

## Outputs inspection half CIFAR100

```
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.0.5

library(dplyr)

## Warning: package 'dplyr' was built under R version 4.0.5

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##       filter, lag

## The following objects are masked from 'package:base':
##       intersect, setdiff, setequal, union

library(tidyr)

## Warning: package 'tidyr' was built under R version 4.0.5

library("ggpubr")

## Warning: package 'ggpubr' was built under R version 4.0.5

library(LDATS)

## Warning: package 'LDATS' was built under R version 4.0.5

library(ggVennDiagram)

## Warning: package 'ggVennDiagram' was built under R version 4.0.5

library(stringr)
library(abind)

## Warning: package 'abind' was built under R version 4.0.3
```

```

source("utils.R")

## Warning: package 'hash' was built under R version 4.0.5

## hash-2.2.6.1 provided by Decision Patterns

## Warning: package 'reticulate' was built under R version 4.0.5

## Warning: package 'berryFunctions' was built under R version 4.0.5

## 
## Attaching package: 'berryFunctions'

## The following object is masked from 'package:ggVennDiagram':
## 
##     circle

## The following object is masked from 'package:dplyr':
## 
##     between

## Warning: package 'purrr' was built under R version 4.0.3

## Warning: package 'reshape2' was built under R version 4.0.3

## 
## Attaching package: 'reshape2'

## The following object is masked from 'package:tidyverse':
## 
##     smiths

```

Visualization on CIFAR100. We are using data of three neural networks trained on reduced CIFAR100 training set. Half of the CIFAR100 training set was extracted as a validation set. We then divided both the reduced training set and validation set into 5 disjoint subsets and trained an ensemble on each of them. This was done in 10 replications, each time with random split of the training set into validation and new training set. In this visualization, we are trying to inspect the outputs deeper, mainly to make sense of strange behavior of nll metric for ensemble outputs.

```

base_dir <- "../data/data_train_val_half_c100"
repls <- 0:9
folds <- 0:4
classes <- 100

nets_outputs <- load_network_outputs(base_dir, repls)
ens_outputs_gathered <- load_ensemble_outputs(base_dir, repls, folds, gather=(1 + nets_outputs$test_labels))
net_results <- read.csv(file.path(base_dir, "net_accuracies.csv"))
ens_results <- read.csv(file.path(base_dir, "ensemble_accuracies.csv"))

```

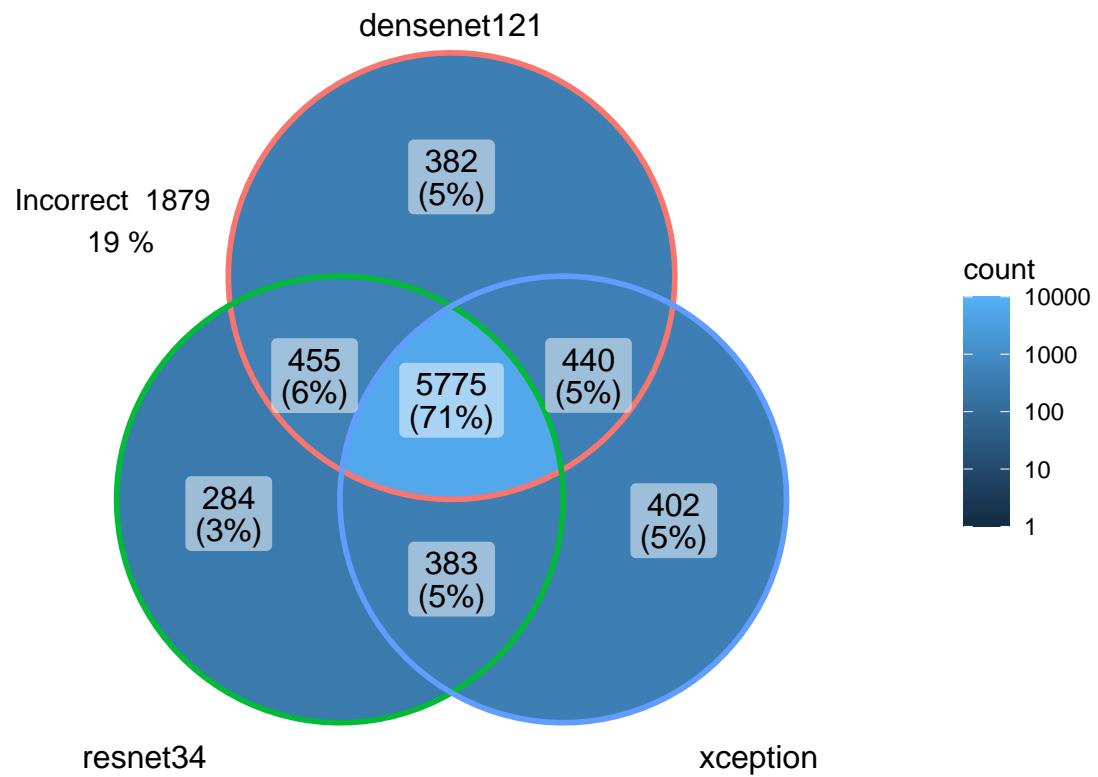
```

sort_ind <- function(lst)
{
  return(sort(lst, index.return=TRUE, decreasing=TRUE)$ix)
}
nets_test_top_indices <- apply(X=nets_outputs$test_outputs, MARGIN=c(1, 2, 3), FUN=sort_ind)[1, , , ]
r_n <- length(repls)
samples_n <- dim(nets_outputs$test_labels)[2]
nets_n <- length(nets_outputs$networks)
test_labs <- nets_outputs$test_labels + 1
dim(test_labs) <- c(r_n, 1, samples_n)
test_labs <- aperm(abind(array(rep(aperm(test_labs, perm=c(2, 1, 3)), nets_n), c(r_n, samples_n, nets_n))
nets_test_cor_preds <- test_labs == nets_test_top_indices

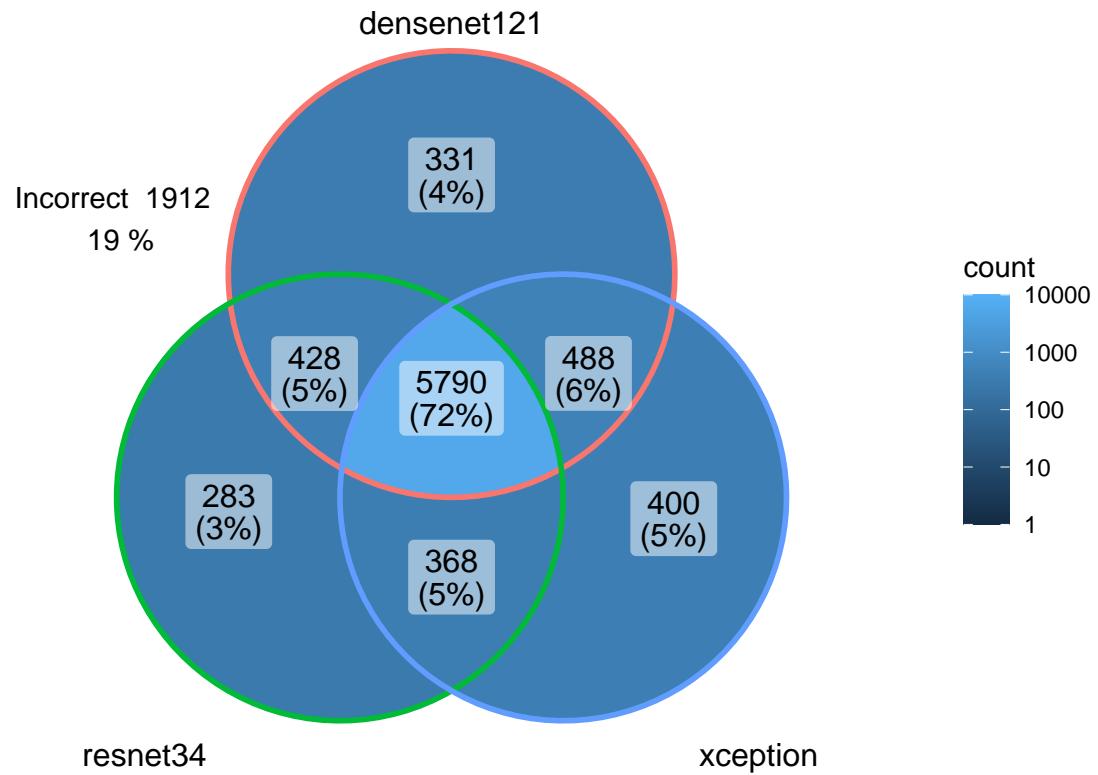
for (ri in 1:r_n)
{
  nets_cor_list <- list()
  incor <- 1:samples_n
  for (ni in 1:nets_n)
  {
    cor_list <- which(nets_test_cor_preds[ri, ni, ])
    nets_cor_list[[nets_outputs$networks[ni]]] = cor_list
    incor <- setdiff(incor, cor_list)
  }
  incor_n <- length(incor)
  venn_diag <- ggVennDiagram(nets_cor_list) + scale_fill_gradient(trans="log10", name="count", limits=c(-8, 10))
  annotate(geom="text", x=-4, y=5, label=paste("Incorrect ", incor_n, "\n", round(incor_n / samples_n)))
  ggtitle(paste("Correct predictions by network - replication ", ri)) +
    scale_x_continuous(limits=c(-8, 10))
  print(venn_diag)
}

