

1 ggtreeExtra: A universal package to visualize compact circular layers of phylogenetic tree

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1.1 Examples of mapping and visualizing associated data on circular layout tree.

It can help to find patterns and generate new hypotheses by integrating and visualizing associated data to the phylogenetic tree. The associated data type of phylogenetic tree can be roughly divided into continuous data and categorical data (discrete data). The continuous data represents measurements, they can be measured but not be counted, such as the height, weight, abundance of species, gene expression and the number of target genes etc. The categorical data represents characteristics, they can not be measured but they can be counted, such as endemic region information of virus, taxonomy information of species, type of target gene and sampling location information etc. Certainly, categorical data can also take on numerical values (for example, 1 for target gene A, 2 for target gene B). The associated data are also often multi-dimensional. To display them on the outer of phylogenetic tree more compactly, we developed *ggtreeExtra* to annotate multi-dimensional data to the outer ring of circular phylogenetic tree. It is a universal tool, based on the grammar of graphics(Wilkinson 2012). So user can easily map the variables (abundance of species, length of genome, sampling location) of associated data to aesthetic attributes (size, color, shape) of outer geometric objects (bar, point, boxplot) of circular phylogenetic tree using *ggtreeExtra*. Here, we present several examples to elucidate how to map and display the associated data on the outer rings of circular phylogenetic trees using *ggtreeExtra*. More examples can be found on the chapter10 of online book¹.

1.1.1 Displaying associated data to circular phylogenetic tree using layer overlay method.

This example reproduce Fig.2 of (Morgan, Segata, and Huttenhower 2013). The data sets are provided by GraPhlAn (Asnicar et al. 2015), which contain the relative abundance of bacteria (continuous data) at different body sites (categorical data). The associated data sets were imported with *data* parameter of *geom_fruit* and displayed with the corresponding geometric function. We used heat map to display the abundance of tip species (the abundance of species was mapped to the transparency of heat map, the different body sites was mapped to the color of heat map) (Fig. S1.C), and the most abundance of species (continuous data) at specific site was visualized with bar plot (the abundance of species was mapped to the length of bar plot, the different body sites was mapped to the color of bar plot) (Fig. S1.D). The outer graphic layers can be aligned automatically according to the tip labels of phylogenetic tree. The tree annotated by *geom_fruit* can also be fully annotated with other multiple layers (high light (Fig. S1.A), clade labels (Fig. S1.B)). The layers and outer graphic layers can be added step-by-step with + (Fig. S1). More step-by-step instructions are also available at the vignette².

```
library(ggtreeExtra)
library(ggtree)
library(treeio)
library(tidytree)
library(ggstar)
library(ggplot2)
library(ggnewscale)
library(patchwork)

tree <- read.tree("../data/HMP_tree/hmptree.nwk")
# the abundance and types of microbes
dat1 <- read.csv("../data/HMP_tree/tippoint_attr.csv")
# the abundance of microbes at different body sites.
```

¹<http://yulab-smu.top/treedata-book>

²<http://bioconductor.org/packages/release/bioc/vignettes/ggtreeExtra/inst/doc/ggtreeExtra.html>

```

dat2 <- read.csv("../data/HMP_tree/ringheatmap_attr.csv")
# the abundance of microbes at the body sites of greatest prevalence.
dat3 <- read.csv("../data/HMP_tree/barplot_attr.csv")

# adjust the order
dat2$Sites <- factor(dat2$Sites, levels=c("Stool (prevalence)", "Cheek (prevalence)",
                                             "Plaque (prevalence)", "Tongue (prevalence)",
                                             "Nose (prevalence)", "Vagina (prevalence)",
                                             "Skin (prevalence)"))
dat3$Sites <- factor(dat3$Sites, levels=c("Stool (prevalence)", "Cheek (prevalence)",
                                             "Plaque (prevalence)", "Tongue (prevalence)",
                                             "Nose (prevalence)", "Vagina (prevalence)",
                                             "Skin (prevalence)"))

# extract the clade label information. Because some nodes of tree are annotated to genera,
# which can be displayed with high light using ggtree.
nodeids <- nodeid(tree, tree$node.label[nchar(tree$node.label)>4])
nodedf <- data.frame(node=nodeids)
nodelab <- gsub("[\\.\.0-9]", "", tree$node.label[nchar(tree$node.label)>4])
# The layers of clade and hightlight
poslist <- c(1.6, 1.4, 1.6, 0.8, 0.1, 0.25, 1.6, 1.6, 1.2, 0.4,
            1.2, 1.8, 0.3, 0.8, 0.4, 0.3, 0.4, 0.4, 0.4, 0.6,
            0.3, 0.4, 0.3)
labdf <- data.frame(node=nodeids, label=nodelab, pos=poslist)
# The circular layout tree.
p <- ggtree(
  tree,
  layout="fan",
  size=0.15,
  open.angle=5
) +
  geom_hilight(
    data=nodedf,
    mapping=aes(node=node),
    extendto=6.8,
    alpha=0.3,
    fill="grey",
    color="grey50",
    size=0.05
) +
  geom_cladelab(
    data=labdf,
    mapping=aes(node=node, label=label, offset.text=pos),
    barsize=NA,
    fontsize=0.7,
    angle="auto",
    hjust=0.5,
    horizontal=FALSE,
    fontface="italic"
) +
  ggtitle("A")

p1 <- p %<+% dat1 +
  geom_star(
    mapping=aes(fill=Phylum, starshape=Type, size=Size),
    starstroke=0.05
) +
  scale_fill_manual(
    values=c("#FFC125", "#87CEFA", "#7B68EE", "#808080", "#800080",

```

```

        "#9ACD32", "#D15FEE", "#FFC0CB", "#EE6A50", "#8DEEEE",
        "#006400", "#800000", "#B0171F", "#191970"),
  guide=guide_legend(keywidth = 0.5, keyheight = 0.5, order=1,
  override.aes=list(starshape=15)),
  na.translate=FALSE
) +
scale_starshape_manual(
  values=c(15, 1),
  guide=guide_legend(keywidth = 0.5, keyheight = 0.5, order=2),
  na.translate=FALSE
) +
scale_size_continuous(
  range = c(0.5, 1.5),
  guide = guide_legend(keywidth = 0.5, keyheight = 0.5, order=3,
  override.aes=list(starshape=15))
) +
new_scale_fill() +
ggtitle("B") +
theme(legend.position="none")

p2 <- p1 +
  geom_fruit(
    data=dat2,
    geom=geom_tile,
    mapping=aes(y=ID, x=Sites, alpha=Abundance, fill=Sites),
    color = "grey50",
    offset = 0.04,
    size = 0.02
) +
scale_alpha_continuous(
  range=c(0, 1),
  guide=guide_legend(keywidth = 0.3, keyheight = 0.3, order=5)
) +
scale_fill_manual(
  values=c("#0000FF", "#FFA500", "#FF0000", "#800000",
          "#006400", "#800080", "#696969"),
  guide=guide_legend(keywidth = 0.3, keyheight = 0.3, order=4)
) +
ggtitle("C") +
theme(legend.position="none")

p3 <- p2 +
  geom_fruit(
    data=dat3,
    geom=geom_bar,
    mapping=aes(y=ID, x=HigherAbundance, fill=Sites),
    pwidth=0.38,
    orientation="y",
    stat="identity"
) +
geom_treescale(fontsize=1.2, linesize=0.3, x=4.9, y=0.1) +
ggtitle("D") +
theme(legend.position="none")

p4 <- (p + p1 + plot_layout(width=c(3.4,4)))/ (p2 + p3 + plot_layout(width=c(3.4,4))) +
  plot_layout(heights=c(3,4))
p4

