

ggpedigree: Visualizing Pedigrees with ‘ggplot2’ and ‘plotly’

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Summary

Pedigree diagrams underpin research and practice across genetics, animal breeding, genealogy, forensics, and counseling. They help medical geneticists trace Mendelian disease inheritance and identify at-risk relatives; enable dairy breeders to plan matings that improve milk yield; support genealogists in reconstructing ancestry; assist forensic scientists establish familial connections in criminal investigations; and facilitate family therapists and counselors in understanding their clients’ relationships (McGoldrick, Gerson, & Petry, 2020). Early R tools such as kinship2 (Sinnwell, Therneau, & Schaid, 2014) plot simple nuclear families effectively, but they do not scale to today’s pedigrees that can exceed 1,000s of individuals. As datasets have grown, researchers now work with increasingly complex family structures, including large-scale plant breeding pedigrees (Shaw, Graham, Kennedy, Milne, & Marshall, 2014), web-based pedigree management systems (Ranaweera, Makalic, Hopper, & Bickerstaffe, 2018), interactive pedigree editors (Carver et al., 2018), and behavior genetic studies of extended family structures (Garrison et al., 2023; M. Hunter, Garrison, Burt, & Rodgers, 2021). Such complexity exposes limits in existing tools, which often struggle to handle large datasets and complex relationships. ggpedigree addresses this need by combining a vectorised layout algorithm, ggplot2 output, and optional plotly interactivity.

Statement of need

Pedigree visualization has traditionally relied on specialized proprietary software (e.g., Progeny, GenoPro) or R packages like kinship2 (Sinnwell et al., 2014) or pedtricks (J. Martin, Wolak, Johnston, & Morrissey, 2025), which generate static plots using base graphics and minimal layout control. While these tools are functional for small, nuclear pedigrees, their limitations become pronounced when working with complex, modern pedigree datasets. First, base R graphics lack the modular design and extensibility needed for generating publication-quality pedigree figures. Second, most existing R-based tools offer no interactivity, making it difficult to explore large pedigrees dynamically. Third, current solutions are poorly integrated with tidyverse workflows and do not produce ggplot2-based objects that users can customize or extend (Hadley Wickham, 2016).

Recent advances in behavior genetics and genetic epidemiology (Garrison et al., 2023; McArdle & McDonald, 1984) have introduced new visualization demands, particularly for extended families and highly interconnected pedigrees. Modern pedigree datasets can include hundreds or thousands of individuals across multiple generations, with overlapping relationships, consanguinity, remarriages, and twin sets that strain existing layout algorithms. Additionally, researchers increasingly need to overlay phenotypic information, genetic relatedness values, and model outputs onto pedigree plots to support interpretation and communication of their findings.

Several R packages attempt to address these challenges with built-in pedigree plotting functions.

kinship2 (Sinnwell et al., 2014) remains widely used but produces static base graphics and relies on non-vectorized recursive layout functions that does not scale well to large families. A partial ggplot2 implementation exists but is incomplete, non-vectorized, and not actively maintained. The geneHapR (Zhang, Jia, & Diao, 2023) package focuses on haplotype visualization rather than general pedigree structure. The pedgene package provides some plotting functions but is primarily designed for association testing. None of these packages offers the combination of modern ggplot2 integration, interactive capabilities, and extensibility that ggpedigree provides. ggpedigree addresses these limitations by providing a comprehensive visualization framework built on modern R graphics infrastructure. It leverages the extensive customization capabilities of ggplot2 while adding specialized functionality for pedigree-specific visualization challenges.

Software Architecture

ggpedigree is built on a modular architecture that separates data processing, layout calculation, and visualization layers. The core workflow involves: (1) data standardization and family structure analysis using BGmisc functions, (2) coordinate calculation using algorithms adapted from kinship2, (3) relationship connection mapping, and (4) layer-based plot construction using ggplot2 geometry functions. This design allows users to customize any aspect of the visualization while maintaining computational efficiency for large pedigrees. The package integrates tightly with the broader R ecosystem, particularly the tidyverse (H. Wickham et al., 2019) and BGmisc (Garrison, S. M., Hunter, M. D., Lyu, X., Trattner, J. D., & Burt, S. A., 2024). All functions return standard R objects (ggplot or plotly) that can be further customized using familiar syntax, ensuring accessibility for users already comfortable with modern R data science workflows.