```

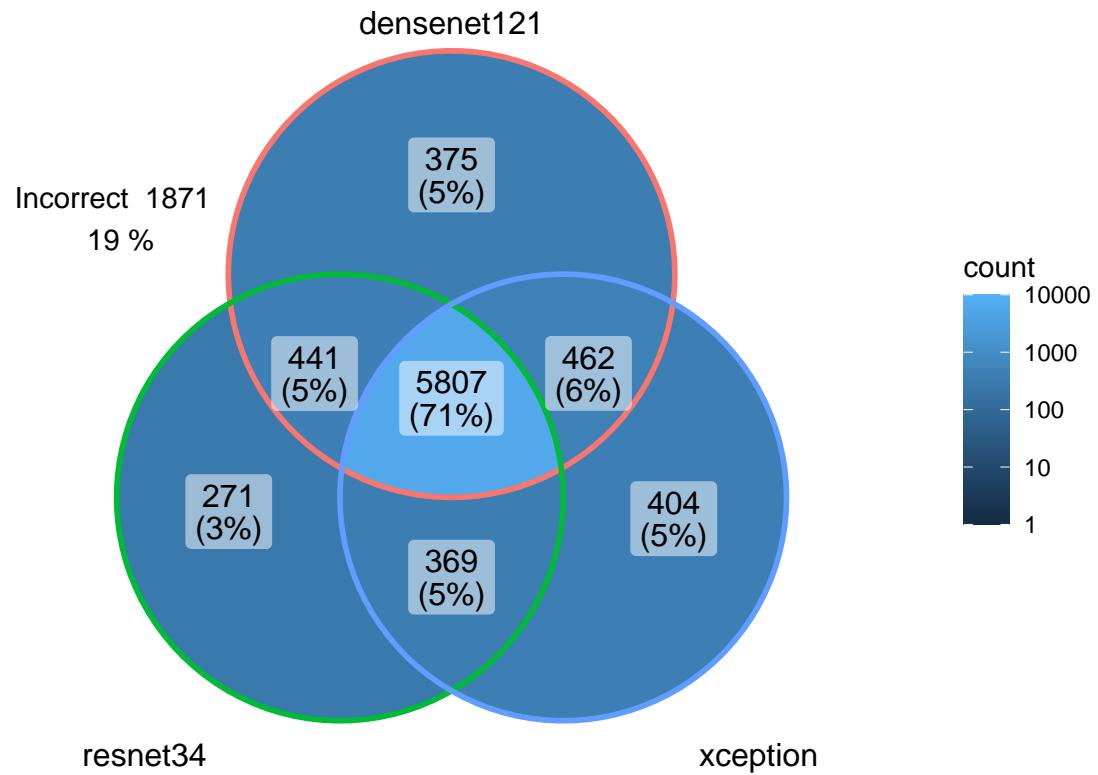
## Correct predictions by network – replication 1