```

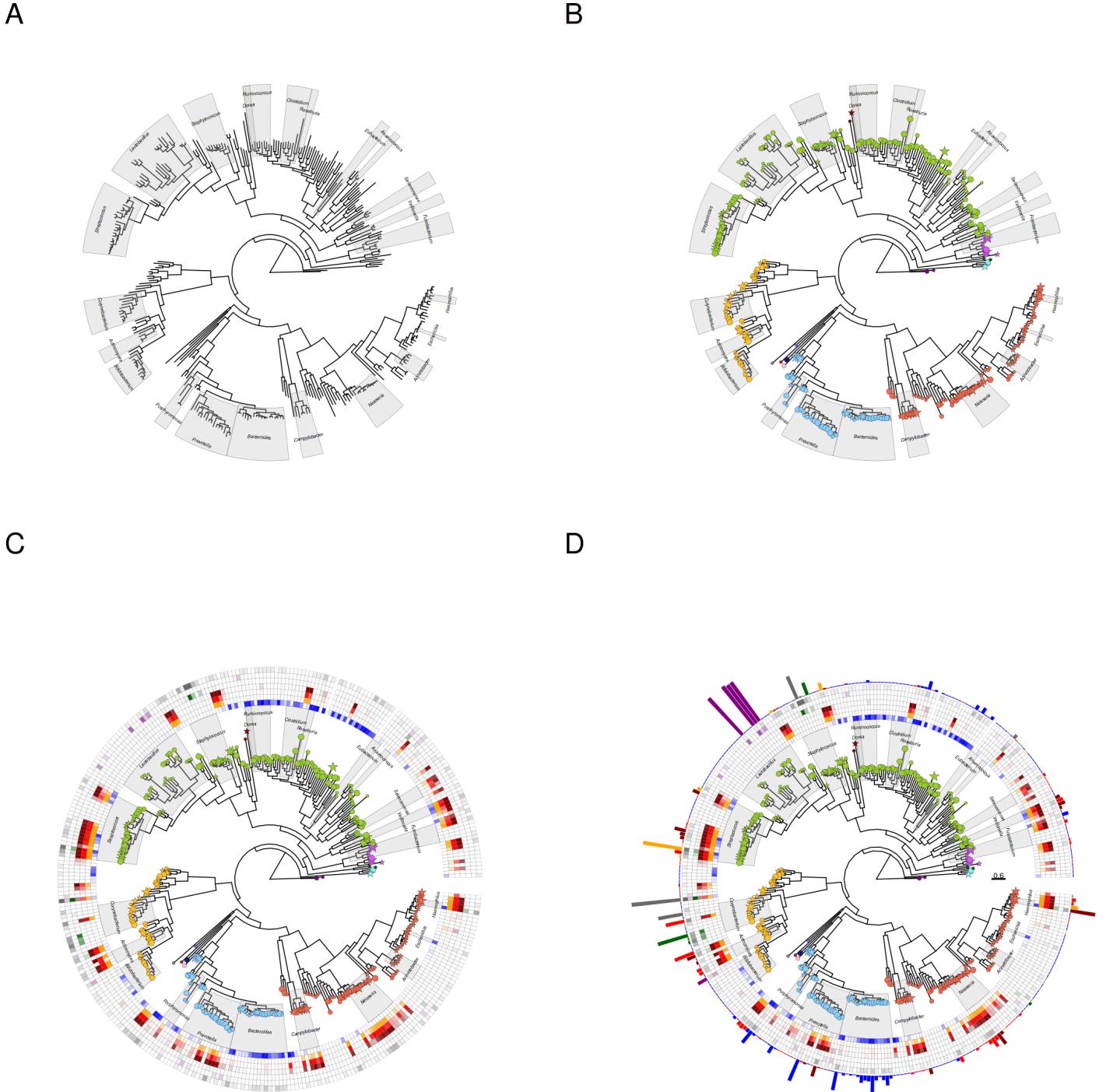


Fig. S1: The abundance of microbes at different sites of human. The shape of tip labels indicated the commensal microbes or potential pathogens. The transparency of heat map indicates the abundance of microbes, and colors of heat map indicate the different sites of human. The bar plot indicates the relative abundance at body site of the most abundance. Fig.S1.A contains the tree layer, high light layer and clade label layer; Fig.S1.B is a tip point layer added on the basis of Fig.S1.A; Fig.S1.C is a heatmap layer added on the basis of Fig.S1.B; Fig.S1.D is a bar chart layer added on the basis of Fig.S1.C. This example shows the abilities of annotate phylogenetic tree through layer overlay using *ggtree* and *ggtreeExtra*.

1.1.2 Annotating circular phylogenetic tree with the associated data contained in *ggtree* graphic object.

The *ggtree* graphic object sometimes can contain the external data and tree data, the external data can also be integrated into the tree data using %<+% of *ggtree* (Yu et al. 2018). Then variables of data integrated can also be displayed to the outer ring of circular phylogenetic tree using *geom_fruit*. We reproduce Figure 3.3 of (Smith and Wrighton 2019) to show how to integrate the associated data to tree data and visualize it. The associated data sets contain the information of

Ecosystem type, sequencing type and sample treatment method (all categorical data). The first column of external data is tip labels (the element of the column must be unique), So it can be integrated to tree data by %<+% operator of *ggtree* (Yu et al. 2018). Then *geom_fruit* can extract the tree data integrated automatically, and the attributes of data can be mapped using related attributes of geometry layers. The *y* of *aes* can be ignored, it will be added automatically. In this example, the external ring heat maps represent the different types in corresponding categories (the type of Ecosystem was mapped to the color of innermost ring heat map, the type of sequencing was mapped to the color of middle ring heat map, the type of sample treatment was mapped to the color of outermost ring heat map) (Fig. S2). And the source data is available at this repository³.

```

library(ggtreeExtra)
library(ggtree)
library(treeio)
library(ggplot2)
library(ggnewscale)
tree <- read.raxml("../data/Methanotroph/Methanotroph_rpS3_Modified_Alignment_RAxML")
# Root the tree to the archaea sequences
tree@phylo <- root(tree@phylo, node=1402, edgelabel=TRUE)
df <- read.csv("../data/Methanotroph/metadata.csv")
# reset the levels of columns to reproduce the order of original figure.
df$Specific.Ecosystem <- factor(df$Specific.Ecosystem,
                                 levels=c("Agriculture", "Alkaline/Hypersaline",
                                         "Contaminated/Wastewater", "Endosymbiont",
                                         "Freshwater", "Forest", "Geothermal",
                                         "Marine", "Natural Seep", "Peat",
                                         "Permafrost", "Wetland", "Unknown"))
df$MetaType <- factor(df$MetaType,
                       levels=c("Metatranscriptome", "Metagenome",
                               "Single-amplified genome", "Fosmid", NA))
df$Treatment <- factor(df$Treatment, levels=c("Native", "Enrichment", "Isolate", "Unknown"))

p <- ggtree(tree, layout="fan", open.angle=30)
p <- rotate_tree(p, 90)
p <- p + geom_treescaling(x=0.2, y=727*6/11, width=1, offset=20) +
  geom_nodepoint(aes(subset=as.numeric(bootstrap)>=70),
                 shape=21, fill="grey", size=1)

# we can use %<+% to integrate the external datasets to tree structure.
# and the y can not be specified.
p <- p %<+% df +
  geom_fruit(
    geom=geom_tile,
    mapping=aes(fill=Specific.Ecosystem),
    offset=0.13,
    width=0.35,
    axis.params=list(axis="none", text="Ecosystem", text.angle=0,
                     hjust=0, text.size=3, family="Times", fontface="bold")
  ) +
  scale_fill_manual(
    values=c("green3", "turquoise", "maroon", "orchid",
            "deepskyblue", "forestgreen", "salmon", "cadetblue3",
            "slategray4", "yellowgreen", "gray90", "chocolate2",
            "yellow"),
    guide=guide_legend(title="Ecosystem", keywidth=0.5, keyheight=0.5, order=4),
    na.translate=FALSE
  ) +
  new_scale_fill() +
  geom_fruit(
    geom=geom_tile,

```

³https://github.com/TheWrightonLab/Methanotroph_rpS3Analyses_SmithWrighton2018

```

mapping=aes(fill=MetaType),
offset=0.13,
width=0.35,
axis.params=list(axis="none", text="Sequencing Type", text.angle=0,
                  hjust=0, text.size=3, family="Times", fontface="bold")
) +
scale_fill_manual(
  values=c("red", "black", "dodgerblue", "gray50"),
  guide=guide_legend(title="Sequencing Type", keywidth=0.5, keyheight=0.5, order=3),
  na.translate=FALSE
) +
new_scale_fill()+
geom_fruit(
  geom=geom_tile,
  mapping=aes(fill=Treatment),
  offset=0.13,
  width=0.35,
  axis.params=list(
    axis="x",
    text="Sample Treatment",
    text.angle=0,
    hjust=0,
    text.size=3,
    family="Times",
    fontface="bold"
  )
) +
scale_fill_manual(
  values=c("red", "gray50", "black", "yellow"),
  guide=guide_legend(
    title="Sample Treatment",
    keywidth=0.5,
    keyheight=0.5,
    order=2),
  na.translate=FALSE
) +
theme(legend.background=element_rect(fill=NA), # the background of legend.
      legend.title=element_text(size=9, family="Times", face="bold"),
      legend.text=element_text(size=7, family="Times"), # the text size of legend.
      legend.spacing.y = unit(0.02, "cm"),
      legend.margin=margin(0.1, 0.9, 0.1,-0.9, unit="cm"), # t, r, b, l, cm
      legend.box.margin=margin(0.1, 0.9, 0.1, -0.9, unit="cm"),
      plot.margin = unit(c(-1.2, -1.2, -1.2, 0.1),"cm"))

# To build the external annotation of clades.
p <- p + geom_cladelabel(node=793, angle="auto", label="Gammaproteobacteria", horizontal=FALSE,
                           offset=1.4, align=T, fontsize = 3, hjust=0.5, offset.text=0.12, barsize=1,
                           family="Times", fontface="bold", color="gray30")+
  geom_cladelabel(node=791, angle="auto", label="Alphaproteobacteria", horizontal=FALSE,
                  offset=1.4, align=T, fontsize = 3, hjust=0.5, offset.text=0.12, barsize=1,
                  family="Times", fontface="bold", color="black")+
  geom_cladelabel(node=1384, label="Ca.Methylomirabilis", horizontal=TRUE,
                  offset=1.4, align=T, fontsize = 3, angle="auto", barsize=1,
                  family="Times", fontface="bold", color="gray30")+
  geom_cladelabel(node=1394, label="Methylacidiphilae", angle="auto", horizontal=TRUE,
                  offset=1.4, align=T, fontsize = 3, barsize=1,
                  family="Times", fontface="bold", color="black")+
  geom_cladelabel(node=1440, label="ANME-1", angle="auto", horizontal=TRUE,

