a new $n \times 2$ matrix via the FGC rule. Its contract is minimal: numeric arrays and scalar parameters defining center, radii, magnification, compression, method, and revolution. Internally it converts to polar form, applies the piecewise radial map

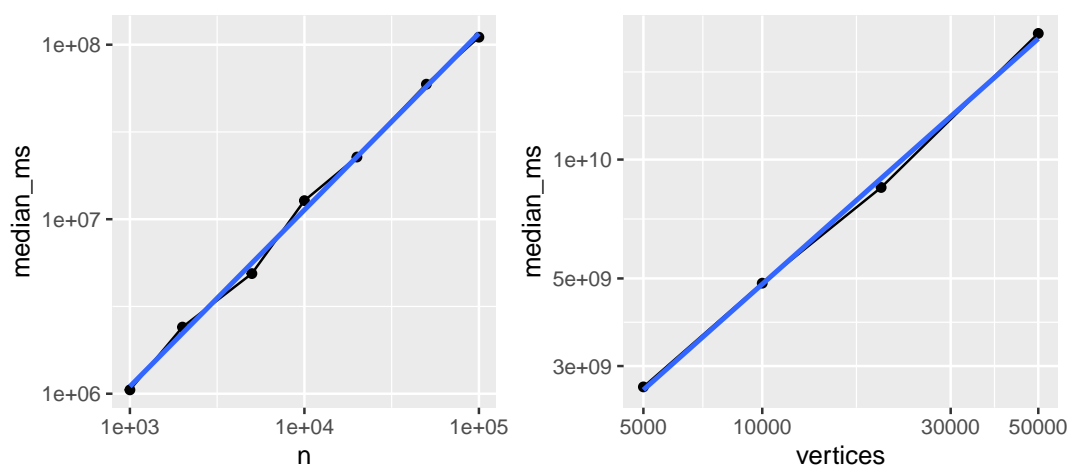
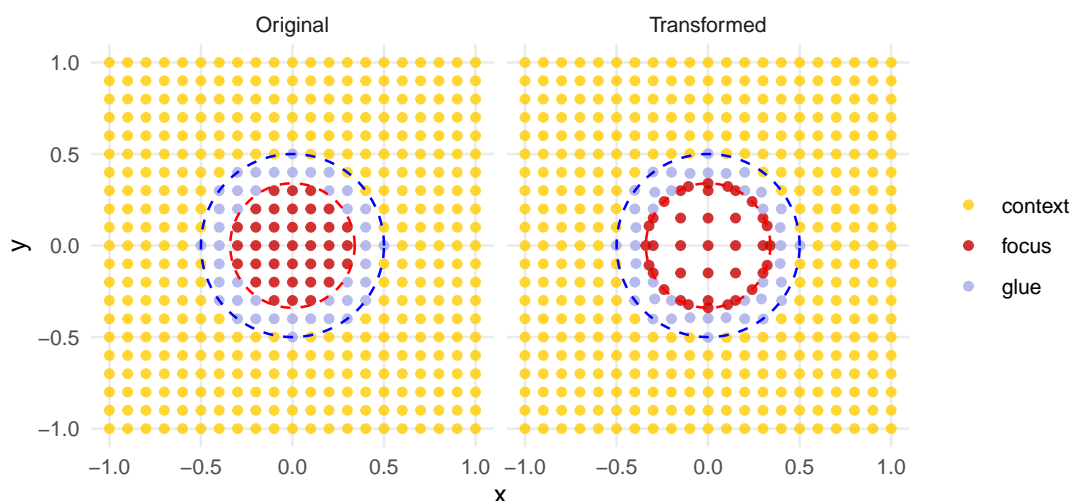


Figure 7: Benchmark performance of `sf_fisheye()` and `sf_fisheye_fg()`

with smooth boundary conditions, optionally perturbs angle via bell-shaped rotation, and converts back to Cartesian.