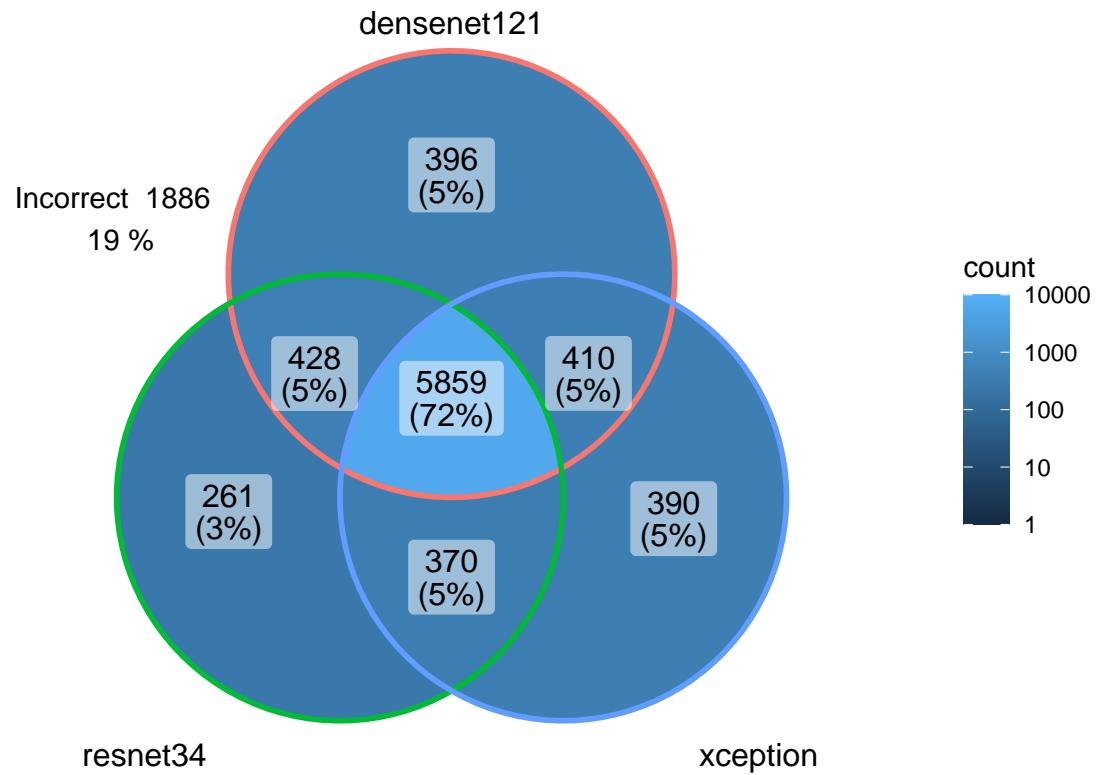
## Correct predictions by network – replication 2



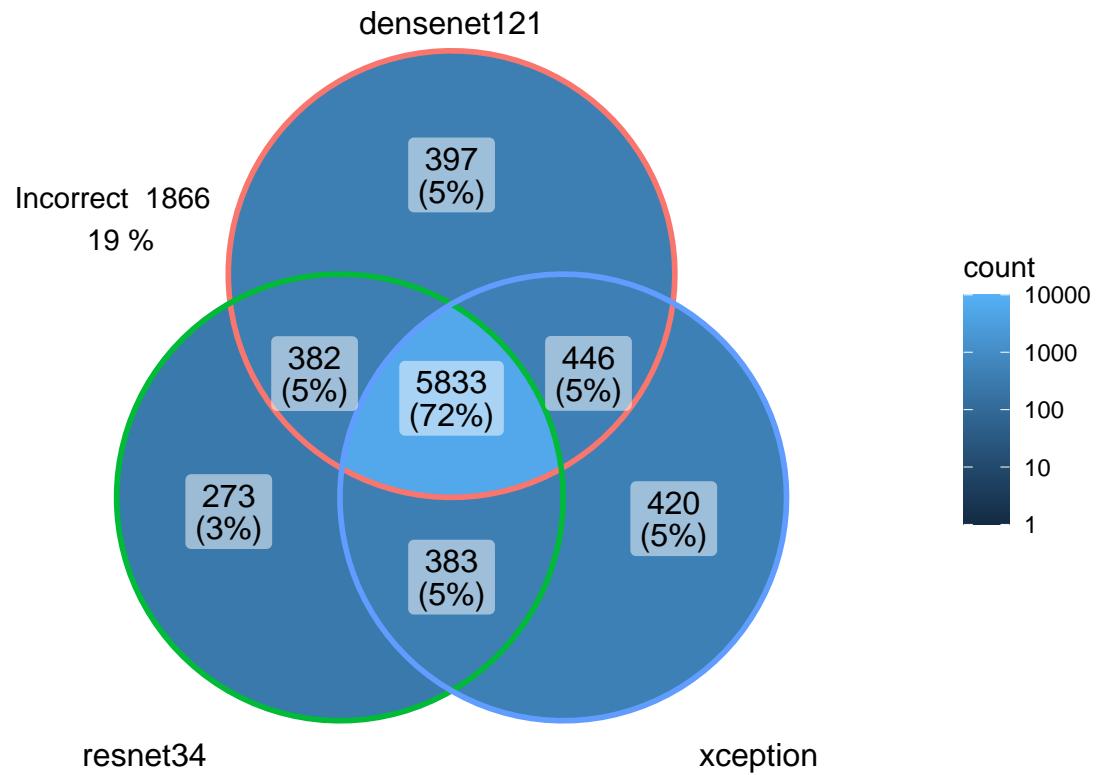
## Correct predictions by network – replication 3