```

```
    offset=1.4, align=T, fontsize = 3, barsize=1,  
    family="Times", fontface="bold", color="gray30") +  
geom_cladelabel(node=1405, label="ANME-2", angle="auto", horizontal=TRUE,  
    offset=1.4, align=T, fontsize = 3, barsize=1,  
    family="Times", fontface="bold", color="black")
```

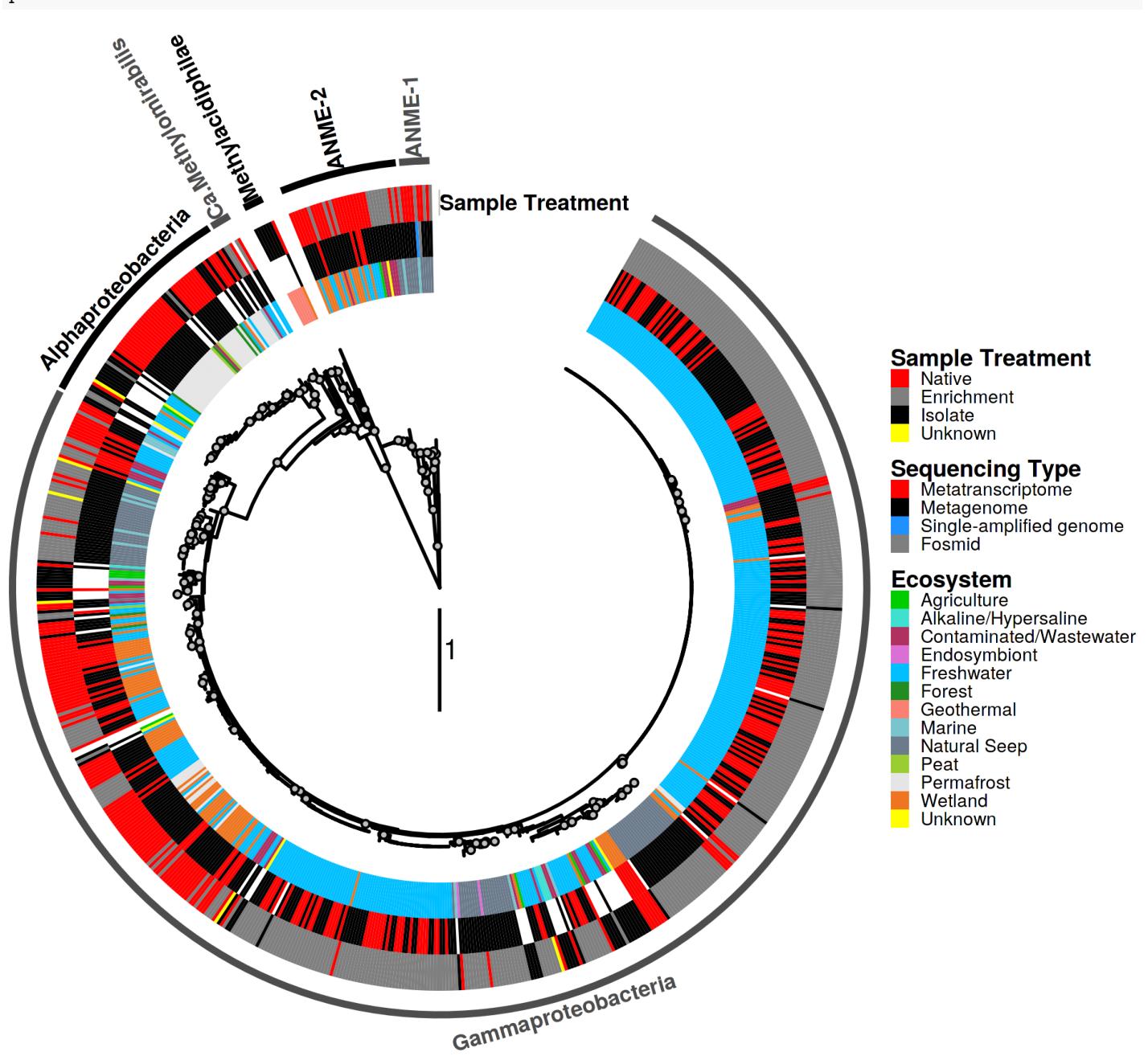


Fig. S2: Phylogeny of methanotroph ribosomal protein S39 (rpS3) genes from Figure 3.3 (Smith and Wrighton 2019). The external ring of circular tree is built with discrete datasets, which were added to tree structure in advance

1.1.3 Annotating large phylogenetic tree with multiple associated data.

The number of tips of phylogenetic tree annotated by *ggtreeExtra* is strictly not limited, except for the above-mentioned small phylogenetic tree, large phylogenetic tree can also be annotated by *ggtreeExtra*, since *ggtreeExtra* inherits the design concept of *ggtree* (Yu et al. 2017) and *ggplot2* (Wickham 2016). Each outer graphic layer of circular phylogenetic tree is independent, the associated data can be visualized simultaneously at the different outer layer. Then the layers can

be aligned and stacked on the outer rings of circular phylogenetic tree according to tree structure. These features allow ggtreeExtra to plot fully annotated large-scale phylogenetic tree with multi-dimensional data.

1.1.3.1 The first example in this section We reproduce the Fig 1 of (Segata et al. 2013), which contains a phylogenetic tree built with 3737 genomes, type of microbes genome and the number of proteins for each microbes genome. The type of genome (discrete data) is displayed in the outermost ring using scatter plot, which is provided by *ggstar* (Xu 2020) (The type of genomes was mapped to the color of point), the fraction of 400 proteins (continuous data) is visualized with bar chart (The number of proteins was mapped to the length of bar, the taxonomy of genomes was mapped to the color of bar) (Fig. S3).

```

library(ggtreeExtra)
library(ggtree)
library(treeio)
library(tidytree)
library(ggstar)
library(ggplot2)
library(ggnewscale)

tree <- read.tree("../data/PhyloPhlAn/ppal_tol.nwk")

# The attributes of tip point
da1 <- read.csv("../data/PhyloPhlAn/tipoint_attr.csv")

# The attributes of bar plot
da2 <- read.csv("../data/PhyloPhlAn/barplot_attr.csv")

# The attributes of triangle point
da3 <- read.csv("../data/PhyloPhlAn/ringpoint_attr.csv")

# to reproduce, reorder column Type
da3$Type <- factor(da3$Type, levels=c("Mislabelled", "Insertions",
                                         "Corrections", "Refinements"))

# extract the node label to annotate clade
nodelab <- tree$node.label[nchar(tree$node.label)>0]
nodeids <- nodeid(tree, nodelab)
# the position of high light extends to
extos <- c(rep(5.66, 12), 7.1, rep(5.66, 30))
nodedf <- data.frame(node=nodeids, extos=extos)

# the position of clade label
textex <- c(0.8, 0.8, 1.0, 0.6, 0, 0.8, 0.8, 0.2, -0.05, 0.5,
           0.1, 0.1, -0.1, 0.6, 0.4, 0, 1.2, 0.8, 1.0, 0.6,
           0.6, 1.0, 0.8, 0.2, -0.1, 1.2, 1, 1, 0.6, 0.6,
           0.4, 0.5, 1.2, 0.6, 1.4, 1.3, 0, -0.1, 0.8, 0.6,
           0.8, 1.4, 2.4)

cladedf <- data.frame(node=nodeids, label=nodelab, pos=textex)

# To reproduce the color of original figures, we generated the same colors,
# user can custom their colors.
phylacolor <- c(Proteobacteria="#9ACD32", Firmicutes="#EE6A50",
                Actinobacteria="#87CEFA", Bacteroidetes="#FFC125",
                Euryarchaeota="#D15FEE", Tenericutes="#8DEEEE",
                Cyanobacteria="#800000", Spirochaetes="#006400",
                Chlamydiae="#800080", Crenarchaeota="#808080",
                Fusobacteria="#FFC0CB", Thermi="#B0171F", Other="Black",
                Chloroflexi="#191970", Thermotogae="#7B68EE",