Fisheye FGC Transformation

$r_{in} = 0.34$, $r_{out} = 0.5$



It attaches diagnostic attributes (zone labels, original and new radii) consumed by plotting utilities but not affecting geometry reconstruction.

```
#>      x_new y_new
#> [1,] -1.0  -1
#> [2,] -0.9  -1
#> [3,] -0.8  -1
#> [4,] -0.7  -1
#> [5,] -0.6  -1
#> [6,] -0.5  -1

#> [1] "dim"          "dimnames"        "zones"           "original_radius"
#> [5] "new_radius"
```

Numeric stability at zone boundaries is ensured by clamping expansions in the focus so radii do not exceed r_{in} , and using smooth power curves in the glue so derivatives match across boundaries. The radial mapping is vectorized and runs in linear time in the number of vertices as seen in the benchmark 7.

Table 1: Coordinate transformation across zones for selected points

x	y	x_new	y_new	zone	r_orig	r_new
-1.0	-1	-1.000	-1.000	context	1.414	1.414
-0.9	-1	-0.900	-1.000	context	1.345	1.345
-0.8	-1	-0.800	-1.000	context	1.281	1.281
-0.7	-1	-0.108	-0.323	focus	0.316	0.340
-0.6	-1	0.000	-0.340	focus	0.300	0.340
-0.5	-1	0.108	-0.323	focus	0.316	0.340
-0.4	-1	0.000	-0.500	glue	0.500	0.500
-0.3	-1	-0.300	-0.400	glue	0.500	0.500
-0.2	-1	-0.200	-0.400	glue	0.447	0.448

Geometry reconstruction

At the top level is an all-in-one function `sf_fisheye()`, which presents the user-facing interface while keeping the numeric core untouched. It validates input, selects working CRS, resolves center, constructs normalization closures, and invokes `st_transform_custom()` to rebuild geometries.

The geometry walker `st_transform_custom()` acts as a drop-in analogue to `sf::st_transform()` but applies an arbitrary coordinate function. For each feature, it extracts coordinates via `sf::st_coordinates()`, yielding a matrix with columns $(x, y, L1, L2, \dots)$ where $L1$ and $L2$ index polygon rings and multi-polygon parts. Geometries are split by type:

- **POINT:** direct warp
- **LINestring:** warp each vertex, retain order
- **POLYGON:** process each ring (identified by $L1$) independently
- **MULTIPOLYGON:** nested by $(L1, L2)$ combinations

After transformation, polygon rings are explicitly closed by forcing first and last vertices to equality: $(x'_1, y'_1) = (x'_n, y'_n)$. This prevents numerical drift when the warp changes ring curvature. Geometries are rebuilt using `sf` constructors (`st_point()`, `st_linestring()`, `st_polygon()`, `st_multipolygon()`), combined into an `sfc` with original CRS, and spliced back into an `sf` if appropriate. Attributes are preserved because only the geometry column is replaced.

Table 1 illustrates coordinate transformations across zones for a vertical transect, showing radial expansion in the focus, smooth compression in the glue, and identity mapping in the context.

Design and extensibility

Utilities in `utils.R` provide `create_test_grid()` for diagnostics, `classify_zones()` for labeling, and `plot_fisheye_fg()` for visualization. Dataset documentation in `data.R` accompanies example layers (`vic`, `vic_fish`, `conn_fish`) used in tests.

For multi-layer maps, the normal process is combine all the layers into a single `sf` object and apply `sf_fisheye()`, then split the result later. One minimalist example for this approach is show in the code block below.

```
# Multi-layer example
bind <- dplyr::bind_rows(
  object_1 |> dplyr::mutate(layer="object_1"),
  object_2 |> dplyr::mutate(layer="object_2"))
```

```

bind_w <- sf_fisheye(
  bind,
  center = melb,
  r_in = 0.34,
  r_out = 0.55,
  zoom = 1.8,
  squeeze = 0.35)

object_1_transformed <- bind_w |>
  dplyr::filter(.layer == "object_1") |>
  dplyr::select(-.layer)

object_2_transformed <- bind_w |>
  dplyr::filter(.layer == "object_2") |>
  dplyr::select(-.layer)

```

The test suite mirrors the modular structure, covering boundary behaviour, zone labeling, CRS round-trips, ring closure, and performance. Functions follow tidyverse-oriented conventions (snake_case parameters, small exported surface). Behaviour is validated by tests; we aim for stability across versions but do not promise guarantees.