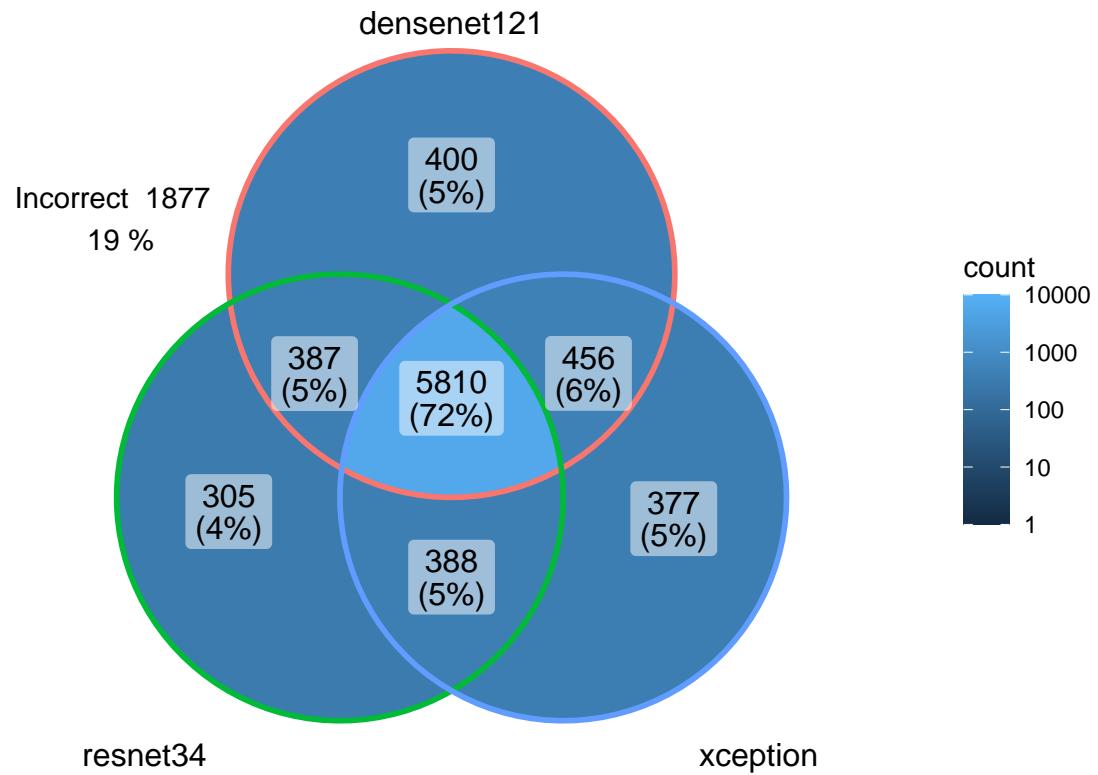
## Correct predictions by network – replication 4



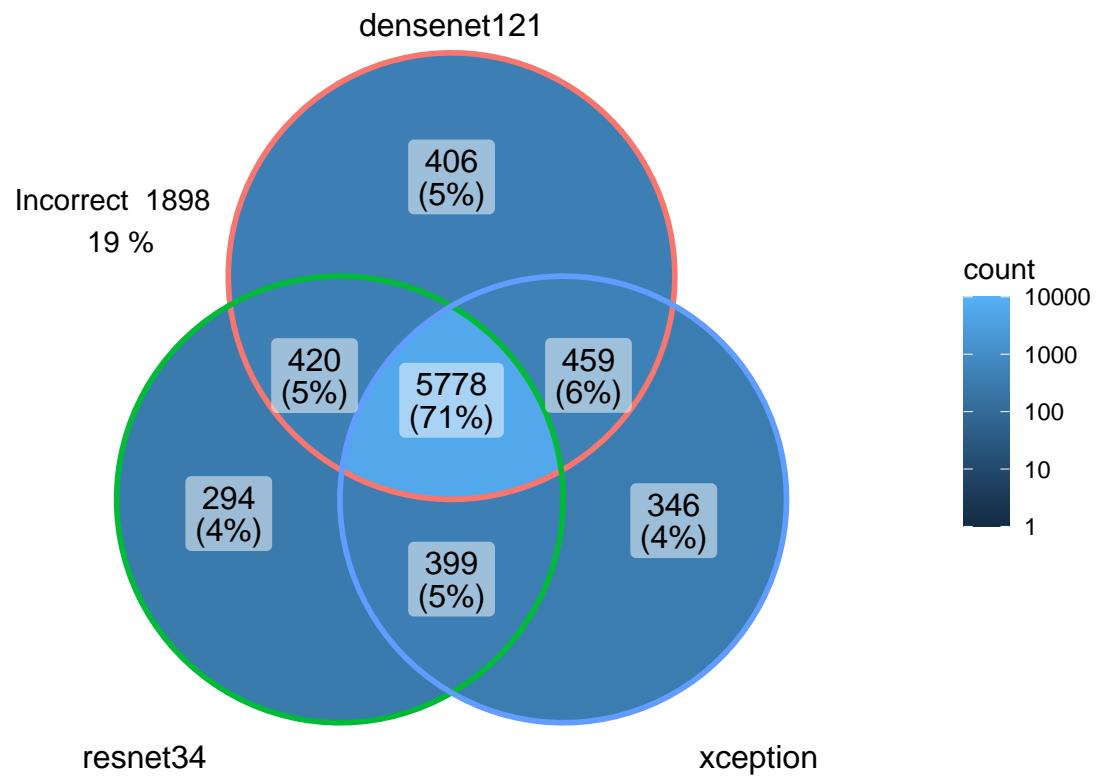
## Correct predictions by network – replication 5



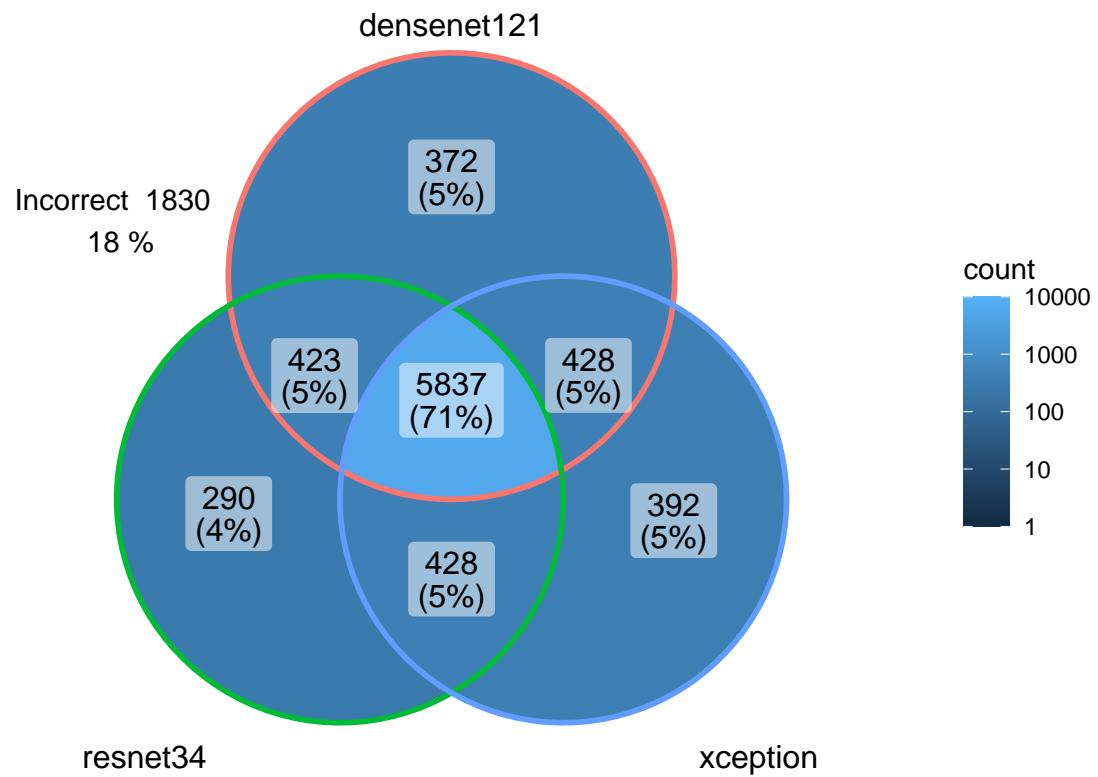
## Correct predictions by network – replication 6