```

```

Aquificae="#00CD00", Synergistetes="#8B4513",
Chlorobi="#BC8F8F", Verrucomicrobia="#303030",
Planctomycetes="#8E8E38", Acidobacteria="#CDCDC1")

tipcolors <- c("#9ACD32", "#EE6A50", "#87CEFA", "#FFC125", "#D15FEE",
               "#8DEEEE", "#800000", "#006400", "#800080", "#808080",
               "#FFC0CB", "#B0171F", "#191970", "#7B68EE", "#00CD00",
               "#8B4513", "#BC8F8F", "#303030", "#8E8E38", "#CDCDC1")

names(tipcolors) <- c("Proteobacteria", "Firmicutes", "Actinobacteria",
                      "Bacteroidetes", "Euryarchaeota", "Tenericutes",
                      "Cyanobacteria", "Spirochaetes", "Chlamydiae",
                      "Crenarchaeota", "Fusobacteria", "Thermi",
                      "Chloroflexi", "Thermotogae", "Aquificae",
                      "Synergistetes", "Chlorobi", "Verrucomicrobia",
                      "Planctomycetes", "Acidobacteria")

# We built a circular layout tree with open a small (6) angle.
p <- ggtree(
  tree,
  layout="fan",
  open.angle=6,
  size=0.1
) +
  geom_hilight(
    data=nodedf,
    mapping=aes(
      node=node,
      extendto=extos
    ),
    alpha=0.3,
    fill="grey",
    color="grey50",
    size=0.05) +
  geom_cladelab(
    data=cladedf,
    mapping=aes(
      node=node,
      label=label,
      offset.text=pos
    ),
    barsize=NA,
    extend=0,
    fontsize=1.2,
    angle="auto",
    hjust=0.5,
    horizontal=FALSE,
    fontface="italic"
  )
p <- p %<+% da1 +
  geom_tippoint(
    mapping=aes(fill=Phylum),
    size=1.2,
    shape=21,
    stroke=0.05,
    show.legend=FALSE
) +

```

```

scale_fill_manual(values=tipcolors) +
new_scale_fill() +
geom_fruit(
  data=da2,
  geom=geom_bar,
  mapping=aes(
    x=Abundance,
    y=ID,
    fill=Phyla
  ),
  offset=-0.2,
  pwidth=0.1,
  stat='identity',
  orientation="y"
) +
scale_fill_manual(
  values=phylacolor,
  guide=guide_legend(
    keywidth=0.6,
    keyheight=0.6,
    order=1,
    ncol=1
  )
) +
new_scale_fill() +
geom_fruit(
  data=da3,
  geom=geom_star,
  mapping=aes(
    x=Pos,
    y=ID,
    fill=Type,
    size=Pos
  ),
  starshape=26,
  alpha=0.8,
  starstroke=0,
  offset=0.024,
  pwidth=0.008
) +
scale_fill_manual(
  values=c("blue", "black", "green", "red"),
  guide=guide_legend(
    keywidth=1, keyheight=1, order=2,
    override.aes=list(alpha=1, size=2)
  ),
  na.translate=FALSE
) +
scale_size_continuous(range=c(1.4,2.2), guide="none") +
geom_treescale(fontsize=1.2, linesize=0.3) +
theme(legend.position=c(0.94, 0.5),
      legend.title=element_text(size=7),
      legend.background=element_rect(fill=NA),
      legend.text=element_text(size=6),
      legend.spacing.y = unit(0.02, "cm"))

```

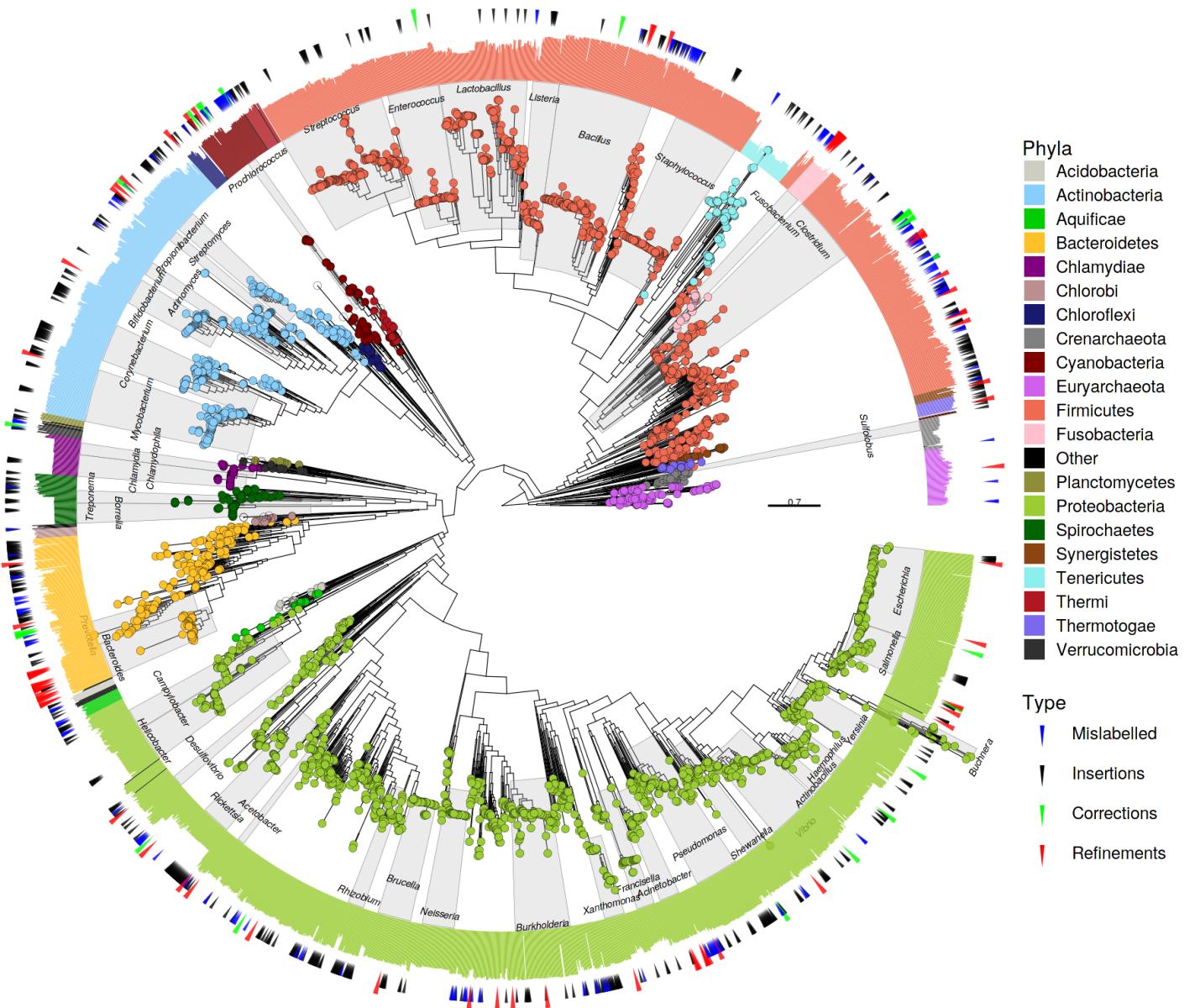


Fig. S3: The bacterial and archaeal (3737 genomes) phylogenetic tree built with 400 conserved proteins. The tip points are built using discrete data, the different colors represent the different phyla. The external bar is created using continuous data, the length represents the fraction of the 400 proteins contained in each genome. The external triangle is also created using type of microbes genome (discrete data), the different colors represent the different types of genomes (mislabelled and confidently replaced see Fig 1 of (Segata et al. 2013).

1.1.3.2 The second example in this section This example reproduces Fig 2 of (Asnicar et al. 2015). The phylogenetic tree also was built using 3737 microbes (Segata et al. 2013). The associated data contains present or not (discrete data) and type (discrete data) of target gene, the type (discrete data) and capability (continuous data) of fatty acid metabolism, and the length of microbes genome. The present or not of different target gene was visualized with first inner heat map (the type of target gene was mapped to color of heat map, and the subtype of target gene was mapped to the x value(take on numerical values)). The type and capability of fatty acid metabolism was also visualized with middle heat map (the type was mapped to color of heat map, the capability of fatty acid metabolism was mapped to the transparency of heat map). The length of genome was displayed with bar chart (the length of genome was mapped to length of bar, the phylum information of genome was mapped to the color of bar). (Fig. S4).

```
library(ggtreeExtra)
library(ggtree)
library(treeio)
library(tidytree)
```



```

"#00CD00", "Black")

p <- ggtree(tree, layout="circular", size=0.1) + highlights

p <- p +
  geom_fruit(
    data=dt1,
    geom=geom_point,
    mapping=aes(y=ID, fill=Phyla),
    shape=21,
    size=1.2,
    stroke=0.05,
    position="identity",
    show.legend=FALSE
  ) +
  scale_fill_manual(values=colors) +
  cladelabels +
  new_scale_fill() +
  geom_fruit(
    data=dt2,
    geom=geom_tile,
    mapping=aes(y=ID, x=ring, fill=Type1),
    offset=-0.02,
    pwidth=0.14,
    addbrink=TRUE
  ) +
  scale_fill_manual(
    name="ATP synthesis",
    values=c("#339933", "#dfac03"),
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=1)
  ) +
  new_scale_fill() +
  geom_fruit(
    data=dt3,
    geom=geom_tile,
    mapping=aes(y=ID, alpha=Abundance, x=Type2, fill=Type2),
    offset=0.001,
    pwidth=0.18
  ) +
  scale_fill_manual(
    name="Fatty Acid metabolism",
    values=c("#b22222", "#005500", "#0000be", "#9f1f9f", "#793a07"),
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=2)
  ) +
  scale_alpha_continuous(range=c(0, 0.4),
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=3)) +
  new_scale_fill() +
  geom_fruit(data=dt4,
    geom=geom_bar,
    mapping=aes(y=ID, x=Length, fill=Phyla),
    stat="identity",
    orientation="y",
    pwidth=0.3,
    position=position_dodge())
  scale_fill_manual(values=colors,
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=4)) +
  geom_treescale(fontsize=1.2, linesize=0.3) +
  theme(legend.position=c(0.95, 0.5),