3.3 Parameters

The principal user interface is `sf_fisheye()`, which accepts an `sf` or `sfc` object and returns an object of the same top-level class whose geometry has been warped in a projection-aware manner. For clarity, we group arguments into data/CRS handling, centre selection, and radial warping, and we make explicit the invariants enforced by the implementation.

Data and CRS. The argument `sf_obj` supplies the features to be transformed. Before any calculation, empty geometries are removed and Z/M dimensions are dropped using `sf::st_zm()`, so that downstream computation operates on a strict $n \times 2$ coordinate matrix. The optional `target_crs` sets the working projected CRS; if provided, the input is transformed via `sf::st_transform()` and the original CRS is restored on return. When `target_crs = NULL` and the input is geographic (lon/lat), a projected working CRS is chosen deterministically from the layer's centroid: the default value is GDA2020 /; otherwise a UTM zone is inferred by longitude and hemisphere. This choice ensures the fisheye operates in metric units with bounded distortion across the extent of interest. The `preserve_aspect` flag governs normalisation: with `TRUE` (default) a uniform scale $s = \max(s_x, s_y)$ is applied, where s_x, s_y are bbox half-spans; with `FALSE`, independent scales are used per axis. Uniform scaling preserves circular symmetry of the focus and glue; per-axis scaling yields an elliptical interpretation that can be useful for long, narrow extents but should be used deliberately. Degenerate cases ($s_x = 0$ or $s_y = 0$) are handled by substituting a unit scale to avoid division by zero.

Centre selection. The lens centre may be specified in several forms. The preferred interface is `center`, which takes precedence over legacy `cx`, `cy`. If `center` is a numeric pair and `center_crs` is provided (e.g., "EPSG:4326"), the point is transformed into the working CRS. If `center_crs` is omitted, a heuristic interprets pairs that lie within $|\text{lon}| \leq 180$, $|\text{lat}| \leq 90$ as WGS84 and transforms them accordingly; otherwise the values are assumed to be already in working-CRS map units. Any `sf`/`sfc` geometry may be used as `center`; non-point centres are combined and reduced to a centroid and then transformed to the working CRS, which is often convenient when the focal area is a polygon (e.g., a CBD boundary) or a set of points (e.g., incident locations). Finally, when the argument `{normalized_center = TRUE}`, `center` is interpreted as a pair in $[-1, 1]$ relative to the bbox midpoint and the chosen normalisation (uniform or per-axis). Normalised centres make parameter sets portable across datasets of different extents and are a natural fit for parameter sweeps in reproducible pipelines. If no centre is supplied, the bbox midpoint is used; this default is stable under reprojection.

Radial warping. The radii r_{in} and r_{out} define the focus and glue boundaries in the normalised coordinate space and must satisfy $r_{out} > r_{in}$. The interpretation of these radii depends on `preserve_aspect`. With uniform scaling, a circle of radius r_{in} in unit space corresponds to a circle of radius $r_{in}s$ in map units; with per-axis scaling, the corresponding shape is an axis-aligned ellipse with semi-axes $r_{in}s_x$ and $r_{in}s_y$. Inside the focus, distances from the centre are multiplied by `zoom_factor`; to prevent overshoot, the implementation clamps r' so that points do not cross the r_{in} boundary. Across the glue, `squeeze_factor` in $(0, 1]$ controls how strongly intermediate radii compress: smaller values create tighter compression near the boundaries and a more pronounced “shoulder” in the middle of the glue; larger values approach a linear transition. The method selects the family of curves used in the glue. The default “expand” applies a symmetrical power law that expands inward and outward halves of the glue to maintain visual balance around the midpoint; “outward” biases the map towards r_{out} , keeping the outer boundary steadier and pushing more deformation into the inner portion of the glue. The optional `revolution` parameter adds a bell-shaped angular twist inside the glue of magnitude $\rho 4u(1 - u)$, where u is the normalised glue radius. This rotation vanishes at both glue boundaries and peaks at the midpoint, preserving continuity. Positive values rotate counter-clockwise, negative values clockwise; values are specified in radians.