Features

We briefly describe the main features of the ggpedigree package below. A more detailed description of the package's features and usage is available from the [package vignettes](#), including examples of how to create static and interactive pedigree plots, customize aesthetics, and visualize relatedness matrices. Additional example data include squirrel data from the Kluane Red Squirrel Project (McFarlane et al., 2014, 2015) and Targaryen family data from the Song of Ice and Fire universe (G. R. Martin, 1997; G. R. R. Martin, 2018; G. R. R. Martin, Jr, & Antonsson, 2014). These datasets demonstrate the package's capabilities for both empirical and fictional pedigrees.

Core Functions

- Data Standardization and Family Structure Analysis: `ggPedigree()` integrates with BGmisc functions like `ped2fam()` to organize individuals by family, `recodeSex()` to standardize sex coding, and `checkParentIDs()` to validate pedigree structures. The function handles complex scenarios including consanguineous relationships and individuals appearing in multiple pedigree positions.
- Coordinate Calculation: `calculateCoordinates()` computes optimal positioning for individuals using algorithms adapted from `kinship2::align.pedigree`, with enhancements for large multi-generational pedigrees and complex family structures. The function returns coordinate grids that optimize spacing and minimize visual overlap. These steps are vectorized as much as possible to ensure computational efficiency, especially for large pedigrees.
- Relationship Connection Mapping: `calculateConnections()` generates connection paths between family members, mapping parent-child, sibling, spousal, and twin relationships. The function determines midpoints for line intersections and handles overlapping connections with specialized curved segments. These calculations are optimized for large

datasets by using vectorized operations rather than the loop-based approaches used in kinship2.

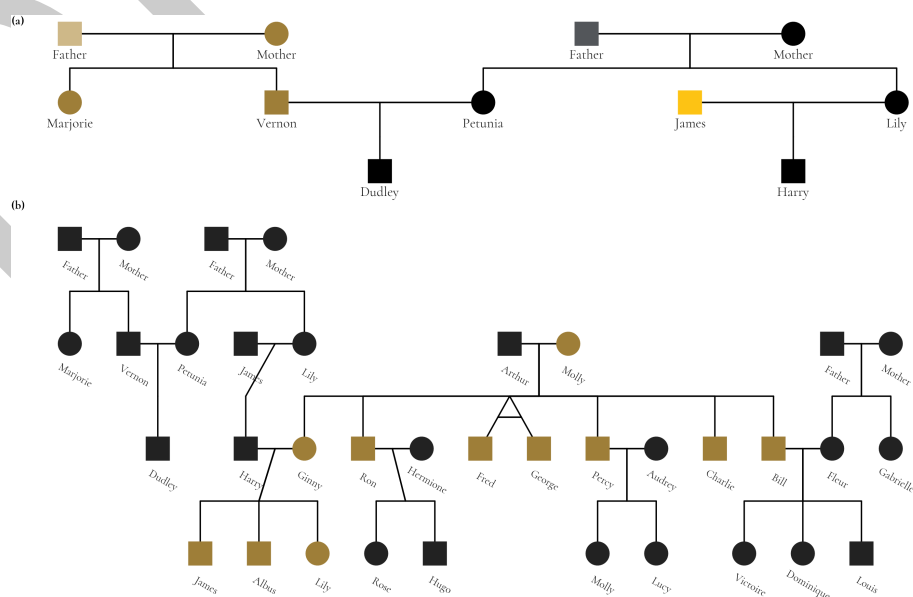
- **Layer-based Plot Construction:** `ggPedigree()` constructs plots using `ggplot2` geometry functions, returning standard `ggplot2` objects that integrate with existing R workflows. `ggPedigreeInteractive()` extends plots into interactive `plotly` widgets with hover tooltips and dynamic exploration capabilities. A comprehensive config system allows customization of over 100 aesthetic and layout parameters.
- **Focal Individual Highlighting:** Advanced functionality to highlight specific individuals and their relatives based on additive genetic, mitochondrial, or other relationship matrices.