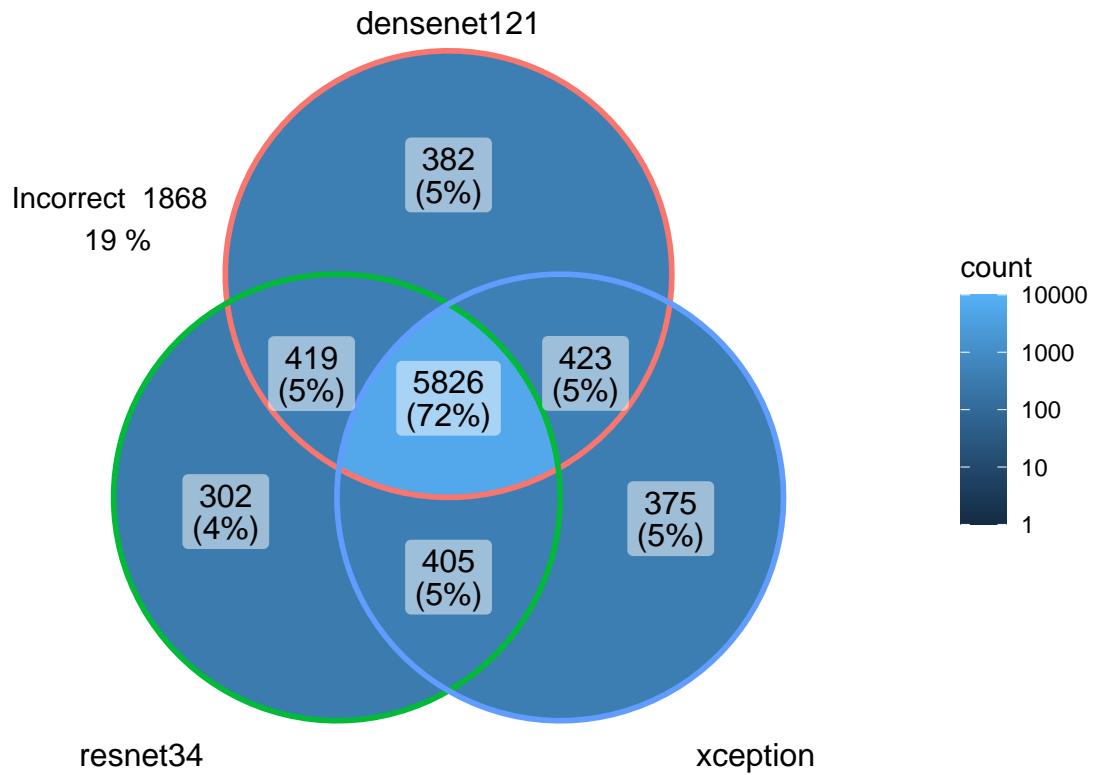
## Correct predictions by network – replication 7



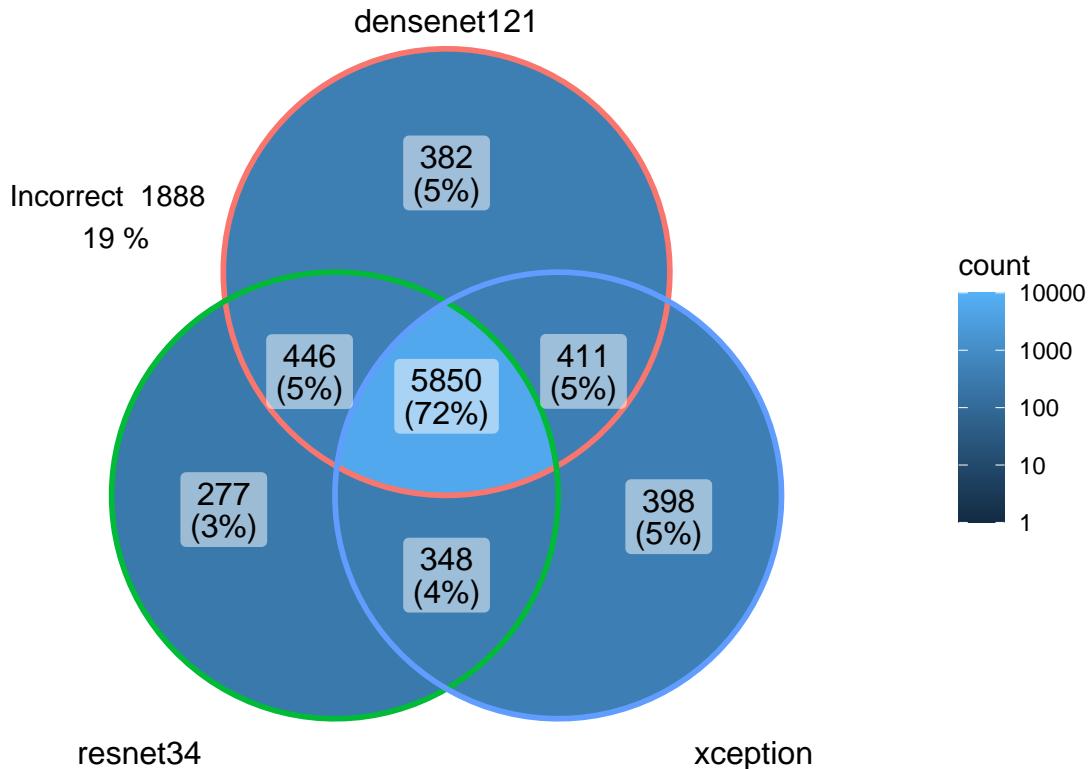
## Correct predictions by network – replication 8