Inter-parameter interactions and invariants. The following constraints and behaviours are enforced: $r_{out} > r_{in} > 0$; `zoom_factor` ≥ 1 (values close to one yield gentle focus); `squeeze_factor` in $(0, 1]$ ($= 1$ approaches linear); and monotonicity of the radial map so that ordering by distance from the centre is preserved. The choice of `preserve_aspect` affects the physical size of radii and thereby the impact of a given parameter set on different datasets; using uniform scaling with a normalised centre yields the most portable configurations. Twisting via `revolution` is confined to the glue; it does not change radii and therefore does not affect the classification of points into zones. Because angles are modified only in the glue, bearings inside the focus and in the context are preserved.

Return value and side effects. The function returns an object of the same top-level class as its input (`sf` or `sfc`). For `sf` inputs, non-geometry columns are preserved verbatim; only the geometry column is replaced. The original CRS is restored before return so that downstream plotting and analysis code does not need to change. On malformed geometries, the implementation emits a warning and returns an empty geometry of the appropriate family to preserve row count and indices. For exploratory diagnostics, the low-level `fisheye_fgc()` returns a coordinate matrix with attributes “zones”, “original_radius”, and “new_radius”; these can be used to plot scale curves and verify parameter effects prior to applying the transform to complex geometries.

3.4 Common choices

Although the parameter space is continuous, certain regimes recur in practice and can serve as reliable starting points. We describe these regimes and articulate the trade-offs that motivate each choice. The recommendations assume the default `preserve_aspect = TRUE`; when per-axis scaling is enabled, translate radii to semi-axes using the `bbox` half-spans.

Quick start (synthetic grid). Set r_{in} to 0.30–0.35 and r_{out} to 0.55–0.70. Pair this with `zoom_factor` between 5 and 10 and `squeeze_factor` near 0.35 for a balanced focus that still shows context. Stick with `method = “expand”` and `revolution = 0` unless you explicitly need outer rigidity or a twist.

In this simple example, we demo the effect of zoom factor 1.5 and 2 on a synthetic grid to demonstrate how the point movement was effected by the zoom factor.

However, as demonstrated below, the `revolution` effect might make the distortion of the linestring object more obvious, comparing to just only the zoom factor effect.

For polygon object, we can see the same distortion effect as the linestring.

As we can see, the `revolution` effect create a vortex or zoom wheel effect into the focus zone, which may be useful in some cases. For manuscripts and dashboards, prefer

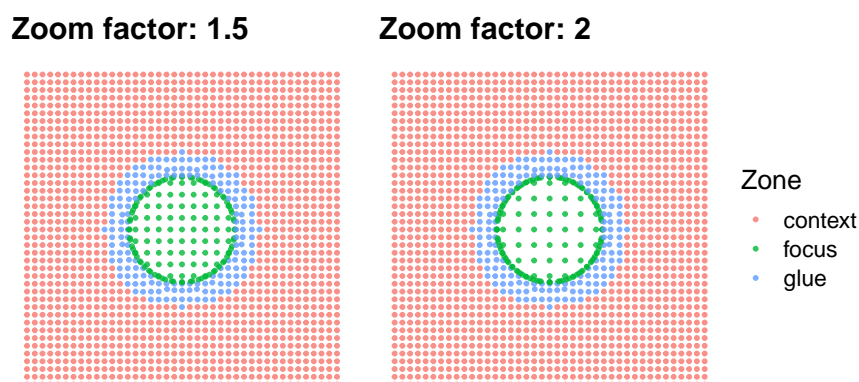


Figure 8: Zoom factors for balanced metropolitan focus within a state.