Specific Visualization Functions

- **Pedigree Plotting:** `ggPedigree()` creates static pedigree plots using `ggplot2`, supporting complex family structures, multiple generations, and customizable aesthetics. It can handle large pedigrees with thousands of individuals while maintaining clarity and readability.
- **Interactive Pedigree Visualization:** `ggPedigreeInteractive()` generates interactive pedigree plots using `plotly`, allowing users to explore large pedigrees dynamically. Users can configure tooltip content to display individual IDs, names, phenotypic information, and other relevant data. It supports tooltips, zooming, and panning for detailed exploration of family structures.
- **Relatedness Matrix Heatmaps:** `ggRelatednessMatrix()` creates customizable heatmap visualizations for genetic relatedness matrices with support for hierarchical clustering, interactive exploration, and seamless integration with `BGmisc` relatedness calculations.
- **Phenotype-Relatedness Analysis:** `ggPhenotypeByDegree()` produces specialized plots for visualizing phenotypic correlations as a function of genetic relatedness, including confidence intervals and statistical summaries for quantitative genetic analysis.

Illustrative Example of ggPedigree

Here, we illustrate some of the features by reproducing the figure from (M. D. Hunter et al., 2025) but restyled to conform to the Wake Forest Style Guidelines. As you can see, the `ggpedigree` package allows for complex family structures to be visualized in a clear and aesthetically pleasing manner.



We have combined two figures using `patchwork` (Pedersen, 2025) to show the pedigree of the

121 Potter family from the Harry Potter series. The first plot (a) highlights the individual mtDNA
122 lines in the Dursley and Evans families, while the second plot (b) shows the entire family
123 structure with mitochondrial relationships highlighted in a gradient color scale. Features such
124 as focal individual highlighting, custom color scales, text labels, fonts, are easily implemented.
125 The package supports both static and interactive plots, allowing users to explore large pedigrees
126 dynamically.

127 The plot includes individuals' names, relationships, and phenotypic information, all while
128 maintaining clarity even with overlapping relationships. Although pleasing, this level of
129 customization is not required, as default settings produce clear and informative pedigree plots.

130 Collectively, these tools provide a valuable resource for behavior geneticists and others who
131 work with extended family data. They were developed as part of a grant and have been used
132 in several ongoing projects, presentations (Garrison, 2024; M. D. Hunter, Garrison, Lyu, Good,
133 & Burt, 2024), and forthcoming papers (Burt et al., 2025; M. D. Hunter et al., 2025; Lyu,
134 Hunter, Good, Carroll, & Garrison, 2025).

135 Availability

136 The ggpedigree package is open-source and available on both GitHub at <https://github.com/R-Computing-Lab/ggpedigree> and the Comprehensive R Archive Network (CRAN) at <https://cran.r-project.org/package=ggpedigree>. It is licensed under the GNU General Public License.