```

```

legend.background=element_rect(fill=NA),
legend.title=element_text(size=7),
legend.text=element_text(size=6),
legend.spacing.y = unit(0.02, "cm"))

```

p

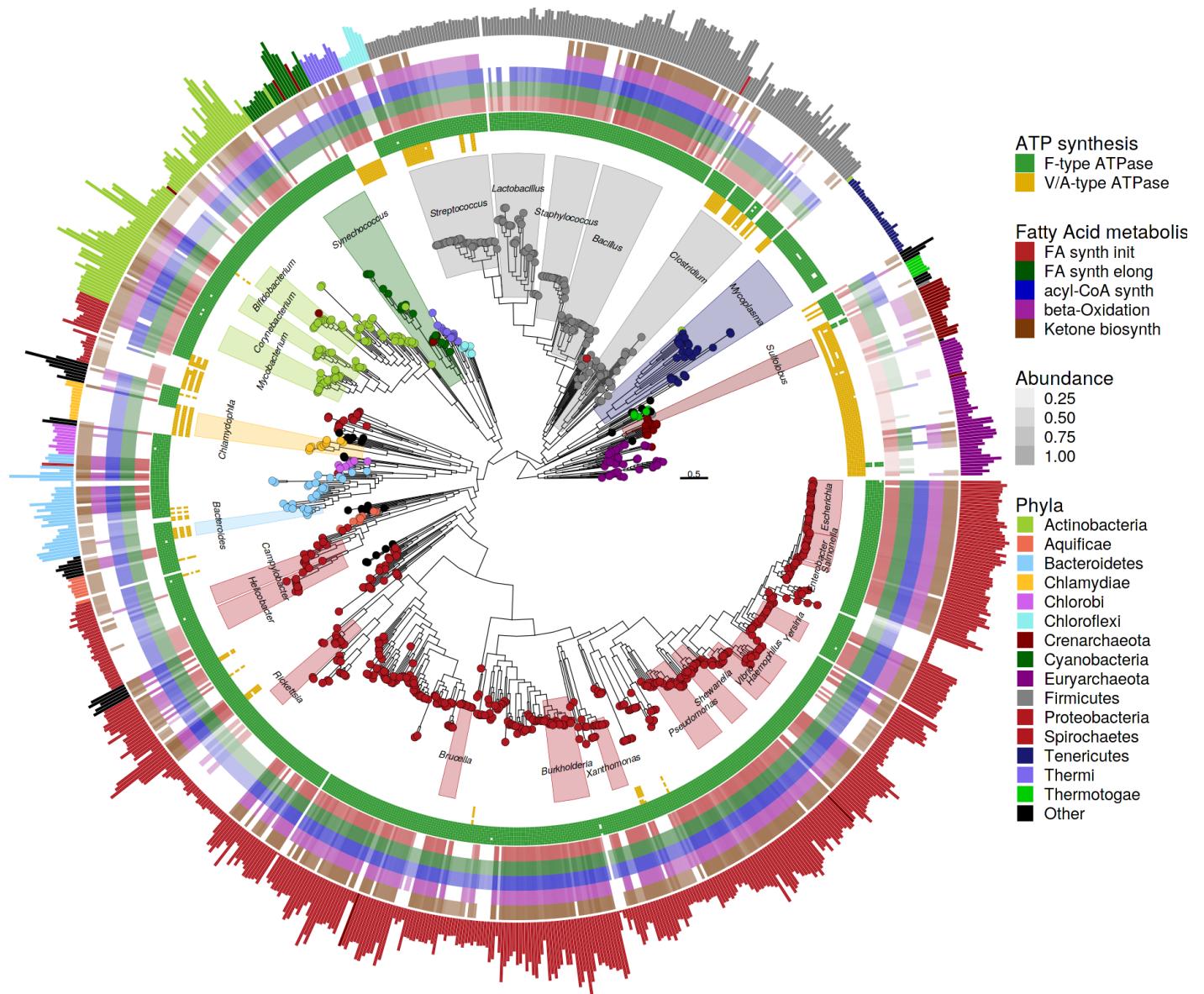


Fig. S4: The phylogenetic tree built using in (Segata et al. 2013) using 3737 microbial genomes. The first and second ring heat map were built with discrete data, it represents the presence or absence of each module, the other heat map rings were created with continuous data, the transparency represents the capability of fatty acid metabolism, the different colors represent the types of fatty acid metabolism. The length of bar represents the genome length of corresponding microbes.

1.1.4 Obtaining more power to present associated data by working with some packages

The *geom_fruit* not only can work with some *geom* function defined in *ggplot2* (Wickham 2016), but also can work with *geom* function defined in *ggplot2*-based packages. This ability will make *gtreeExtra* more power to present associated data. We use an example from Fig.1 of (Song et al. 2020). The *geom_fruit* can work with *geom_phylropic* defined in *ggimage* (Yu 2020) to display the image data on the circular phylogenetic tree. The attributes (such as color, size) of subplot can also be mapped to variables of external dataset. This is often not available in the existing tools. The source data of the

example is available at this repository⁴. More example that ggtreeExtra works with other ggplot2-based packages were list in supplementary file2.

```

library(ggtree)
library(treeio)
library(ggplot2)
library(ggtreeExtra)
library(tidytree)
library(ggnewscale)
library(ggimage)

tr <- read.tree("../data/VertebrateGutMicrobiomes/annotated_host_tree.tre")
corda <- read.csv("../data/VertebrateGutMicrobiomes/mantel.jaccard.pearson.csv")
corda$r <- abs(corda$r)
barda <- read.csv("../data/VertebrateGutMicrobiomes/data_diet_bar.csv", check.names=F)
barda <- reshape2::melt(barda, id.vars="ID", variable.name="Diet", value.name="mete")
barda$Diet <- factor(barda$Diet, levels=c("Fruit", "Invertebrates",
                                             "Nectar", "Plants", "Scavenging",
                                             "Seeds", "Meat (Ectotherms)",
                                             "Meat (Endotherms)",
                                             "Meat (Fish)", "Meat (Unknown)"))

cladeda <- read.csv("../data/VertebrateGutMicrobiomes/data_clade_class.csv", check.names=F)
cladeda$id <- nodeid(tr, cladeda$id)
cladeda$class <- factor(cladeda$class, levels=c("Amphibia", "Chelonia", "Lepidosauria",
                                                 "Crocodylomorpha", "Aves", "Mammalia"))

flightda <- read.csv("../data/VertebrateGutMicrobiomes/data_flight_bar.csv")

phylopicda <- read.csv("../data/VertebrateGutMicrobiomes/data_phylopic_uid.csv")
phylopicda$class <- factor(phylopicda$class, levels=c("Amphibia", "Chelonia", "Lepidosauria",
                                                       "Aves", "Mammalia"))

p <- ggtree(tr, layout="fan", open.angle=15)

p <- p %<+% corda
p$data$width <- ifelse(is.na(p$data$r), 0.1, 0.6)
r <- NULL
p <- p +
  aes(color=r, size=I(width)) +
  scale_colour_viridis_c(
    name="Mantel Correlation",
    option="C",
    guide=guide_colorbar(
      barheight=0.6,
      order=4,
      title.position="top",
      label.position="bottom",
      direction="horizontal"
    )
  )

p1 <- p +
  geom_fruit(
    data=barda,
    geom=geom_bar,
    mapping=aes(x=mete, y=ID, fill=Diet),
    orientation="y",
  )