## Correct predictions by network – replication 9



## Correct predictions by network – replication 10



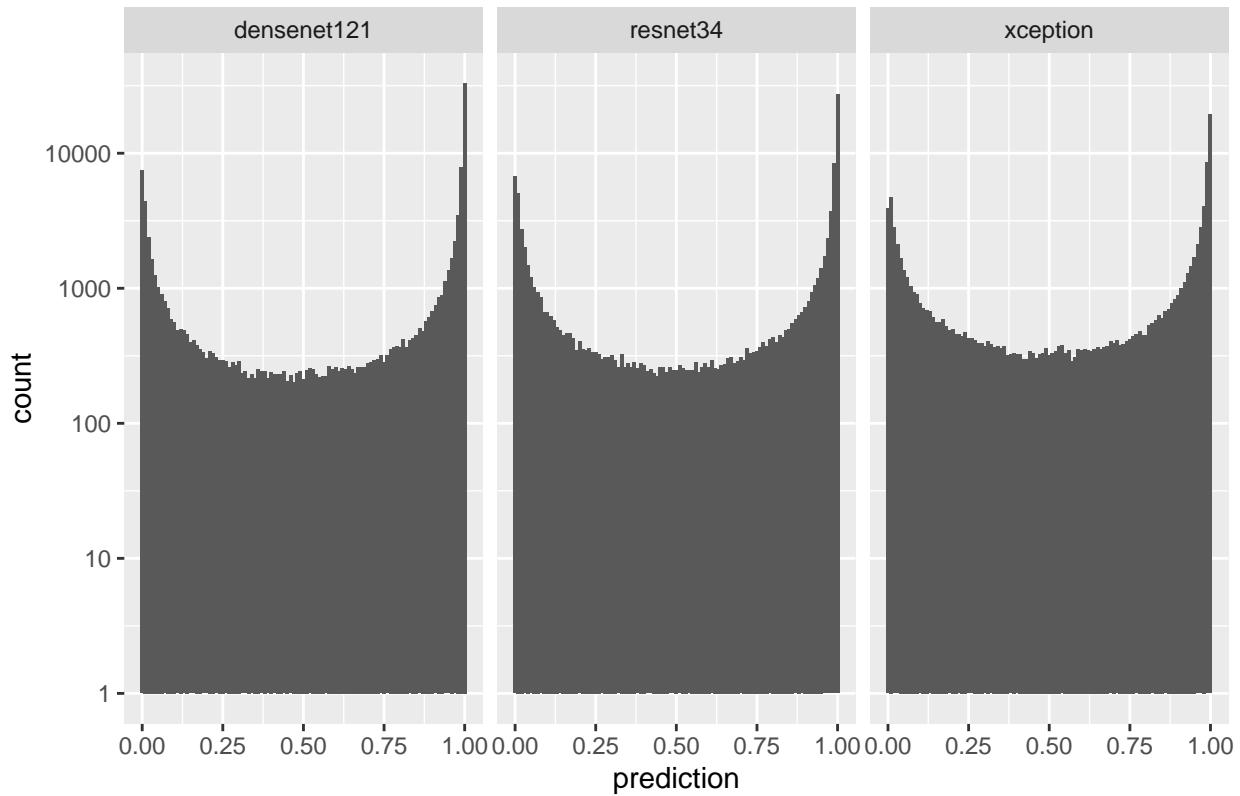
Compared to CIFAR10 examples, here are fewer samples correctly classified by all the networks. For one network exclusive correct classifications densenet and xception perform better than resnet. Pairs of network all have similar values with xception, resnet lagging a bit.

```

preds <- nets_outputs$test_outputs
for (ri in repls + 1)
{
  for (net_i in seq_along(nets_outputs[["networks"]]))
  {
    preds[ri, net_i, ,] <- softmax(preds[ri, net_i, , ])
  }
}
nets_test_cor_probs <- gather(preds, 1 + nets_outputs$test_labels[, ], 3, 4)
nets_test_cor_probs <- melt(nets_test_cor_probs)
nets_test_cor_probs <- nets_test_cor_probs[, c(-3, -4)]
names(nets_test_cor_probs) <- c("replication", "network", "prediction")
nets_test_cor_probs$network <- as.factor(nets_test_cor_probs$network)
levels(nets_test_cor_probs$network) <- nets_outputs$networks

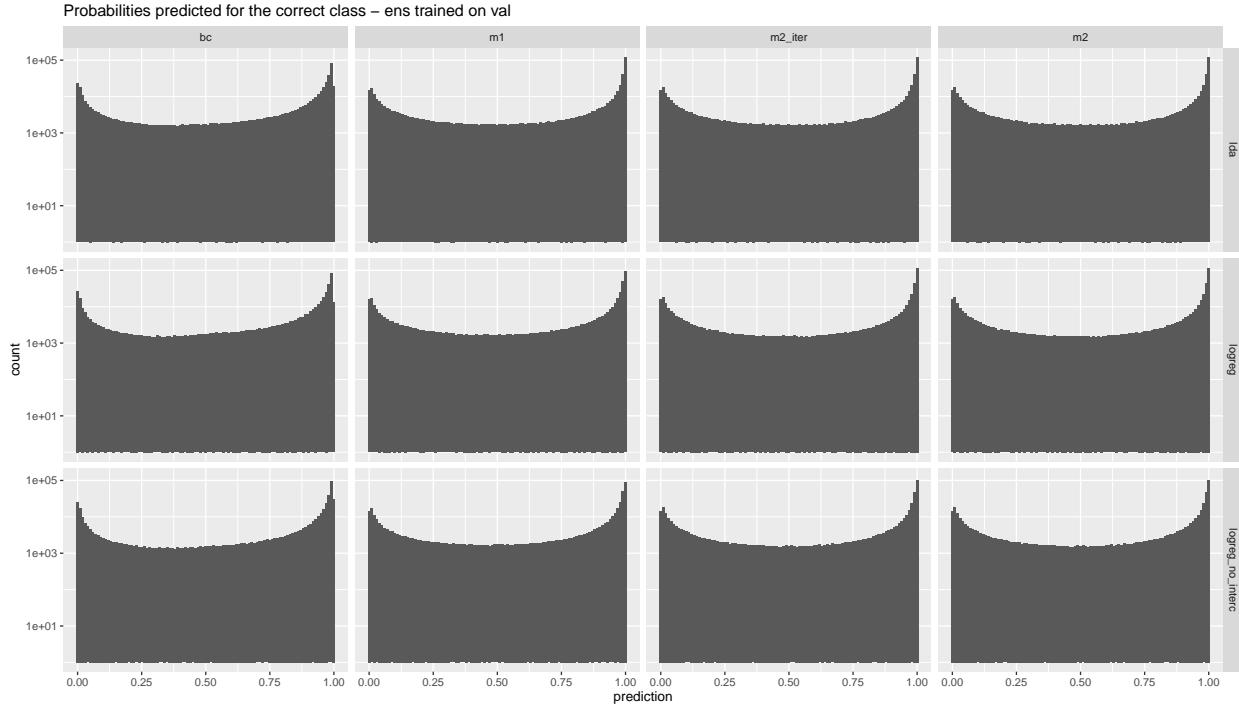
nets_cor_preds_histo <- ggplot(data=nets_test_cor_probs) + geom_histogram(mapping=aes(x=prediction), bins=100)
  ggtitle("Histograms of predicted probability for the correct class") + facet_wrap(~network) + scale_y_continuous(breaks=c(0, 1))
  
