

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

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## Software

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## Summary

Pedigree diagrams underpin research and practice across genetics, animal breeding, genealogy, forensics, and counseling. They help medical geneticists trace the inheritance of Mendelian diseases and identify at-risk relatives; enable dairy breeders to plan matings that improve milk yield; support genealogists in reconstructing ancestry; assist forensic scientists in establishing 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 the limitations of 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, Pedigree Viewer) or R packages like kinship2 (Sinnwell et al., 2014), pedtools (Magnus Dehli Vigeland, 2021a), pedtricks (J. Martin, Wolak, Johnston, & Morrissey, 2025). While these tools are functional for many use cases, their limitations become pronounced when working with complex, modern pedigree datasets or when more detailed customization is required. Most existing R packages focus on base graphics or simple ggplot2 implementations.

First, base R graphics lack the modular design and extensibility needed for generating publication-quality pedigree figures. For example, kinship2 (Sinnwell et al., 2014) uses base graphics and loop-based layout functions that do not scale well. pedtricks (J. Martin et al., 2025), by contrast, returns ggplot2-based plots and is designed for large, multigenerational animal pedigrees, but offers minimal support for annotation layering, per-individual theming, or integration with phenotypic and model-based data

Second, most existing R-based tools offer no interactivity. Static graphics are often sufficient for publication, but interactivity improves exploration and communication during model development or data cleaning. A notable exception is pedtools (Magnus Dehli Vigeland, 2021b), which offers a sister shiny app, QuickPed (Magnus D. Vigeland, 2022). While the R ecosystem includes libraries like plotly that support interactive plotting, these features have yet to be integrated into pedigree functions.

41 Third, current solutions are often poorly integrated with tidyverse workflows and do not expose  
42 the full theming and layering capabilities familiar to users of ggplot2 (Hadley Wickham, 2016).  
43 In animal-focused workflows, rapid rendering often takes precedence over aesthetic flexibility,  
44 which can hinder interpretability in human-focused research.

45 Recent advances in behavior genetics and genetic epidemiology (Garrison et al., 2023; Lyu,  
46 Hunter, Good, Carroll, & Garrison, 2025; McArdle & McDonald, 1984) have introduced new  
47 visualization demands, particularly for extended families and highly interconnected pedigrees.  
48 Modern pedigree datasets can include hundreds or thousands of individuals across multiple  
49 generations, with overlapping relationships, consanguinity, remarriages, and twin sets that strain  
50 existing layout algorithms. Additionally, researchers increasingly need to overlay phenotypic  
51 information, genetic relatedness values, and model outputs onto pedigree plots to support  
52 interpretation and communication of their findings.

53 Several R packages attempt to address these challenges with built-in pedigree plotting functions.  
54 kinship2 (Sinnwell et al., 2014) remains widely used but produces static base graphics and  
55 relies on non-vectorized recursive layout functions that do not scale well to large families.  
56 A partial ggplot2 implementation exists but is incomplete, non-vectorized, and not actively  
57 maintained. pedtricks, a revival of pedantics (Morrissey & Wilson, 2010), provides a ggplot2-  
58 based implementation for large animal pedigrees but lacks extensibility and interactivity. The  
59 geneHapR (Zhang, Jia, & Diao, 2023) package focuses on haplotype visualization rather than  
60 general pedigree structure. The pedgene package (Daniel & Sinnwell, 2024) offers some  
61 plotting functions but is primarily designed for association testing rather than visualization.  
62 The pedigreejs package (Carver et al., 2018) provides an interactive pedigree editor but does  
63 not integrate with R or ggplot2, limiting its utility for R users.

64 None of these packages offers the combination of modern ggplot2 integration, interactive  
65 capabilities, and extensibility that ggpedigree provides. ggpedigree addresses these limitations  
66 by providing a comprehensive visualization framework built on modern R graphics infrastruc-  
67 ture. It leverages the extensive customization capabilities of ggplot2 while adding specialized  
68 functionality for pedigree-specific visualization challenges.

## 69 Software Architecture

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

## 80 Features

81 I describe the main features of the ggpedigree package below. More detailed descriptions of  
82 features and usage is available in the [package vignettes](#), including examples of how to create  
83 static and interactive pedigree plots, customize aesthetics, and visualize relatedness matrices.  
84 Additional example data include squirrel data from the Kluane Red Squirrel Project (McFarlane  
85 et al., 2014, 2015) and Targaryen family data from the Song of Ice and Fire universe (G. R.  
86 Martin, 1997; G. R. R. Martin, 2018).