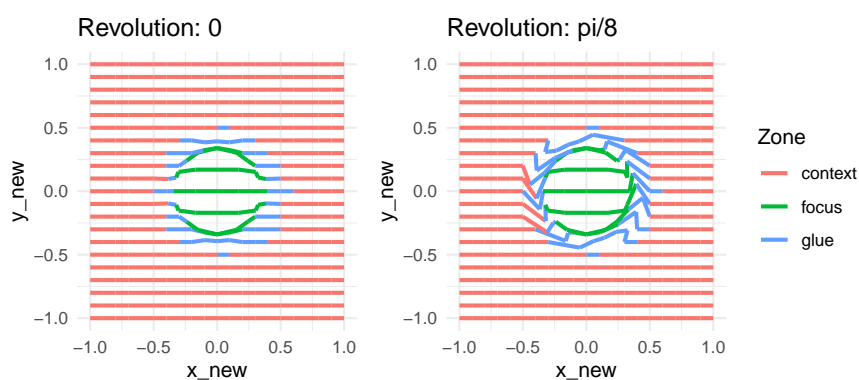


Figure 9: Revolution effect

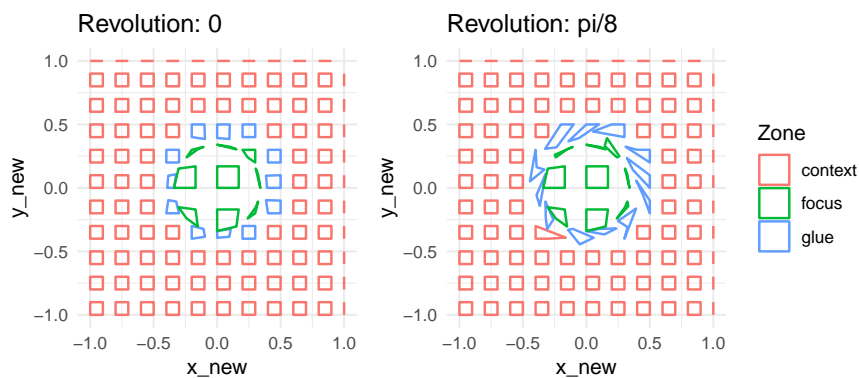


Figure 10: Revolution effect with polygon

Melbourne hospitals

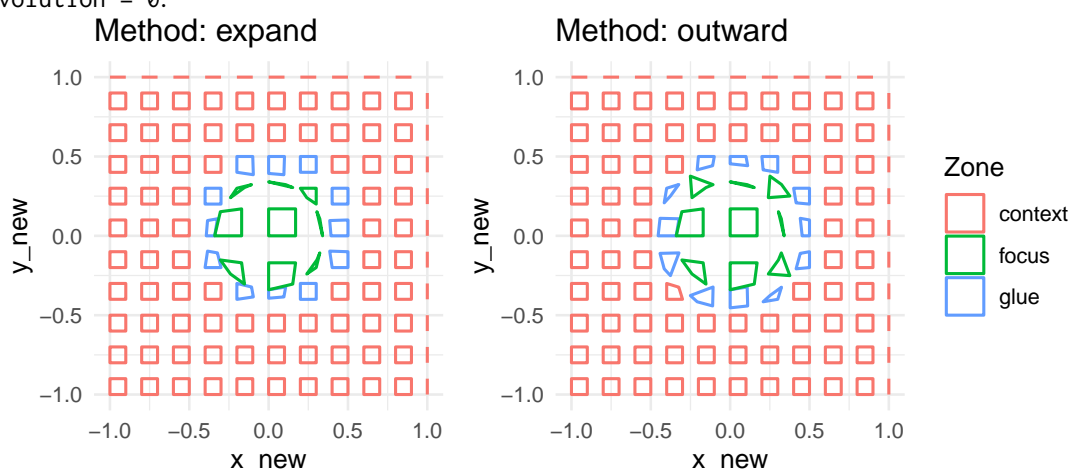


Melbourne RACFs



Figure 11: Standard maps: hospital (red) and RACF (blue) points plotted over Victoria without fisheye.

revolution = 0.