```

## Histograms of predicted probability for the correct class



```
val_ens_cor_probs <- melt(ens_outputs_gathered$val_training)
val_ens_cor_probs <- val_ens_cor_probs[, c(-5, -6)]
names(val_ens_cor_probs) <- c("replication", "combining_method", "coupling_method", "fold", "prediction")
val_ens_cor_probs[, c("combining_method", "coupling_method")] <- lapply(val_ens_cor_probs[, c("combining_
levels(val_ens_cor_probs$combining_method) <- ens_outputs_gathered$combining_methods
levels(val_ens_cor_probs$coupling_method) <- ens_outputs_gathered$coupling_methods

val_ens_cor_preds_histo <- ggplot(data=val_ens_cor_probs) + geom_histogram(mapping=aes(x=prediction), b
```



```
sum(val_ens_cor_probs$prediction <= 0)
```

```
## [1] 0
```

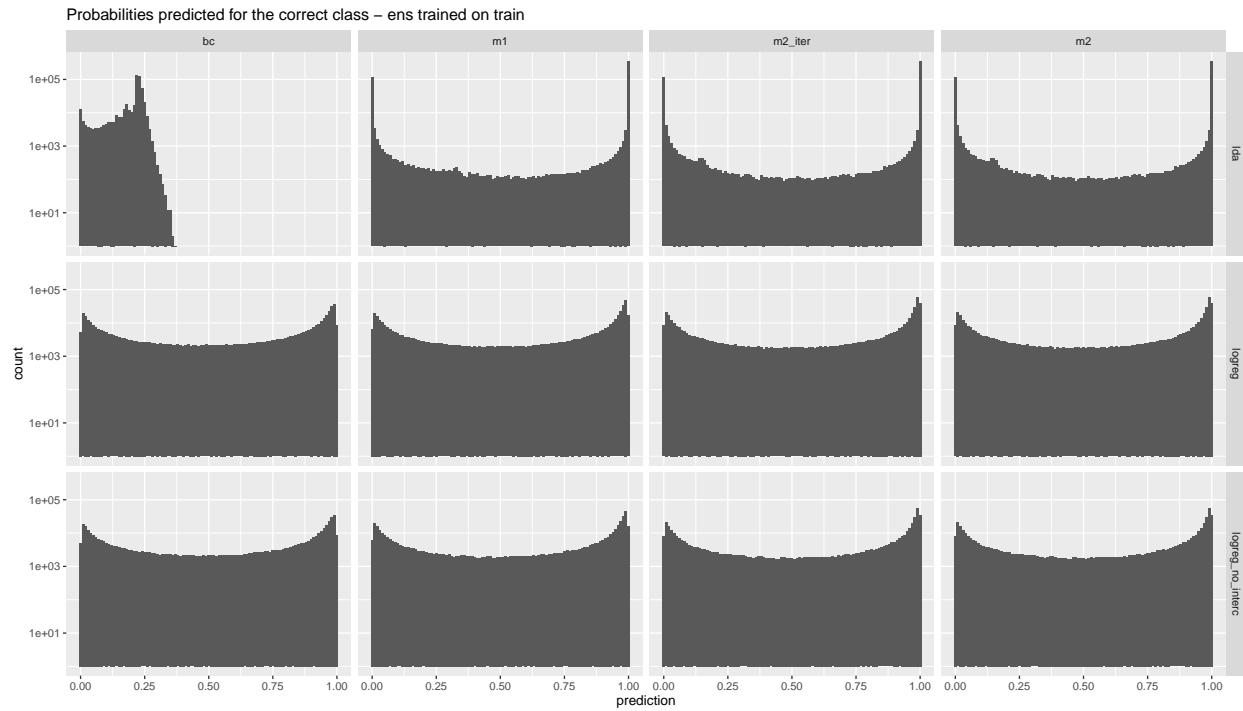
No zero probabilities produced

```
train_ens_cor_probs <- melt(ens_outputs_gathered$train_training)
train_ens_cor_probs <- train_ens_cor_probs[, c(-5, -6)]
names(train_ens_cor_probs) <- c("replication", "combining_method", "coupling_method", "fold", "prediction")
train_ens_cor_probs[, c("combining_method", "coupling_method")] <- lapply(train_ens_cor_probs[, c("combining_method", "coupling_method")], as.factor)
levels(train_ens_cor_probs$combining_method) <- ens_outputs_gathered$combining_methods
levels(train_ens_cor_probs$coupling_method) <- ens_outputs_gathered$coupling_methods
```

```
train_ens_cor_preds_histo <- ggplot(data=train_ens_cor_probs) + geom_histogram(mapping=aes(x=prediction))
train_ens_cor_preds_histo
```

```
## Warning: Transformation introduced infinite values in continuous y-axis
```