```

⁴https://github.com/tanaes/tetrapod_microbiome_analysis

```

    stat="identity",
    colour=NA,
    pwidth=0.25,
    offset=0.008
) +
scale_fill_manual(
  values=c("#a6cee3", "#cab2d6",
          "#1f78b4", "#33a02c",
          "#6a3d9a", "#b2df8a",
          "#fb9a99", "#e31a1c",
          "#ff7f00", "#fdbf6f"),
  guide=guide_legend(keywidth=0.5, keyheight=0.5, order=1)
)

p2 <- p1 +
  new_scale_colour() +
  geom_cladelab(
    data=cladeda,
    mapping=aes(node=id, label=class, colour=class),
    textcolour=NA,
    barsize=4,
    extend=0.2,
    offset=100) +
  scale_colour_manual(
    name="Host Class",
    values=c("#b2df8a", "#33a02c", "#fb9a99",
            "#e31a1c", "#EACB47", "#6a3d9a"),
    guide=guide_legend(
      keywidth=0.5,
      keyheight=0.5,
      order=2,
      override.aes=list(size=3, alpha=1)
    )
  )

p3 <- p2 +
  new_scale_fill() +
  geom_fruit(
    data=flightda,
    geom=geom_tile,
    mapping=aes(y=ID, fill=flight),
    size=0,
    width=14,
    offset=0.11,
    pwidth=0.4,
  ) +
  scale_fill_manual(
    name="Flight Status",
    values=c("black", "white"),
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=3,
                       override.aes=list(color="black", size=0.3)))
p4 <- p3 +
  new_scale_colour() +
  geom_fruit(
    data=phylopicda,
    geom=geom_phylopic,
    mapping=aes(y=taxa, image=uid, color=class),
    size=0.035,

```

```

    offset=0.16,
    alpha=0.8,
    position=position_identityx()
) +
scale_colour_manual(
  values=c("#b2df8a", "#33a02c", "#fb9a99",
          "#EACB47", "#6a3d9a"),
  guide="none"
) +
theme(
  legend.background=element_rect(fill=NA),
  legend.title=element_text(size=9),
  legend.text=element_text(size=6.6),
  legend.spacing.y = unit(0.02, "cm")
)
)

```

p4

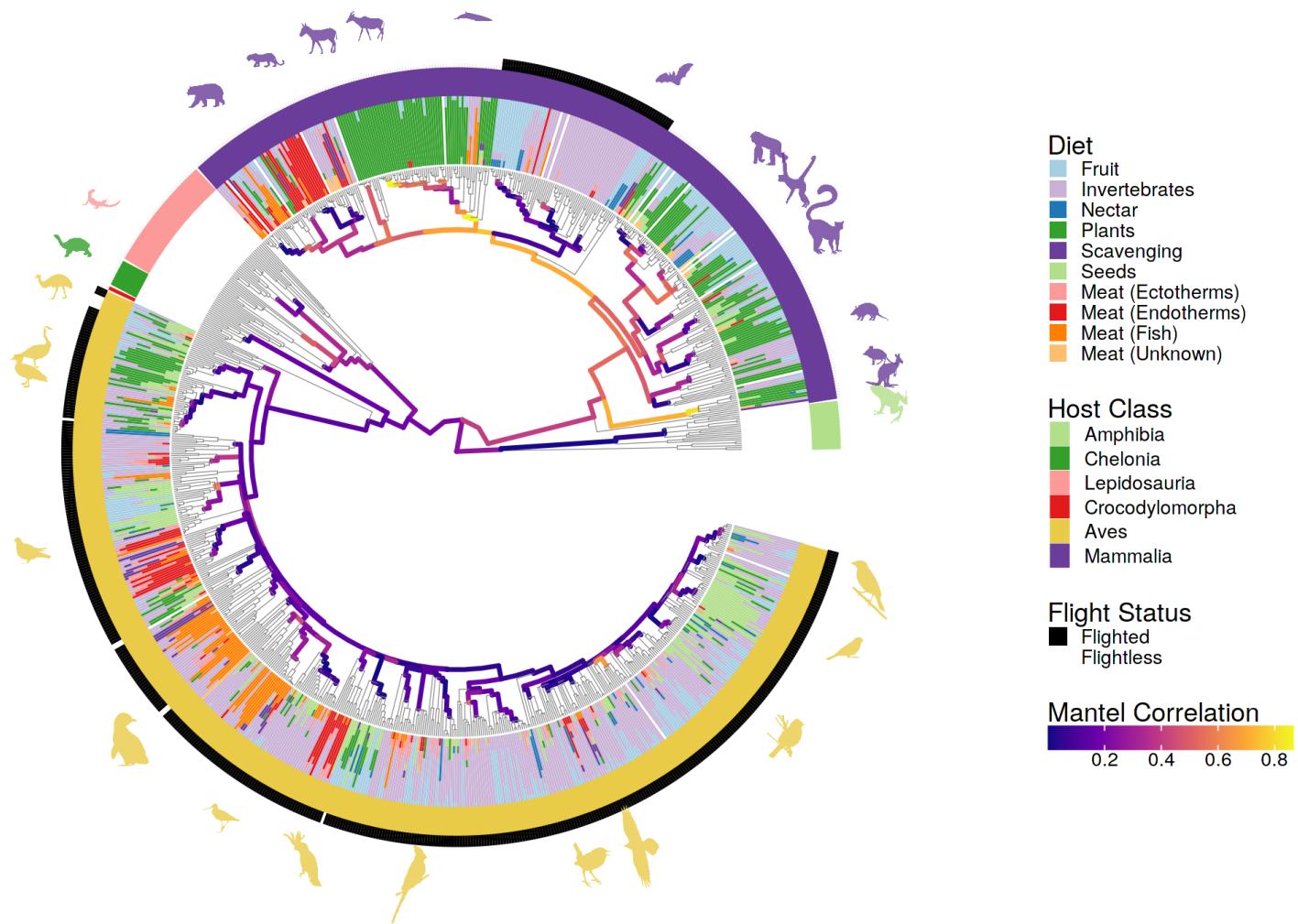


Fig. S5: Host tree was obtained from TimeTree (Kumar et al. 2017). The branch colour represents the Mantel Pearson correlation. The stack bar chart represents the diet composition of host. The inner ring represents the host taxonomic class, and some representative species are also displayed in the outermost circle. The middle ring represents flight status.

1.1.5 Annotating associated data to the phylogenetic tree combined relationship data

In fact, in addition to circular phylogenetic tree, *ggtreeExtra* can annotate other layout tree. Especially, It can annotate inward circular phylogenetic tree, which can combine relationship data, such as the correlation of species or genes, horizontal

gene transfer and syntetic linkage. This feature can give user more convenience and reproducibility to visualize and annotate phylogenetic trees with relationship data for better exploring phylogenetic patterns behind multi-dimensional data. Here, we combine the Fig.1b and Fig.2 of (Helfrich et al. 2018) show the directional interactions and the biosynthetic potential of isolates from *Arabidopsis* leaf microbiome in phylogenetic tree. The interactions of isolates are visualized in the line of inner. The biosynthetic potential of isolates are displayed with heat map (the number of target gene (continuous data) was mapped to the transparency of heat map, the type of target gene (discrete data) was mapped to the x value and color of heat map). The number of interactions of inhibitions or sensitivities per strain is displayed with stacked bar (the number of interactions (continuous data) was mapped to the x value (length) of bar plot, and the type of interactions (discrete data) was mapped to the color of bar plot.). We found some strains from Firmicutes and from Grammaproteobacteria have more inhibitor interactions. However, many strains from Alphaproteobacteria and Betateobacteria prefer the interaction of sensitivity. In addition, These strains that prefer the interactions of sensitivity might be have more BGCs from ribosomally synthesized and post-translationally modified peptide (*RiPP*).

```

library(ggtree)
library(ggtreeExtra)
library(ggplot2)
library(MicrobiotaProcess)
library(ggstar)
library(ggnewscale)
library(grid)

alltax <- read.csv("../data/Arabidopsis_leaf_microbiome/all_stain_taxonomy.csv")
linktab <- read.csv("../data/Arabidopsis_leaf_microbiome/Interaction_link_tab.csv")
weighttab <- read.csv("../data/Arabidopsis_leaf_microbiome/Interaction_weight.csv")
tippoint <- read.csv("../data/Arabidopsis_leaf_microbiome/stain_tippoint.csv")
BGCsda <- read.csv("../data/Arabidopsis_leaf_microbiome/BGCs_heatmap.csv")

tippoint$Taxa <- factor(tippoint$Taxa,
                           levels=c("Actinobacteria",
                                   "Bacteroidetes",
                                   "Firmicutes",
                                   "Deinococcus-Thermus",
                                   "Alphaproteobacteria",
                                   "Betaproteobacteria",
                                   "Gammaproteobacteria"
                           )
)
tippoint$names <- gsub("s_Leaf","",tippoint$Isolation)

BGCsda$BGCs <- factor(BGCsda$BGCs,
                        levels=c("modular.PKS",
                                "modular.PKS.NRPS.hybrid",
                                "non_modular.PKS", "NRPS",
                                "RiPP",
                                "Quorum.sensing",
                                "terpene",
                                "other"
                        )
)
BGCsda$Count <- log10(BGCsda$Count+1)
BGCsda$Count <- ifelse(BGCsda$Count==0, NA, BGCsda$Count)

trda <- convert_to_treedata(alltax)
p <- ggtree(
    trda,
    layout="inward_circular",
    size=0.2,
    xlim=c(18,NA)
)