Similarly, start with "expand" and adopt "outward" only when outer stability is an explicit requirement. Always annotate or at least describe the distortion in figure captions so readers do not mistake warped areas for standard projections.

Simple tweaks. If linework kinks, raise `squeeze_factor` slightly (e.g., 0.45). If the focus feels too tight, lower `zoom_factor` toward 4–6. For reproducible comparisons, keep `normalized_center = TRUE` and reuse the same radii across runs.

4 Exploring the implementation using Shiny

Explain the shiny app structure, and how it can be used interactively to explore a spatial data set.

Screenshot

5 Application to Victorian ambulance transfer records

We focus on one application: Victorian hospitals and residential aged care facilities (RACFs). Transfer counts are simulated; coordinates come from `conn_fish`. We sample 10 hospitals and 10 RACFs to avoid clutter, use a grey basemap for contrast, and highlight how points, polygons, and connection lines behave under the same lens.

First, we can plot separately the sampled hospitals and the age care facilities to see where they actually are.

Afterward, we can plot the connection between the hospital and the age care facilities to see the movement of the patient between the two as shown in the Figure 12.

Victorian hospital–RACF network (sampled)

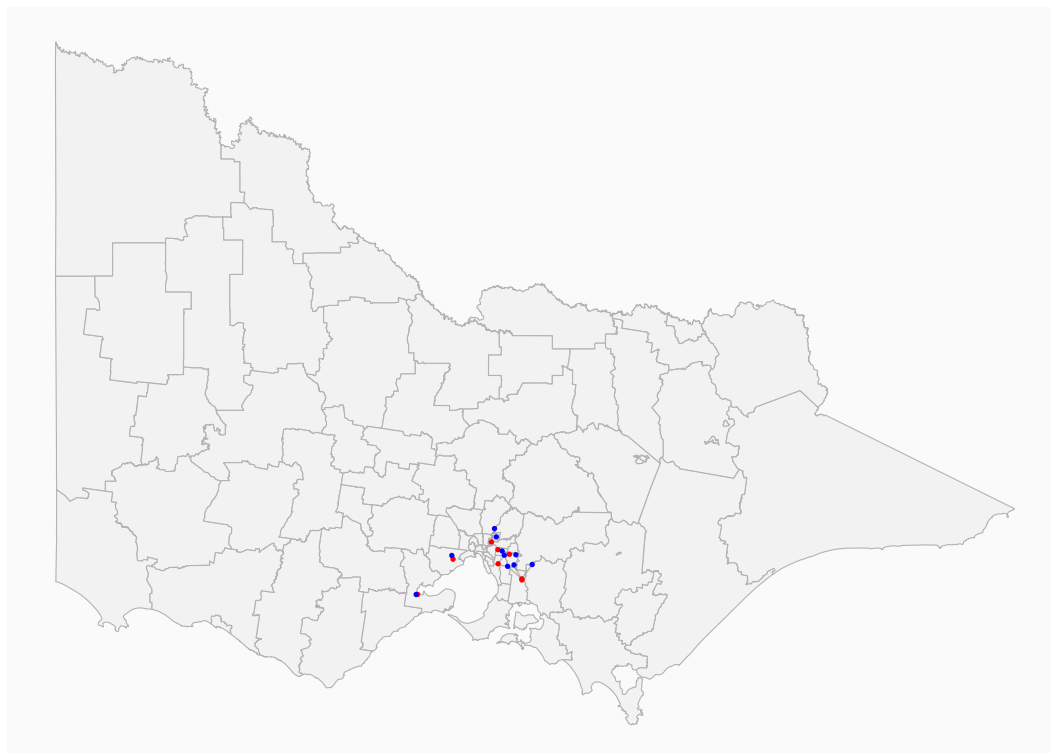


Figure 12: Standard Victorian map with sampled hospitals (red), RACFs (blue), and simulated transfer lines on a grey background.