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210 Appendix

211 Source code for the figures above is available in the vignettes/articles/_paper.Rmd file.
212 The code below reproduces the figure using the ggpedigree package.

```
library(ggpedigree) # ggPedigree lives here
library(BGmisc) # helper utilities & example data
library(tidyverse) # for data manipulation and plotting
library(showtext)
library(sysfonts)
library(patchwork) # for combining plots

# Load the potter pedigree data
data("potter")
# Removing Last names
df_potter <- potter %>%
  mutate(
    name = case_when(
      personID == 1 ~ "Vernon",
      personID == 2 ~ "Marjorie",
      personID == 3 ~ "Petunia",
      personID == 4 ~ "Lily",
      personID == 5 ~ "James",
      personID == 6 ~ "Dudley",
      personID == 7 ~ "Harry",
      personID == 8 ~ "Ginny",
      personID == 9 ~ "Arthur",
      personID == 10 ~ "Molly",
      personID == 11 ~ "Ron",
      personID == 12 ~ "Fred",
      personID == 13 ~ "George",
      personID == 14 ~ "Percy",
      personID == 15 ~ "Charlie",
      personID == 16 ~ "Bill",
      personID == 17 ~ "Hermione",
      personID == 18 ~ "Fleur",
      personID == 19 ~ "Gabrielle",
      personID == 20 ~ "Audrey",
      personID == 21 ~ "James",
      personID == 22 ~ "Albus",
      personID == 23 ~ "Lily",
      personID == 24 ~ "Rose",
      personID == 25 ~ "Hugo",
      personID == 26 ~ "Victoire",
      personID == 27 ~ "Dominique",
      personID == 28 ~ "Louis",
      personID == 29 ~ "Molly",
      personID == 30 ~ "Lucy",
      personID == 101 ~ "Mother",
      personID == 102 ~ "Father",
      personID == 103 ~ "Mother",
      personID == 104 ~ "Father",
      personID == 105 ~ "Mother",
      personID == 106 ~ "Father"
    )
  )
```



```

)
# Load Google fonts for styling
font_add_google(name = "Cormorant", family = "cormorant")
showtext_auto() # Load the showtext package to render Google fonts

# Set the WFU style guidelines for the plot
text_color_wfu <- "#222222" # dark grey for text labels
focal_fill_color_values_wfu <- c(
  "#9E7E38", "#000000", "#FDC314", "#CEB888", "#53565A")
family_wfu <- "cormorant"
text_size_wfu <- 14

# Create Panel A
m1 <- ggPedigree(df_potter %>% filter(personID %in% c(1:7, 101:104)),
  famID = "famID",
  personID = "personID",
  config = list(
    label_include = TRUE,
    label_column = "name",
    point_size = 8,
    # outline_include = TRUE,
    focal_fill_personID = 8,
    outline_multiplier = 1.5,
    segment_linewidth = 0.5,
    label_text_size = 11, # 3,
    label_text_color = text_color_wfu,
    axis_text_color = text_color_wfu,
    label_text_family = family_wfu,
    focal_fill_include = TRUE,
    label_nudge_y = -0.30,
    focal_fill_method = "manual",
    focal_fill_color_values = focal_fill_color_values_wfu,
    focal_fill_force_zero = TRUE,
    label_method = "geom_text",
    focal_fill_na_value = text_color_wfu,
    focal_fill_scale_midpoint = 0.40,
    focal_fill_component = "matID",
    focal_fill_labels = NULL,
    sex_legend_show = FALSE,
    sex_color_include = FALSE
  )
) + guides(shape = "none") + theme(
  plot.title = element_blank(),
  plot.title.position = "plot",
  text = element_text(family = family_wfu,
    size = 14)
) + coord_cartesian(ylim = c(3.25, 1),
  clip = "off")

# Create Panel B
m2 <- ggPedigree(df_potter,
  famID = "famID",
  personID = "personID",
  config = list(
    label_include = TRUE,
    label_column = "name",

```

```

point_size = 8,
focal_fill_personID = 8, # Molly Weasley
outline_multiplier = 1.5,
segment_linewidth = 0.5,
label_text_size = 9,
label_text_family = family_wfu,
label_text_color = text_color_wfu,
axis_text_color = text_color_wfu,
label_nudge_y = -0.3,
label_nudge_x = .1,
focal_fill_include = TRUE,
focal_fill_method = "gradient2",
focal_fill_high_color = "#9E7E38",
focal_fill_mid_color = "#9E7E38",
focal_fill_low_color = text_color_wfu[2],
focal_fill_scale_midpoint = 0.85,
focal_fill_component = "mitochondrial",
focal_fill_force_zero = TRUE,
label_method = "geom_text",
focal_fill_na_value = text_color_wfu,
label_text_angle = -30,
sex_legend_show = FALSE,
sex_color_include = FALSE
)
) + theme(
  legend.position = "none",
  plot.title = element_blank(),
  plot.title.position = "plot",
  text = element_text(family = family_wfu,
    size = text_size_wfu, face = "bold")
) + coord_cartesian(ylim = c(4.25, .9), clip = "off")

# Combine the two plots using patchwork
showtext_auto()
result <- m1 + m2 +
  plot_layout(
    ncol = 1, heights = c(1.1, 2.5),
    guides = "collect", tag_level = "new"
  ) +
  plot_annotation(
    tag_levels = list(c("(a)", "(b)")),
    theme = theme(plot.margin = margin(0, 0, 0, 0), )
  ) +
  guides(shape = "none") &
  theme(
    legend.position = "none",
    plot.margin = unit(c(0, 0, 0.0, 0), "lines"),
    plot.tag = element_text(family = family_wfu,
      size = 2*text_size_wfu, face = "bold")
  )

# save as a png
ggsave(
  filename = "wfu_potter_pedigree.png",
  plot = result,

```



```
width = 9.5, height = 6, dpi = 300, units = "in"  
)
```

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