## 87 Core Functions

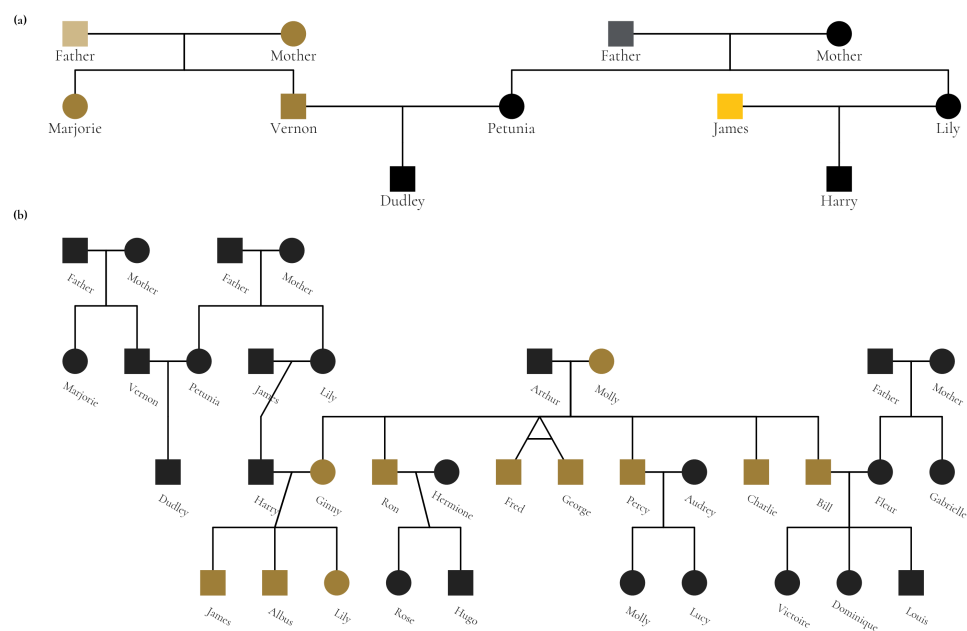
- 88     ▪ Data Standardization and Family Structure Analysis: `ggPedigree()` integrates with  
89     BGmisc functions like `ped2fam()` to organize individuals by family, `recodeSex()` to  
90     standardize sex coding, and `checkParentIDs()` to validate pedigree structures. The  
91     function handles complex scenarios including consanguineous relationships and individuals  
92     appearing in multiple pedigree positions.
- 93     ▪ Coordinate Calculation: `calculateCoordinates()` computes optimal positioning for  
94     individuals using algorithms adapted from `kinship2::align.pedigree`, with enhance-  
95     ments for large multi-generational pedigrees and complex family structures. The function  
96     returns coordinate grids that optimize spacing and minimize visual overlap. These steps  
97     are vectorized as much as possible to ensure computational efficiency, especially for large  
98     pedigrees.
- 99     ▪ Relationship Connection Mapping: `calculateConnections()` generates connection paths  
100     between family members, mapping parent-child, sibling, spousal, and twin relationships.  
101     The function determines midpoints for line intersections and handles overlapping con-  
102     nections with specialized curved segments. These calculations are optimized for large  
103     datasets by using vectorized operations rather than the loop-based approaches used in  
104     `kinship2`.
- 105     ▪ Layer-based Plot Construction: `ggPedigree()` constructs plots using `ggplot2` geome-  
106     try functions, returning standard `ggplot2` objects that integrate with existing R work-  
107     flows. `ggPedigreeInteractive()` extends plots into interactive `plotly` widgets with hover  
108     tooltips and dynamic exploration capabilities. A comprehensive config system allows  
109     customization of over 100 aesthetic and layout parameters.
- 110     ▪ Focal Individual Highlighting: Advanced functionality to highlight specific individuals and  
111     their relatives based on additive genetic, mitochondrial, or other relationship matrices.

## 112 Specific Visualization Functions

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

## 128 Illustrative Example of `ggPedigree`

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



I have combined two figures using patchwork (Pedersen, 2025) to show the pedigree of the Potter family. The first plot (a) highlights the individual mtDNA lines in the Dursley and Evans families, while the second plot (b) shows the entire family structure with mitochondrial relationships highlighted by a gradient color scale. Features such as focal individual highlighting, custom color scales, text labels, and fonts are easily implemented. The package supports both static and interactive plots, allowing users to explore large pedigrees dynamically.

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

## Availability

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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## 229 Appendix

230 Source code for the figures above is available in the vignettes/articles/\_paper.Rmd file.  
231 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,
    focal_fill_personID = 8,
    outline_multiplier = 1.5,
    segment_linewidth = 0.5,
    label_text_size = 12,
    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"
)
```