As we can see, comparing to the scale of Victoria, with the default scale method, the network mostly invisible. To magnify the network, we can apply a fisheye lens.

We could also include an interactive version of the plot above, thanks to the native support and integration of `ggplotly` with `sf` object.

The fisheye clarifies dense metro structure without losing state context. Hospitals and RACFs that previously overlapped separate cleanly; transfer lines remain connected to their endpoints because every layer uses the same centre and radii. Outside the glue zone, geometry remains stable, so readers can still place Melbourne within Victoria. To inspect a different hub (e.g., Geelong), change center to that polygon; multiple centres can be explored by re-running `sf_fisheye()` with alternative centres and comparing panels side by side.

For polygons beyond metro Melbourne, the same lens settings apply; polygon edges bend smoothly but retain adjacency. Because transfer counts are simulated, readers should treat magnitude as illustrative only; spatial relationships (what becomes readable after zoom vs. what remains hidden) follow from the geometry.

In this case, we can see that some point and connection was squished together since they were in the glue zone. We can change that by introducing a new center, and shrink the original focus zone. However, the mechanic of handling overlapping zone between multiple center is not implemented in this version yet.

6 Discussion

Contribution `mapycusmaximus` provides an `sf`-native implementation of the FGC fisheye that is projection-aware, parameterised in normalised units, and safe across points, lines, and polygons. The package separates radial mapping from geometry orchestration, exposes explicit controls over focus, glue, and context, and preserves attributes and CRS invariants



Figure 13: Fisheye view magnifies greater Melbourne while keeping statewide context; lines, points, and polygons stay aligned.

for reproducible pipelines with ggplot2.

Relation to alternatives Unlike cartograms (thematic distortion), hex/regular tile maps (discrete abstraction), or inset/multi-panel layouts (spatial separation), the FGC lens delivers continuous magnification within a single map while preserving topology and bearings. This reduces cognitive load for readers who must relate local phenomena to their broader geography.

Limitations The fisheye introduces non-metric distortion in the focus and glue; therefore, use it for visual exploration and communication, not for metric analysis. Aggressive zoom or squeeze can impair legibility near the glue boundary; conservative defaults and `revolution = 0` are recommended for publication maps. When comparing multiple regions, prefer `normalized_center = TRUE` with fixed radii to ensure visual comparability. At present, exact matching of focus and glue radii across separately transformed layers may require a manual step (the user have to manually merge the two or more layers, perform the fisheye transformation, then separated the transformed layers).

Future work Planned extensions include anisotropic or elliptical profiles, multi-focus blending, first-class raster support via warped grids and resampling, and interactive focus selection for exploratory analysis. We also plan an API for shared normalisation and radius locking across layers (e.g., a `combine_fisheye`) so that multiple layers can be warped with identical scale and then returned transformed. Performance improvements via vectorised geometry walkers or GPU acceleration would benefit dense polygonal datasets. Clear figure captions and scale disclaimers remain essential to communicate the presence and intent of distortion.

7 Conclusion

FGC fisheye transformations offer a concise, CRS-aware way to emphasise local structure without losing geographic context. By starting from a point-wise radial map and integrating carefully with `sf` for geometry reconstruction, the approach keeps figures continuous and overlays aligned. The examples demonstrate clearer narratives for metropolitan focus while maintaining state- or nation-level context.

8 AI Use Declaration

We used AI tools to assist with code refactoring and drafting portions of the text. All methods, parameter settings, and claims were designed and reviewed by the authors, and we verified outputs with the package’s test suite and example renders.

9 Resources

The github link for this paper is [here](#).

The `mapycusmaximus` package is available on [GitHub](#).

The slideshow for this package can be found [here](#).

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