```

```

p <- p %<+% tippoint

p1 <- p +
  geom_tippoint(
    mapping=aes(
      color=Taxa,
      shape=Level
    ),
    size=1,
    alpha=0.8
  ) +
  scale_color_manual(values=c("#EF3B2C", "#1D91C0", "#FEB24C", "grey60",
                             "#7FBC41", "#4D9221", "#276419"),
                     guide=guide_legend(
                       keywidth=0.5,
                       keyheight=0.5,
                       order=2,
                       override.aes=list(shape=c("Actinobacteria"=20,
                                                 "Bacteroidetes" =20,
                                                 "Firmicutes" =20,
                                                 "Deinococcus-Thermus" =20,
                                                 "Alphaproteobacteria" =18,
                                                 "Betaproteobacteria" =18,
                                                 "Gammaproteobacteria" =18
                                               )),
                     size=2
                   ),
  na.translate=TRUE
  )
) +
scale_shape_manual(values=c("Phylum"=20, "Class"=18), guide="none" )

p2 <- p1 +
  new_scale_color() +
  geom_taxalink(
    data=linktab,
    mapping=aes(
      taxa1=Inhibitor,
      taxa2=Sensitive,
      color=Interaction
    ),
    alpha=0.6,
    offset=0.1,
    size=0.15,
    ncp=10,
    hratio=1,
    arrow=grid::arrow(length = unit(0.005, "npc"))
  ) +
  scale_colour_manual(values=c("chocolate2", "#3690C0", "#009E73"),
                      guide=guide_legend(
                        keywidth=0.8, keyheight=0.5,
                        order=1, override.aes=list(alpha=1, size=0.5)
                      )
  )
)

p3 <- p2 +
  geom_fruit(
    data=BGCsda,

```

```

    geom=geom_tile,
    mapping=aes(y=Strain, x=BGCs, alpha=Count, fill=BGCs),
    offset=-0.9,
    pwidth=1,
    size=0.02,
    color = "grey50"
) +
scale_alpha_continuous(range=c(0.1, 1),
                       name=bquote(paste(Log[10], .(("Count+1"), "))),
                       guide=guide_legend(keywidth = 0.4, keyheight = 0.4, order=4)
) +
scale_fill_manual(
  values=c("#66C2A5", "#FC8D62", "#8DA0CB", "#E78AC3",
          "#A6D854", "#FFD92F", "#E5C494", "#B3B3B3"),
  guide=guide_legend(keywidth = 0.4, keyheight = 0.4, order=3)
)

p4 <- p3 +
  geom_tiplab(
    mapping=aes(label=names),
    align=TRUE,
    size=1,
    linetype=NA,
    offset=7.8
  )

p5 <- p4 +
  new_scale_fill() +
  geom_fruit(
    data=weighttab,
    geom=geom_bar,
    mapping=aes(
      x=value,
      y=Strain,
      fill=Number
    ),
    stat="identity",
    orientation="y",
    offset=0.48,
    pwidth=2,
    axis.params=list(
      axis="x",
      text.angle=-45,
      hjust=0,
      vjust=0.5,
      nbreak=4,
    )
  ) +
  scale_fill_manual(
    values=c("#E41A1C", "#377EB8", "#4DAF4A", "#984EA3"),
    guide=guide_legend(keywidth=0.5, keyheight=0.5, order=5)
) +
theme(
  legend.background=element_rect(fill=NA),
  legend.title=element_text(size=6.5),
  legend.text=element_text(size=5),
  legend.spacing.y = unit(0.02, "cm"),
  legend.margin=margin(0.1, 0.9, 0.1,-0.9, unit="cm"),

```

```

legend.box.margin=margin(0.1, 0.9, 0.1, -0.9, unit="cm"),
plot.margin = unit(c(-1.2, -1.2, -1.2, 0.1),"cm")
)

```

p5

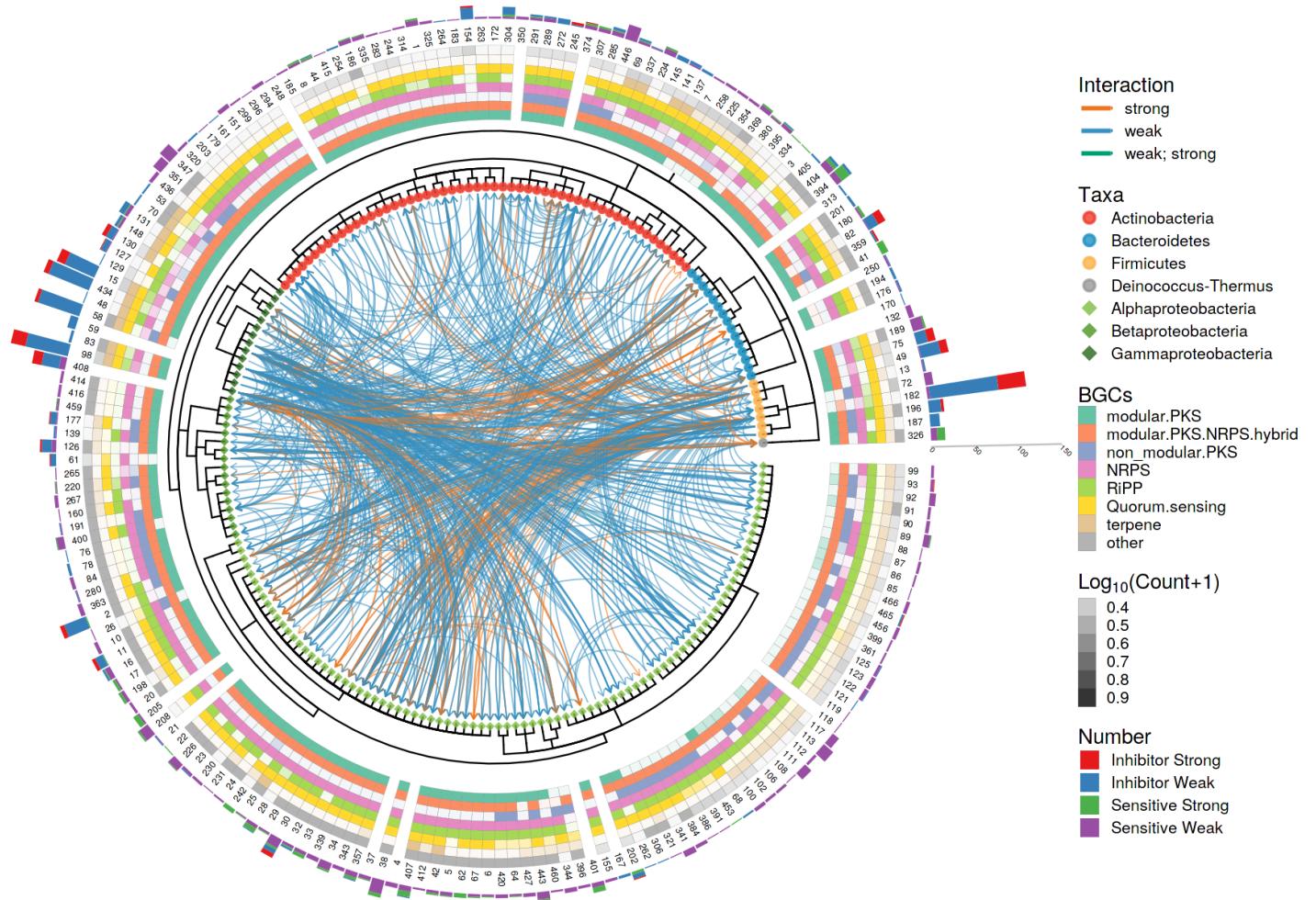


Fig. S6: The Phylogenetic tree of isolates from *Arabidopsis* leaf microbiome (Bai et al. 2015). The lines in inner represent the directional inhibitory interactions, the direction of arrow represents from inhibitor to sensitive, and the color of line represents the weak or strong interactions. The color of tip point represents the taxonomy annotation, circle represents the phylum, square represents the class from Proteobacteria. The heat map of external ring represents the number of detected BGCs (biosynthesis gene clusters). The ring from inner to outside represents the Modular polyketide synthase, Modular PKS non-ribosomal peptide synthetase (NRPS) hybrid, Non modular PKS, NRPS, ribosomally synthesized and post-translationally modified peptide (RiPP), Quorum sensing, Terpene, and Others. The external bar represent the number of interactions of inhibitions or sensitivities per strain.

1.2 Geometric layers that supported by *geom_fruit* of *ggtreeExtra*

As mentioned above, *geom_fruit* is a general function for linking multiple geometric graph to tree. It supports multiple layout type tree, such as circular layout, inward circular layout, rectangular layout etc. It can re-order the input data based on tree structure and displayed the data at the external of tree with the geometric function, which was defined in *ggplot2* (Wickham 2016) and other *ggplot2* extension packages. Here is the list of the geometric functions that work seamlessly with *geom_fruit* (Table S1). Since the *ggplot2* (Wickham 2016) community keeps expanding and more *geom* layers will be implemented in either *ggplot2* (Wickham 2016) or other extensions, *ggtreeExtra* also will support more *geom* layers and will gain more power to explore data in future. Several examples that *ggtreeExtra* works with other geometric function has been listed in supplementary file2.

Table S1: List of geometric layers supported by 'geom_fruit()'

Package	Geom Layer	Description
ggdist	geom_dotsinterval	creates dots, intervals, and quantile dotplots
	geom_pointinterval	creates point and multiple uncertainty interval
	geom_slab	creates slab geom
	geom_slabinterval	creates slab, point and interval meta-geom
ggimage	geom_image	visualizes image files
	geom_phylopic	queries image files from phylopic database and visualizes them
ggpattern	geom_bar_pattern	draws bar charts with support for pattern fills
	geom_boxplot_pattern	draws box and whiskers plot with support for pattern fills
	geom_col_pattern	draws bar charts using 'stat_identity()' with support for pattern fills
	geom_tile_pattern	draws rectangle by using the center of the tile and its size with support for pattern fills
ggplot2	geom_bar	draws bar charts
	geom_boxplot	draws box and whiskers plot
	geom_col	draws bar charts using 'stat_identity()'
	geom_label	draws a rectangle behind the text
	geom_point	creates scatterplots
	geom_raster	a high performance special case for all the tiles are the same size
	geom_text	adds text to the plot
	geom_tile	draws rectangle by using the center of the tile and its size
ggrepel	geom_text_repel	adds text to the plot. The text labels repel away from each other and away from the data points
	geom_label_repel	draws a rectangle underneath the text. The text labels repel away from each other and away from the data points
ggstance	geom_barch	horizontal version of 'geom_bar()'
	geom_boxplotb	horizontal version of 'geom_boxplot()',
	geom_colb	horizontal version of 'geom_col()'
ggstar	geom_star	creates scatterplots
ggsymbol	geom_symbol	creates scatterplots
scatterpie	geom_scatterpie	creates scatter pie plot

1.3 Adding further annotation to specific layers

When the annotation information should be add to same panel, *ggtreeExtra* provides *geom_fruit_list* to add multiple layers to the same external ring of circular tree (Fig S7).

```
library(ggtreeExtra)
library(ggtree)
library(treeio)
library(ggplot2)
set.seed(1024)
tree <- rtree(100)
df <- data.frame(id=tree$tip.label, value=sample(seq_len(200), 100, replace=TRUE),
                  group=c(rep("A",50),rep("B",50)))

p <- ggtree(tree, layout="fan", open.angle=180) + xlim(-2, NA)

p <- p + geom_fruit_list(geom_fruit(data=df,
                                       geom=geom_bar,
                                       mapping=aes(x=value, y=id, fill=group),
                                       orientation="y",
                                       stat="identity"),
                           geom_fruit(data=df,
                                       geom=geom_text,
                                       mapping=aes(x=value, y=id, label=value),
                                       size=2,
                                       offset=0.055,
                                       pwidth=0.26,
                                       position=position_identityx(),
                                       )) +
  scale_fill_manual(values=c("chocolate2", "#009E73"))
```

```
guide=guide_legend(keywidth = 0.5, keyheight = 0.5))
```

p

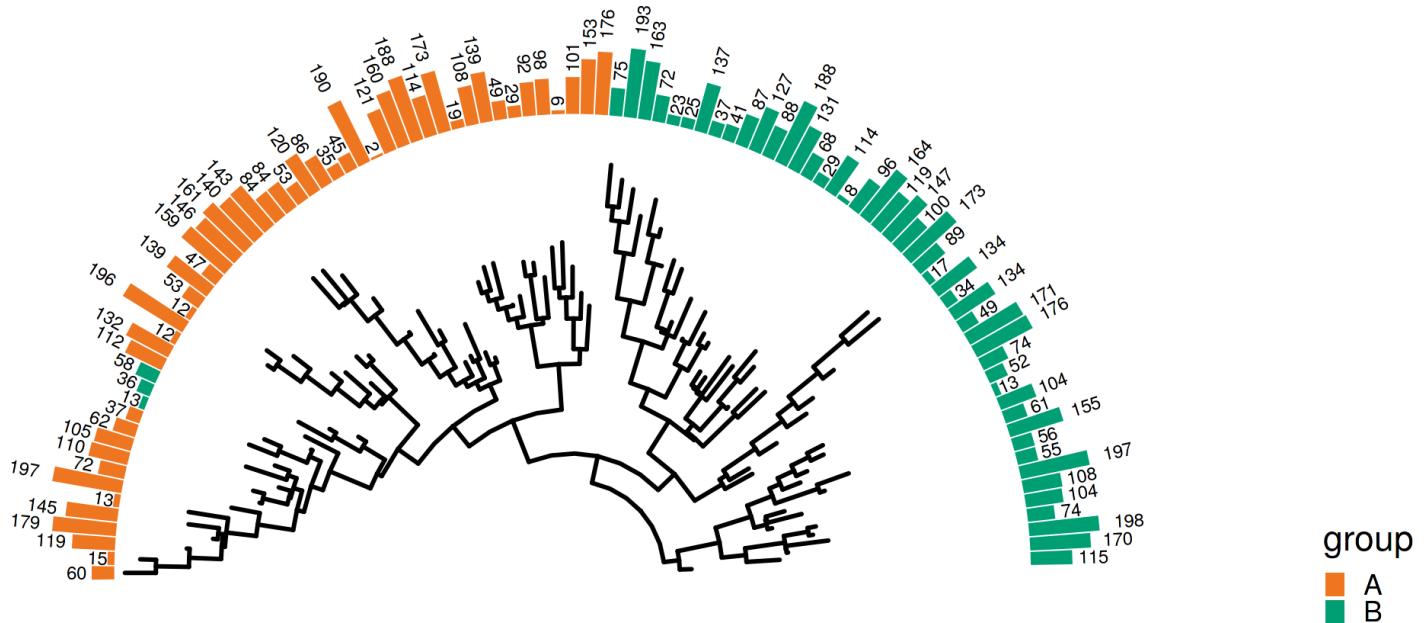


Fig. S7: visualizing multiple layers at them same external ring using *ggtreeExtra*

1.4 Summary

As far as we known, None of R packages support visualizing external data and align the plot to circular phylogenetic tree. There are some tools depending on python or other platform to visualize phylogenetic tree, they provide some features to annotate and visualize circular phylogenetic tree, but they are not general and still have some shortcomings. For example, *GraPhlAn* (Asnicar et al. 2015) and *ETE3* (Huerta-Cepas, Serra, and Bork 2016), both based on python, *GraPhlAn* (Asnicar et al. 2015) provided few graph layers to display associated data, while *ETE3* (Huerta-Cepas, Serra, and Bork 2016) did not support visualizing associated dataset on the outer ring of circular phylogenetic tree. There are some web tools support visualizing phylogenetic tree, such as *iTOL* (Letunic and Bork 2019), *Microreact* (Argimón et al. 2016) and *EvolView* (Subramanian et al. 2019). But the feature to link tree with external datasets is often missing, and they often need user to provide predefined data types, which make it tedious to use and debug. They are also not conducive to integration into the relevant data analysis pipeline.

Here, we developed *ggtreeExtra*, which is a general tool, based on grammar of graphics (Wilkinson 2012). So the variables of associated data can be mapped to the attributes of geometric layer, the layers can be added step-by-step. Then the graphic layers can be stacked and aligned on the outer of circular phylogenetic tree (Fig. S1), these features allow *ggtreeExtra* to plot a fully annotated tree with complex graphic layers easily. In addition, the internal data of tree data or data added by %<+% operator can also be visualized to the outer ring of circular phylogenetic tree with *geom_fruit* (Fig. S2), which will facilitate of annotation of more associated data. Furthermore, *ggtreeExtra* can also be used to annotate huge phylogenetic tree (Fig. S3 and S4), which will facilitate of visualization of large-scale data and multi-dimensional data. Moreover, *ggtreeExtra* is designed to link *ggtree* (Yu et al. 2017) and geometric functions defined in *ggplot2* (Wickham 2016) or other ggplot2-based packages, which can enhance the power of *ggtreeExtra* to present associated data (Fig. S5). Additionally, *ggtreeExtra* also support other layout tree built by *ggtree* (Yu et al. 2017), so it can work with phylogenetic tree combined relationship data using inward circular (Fig. S6), which will help to explore and visualize multi-dimensional data.

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