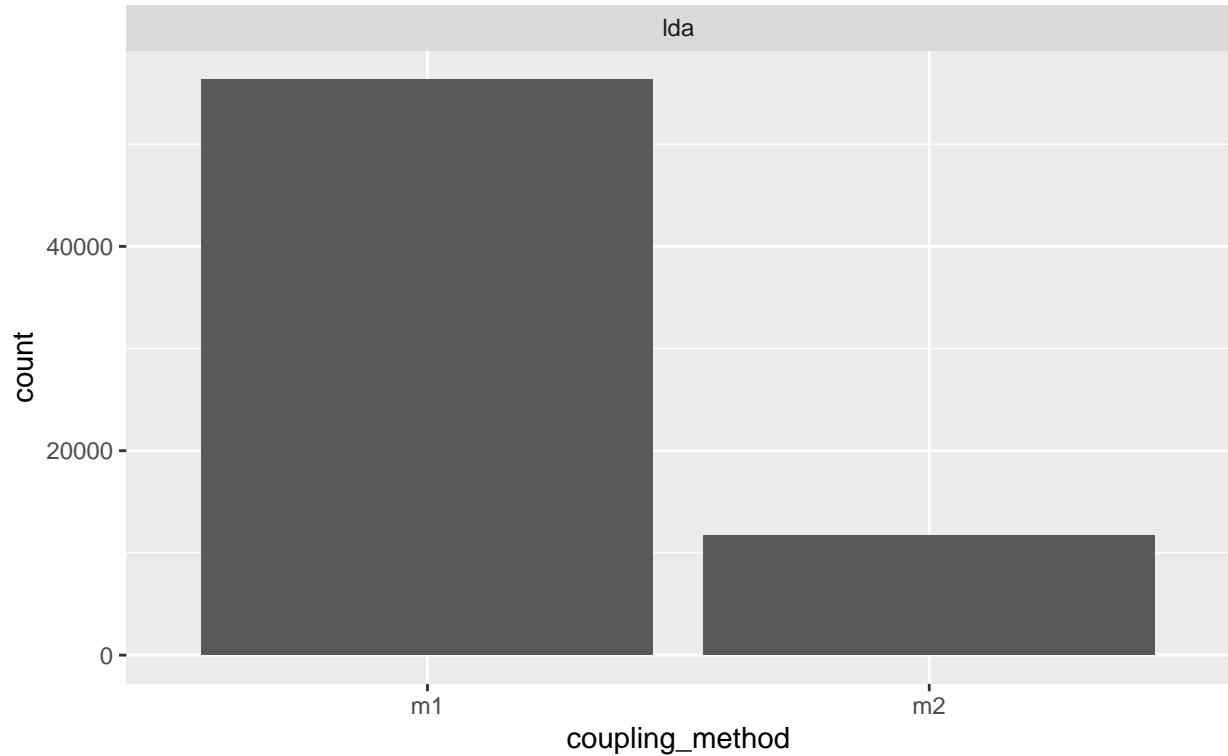
```
## Warning: Removed 63 rows containing missing values (geom_bar).
```



Strange behavior observed for bc coupling method in case of combining method lda is not present for combining methods logreg.

```
train_ens_zero_counts <- ggplot(data=train_ens_cor_probs[train_ens_cor_probs$prediction <= 0, ]) + geom_
## Warning: Ignoring unknown parameters: binwidth, bins, pad
train_ens_zero_counts
```

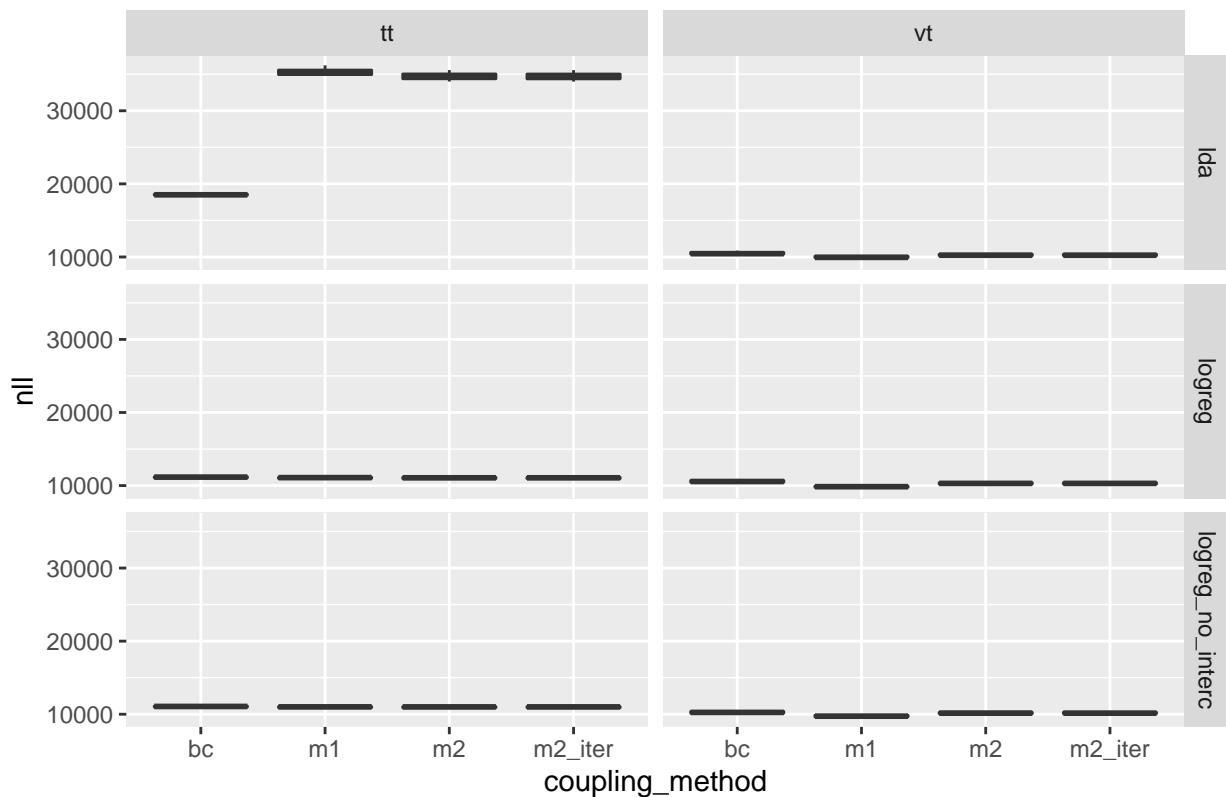
Counts of zero or lower probabilities predicted for the correct class by combiner train methodology



m2\_iter and bc didn't produce any zero probability outputs. Neither were there any zero probability outputs for combining method logistic regression.

```
val_ens_nll <- ggplot(data=ens_results) + geom_boxplot(mapping=aes(x=coupling_method, y=nll)) + facet_grid(~combiner) + ggtitle("Comparison of nll for coupling methods for different combiner train methodologies")  
val_ens_nll
```

## Comparison of nll for coupling methods for different combiner train metho

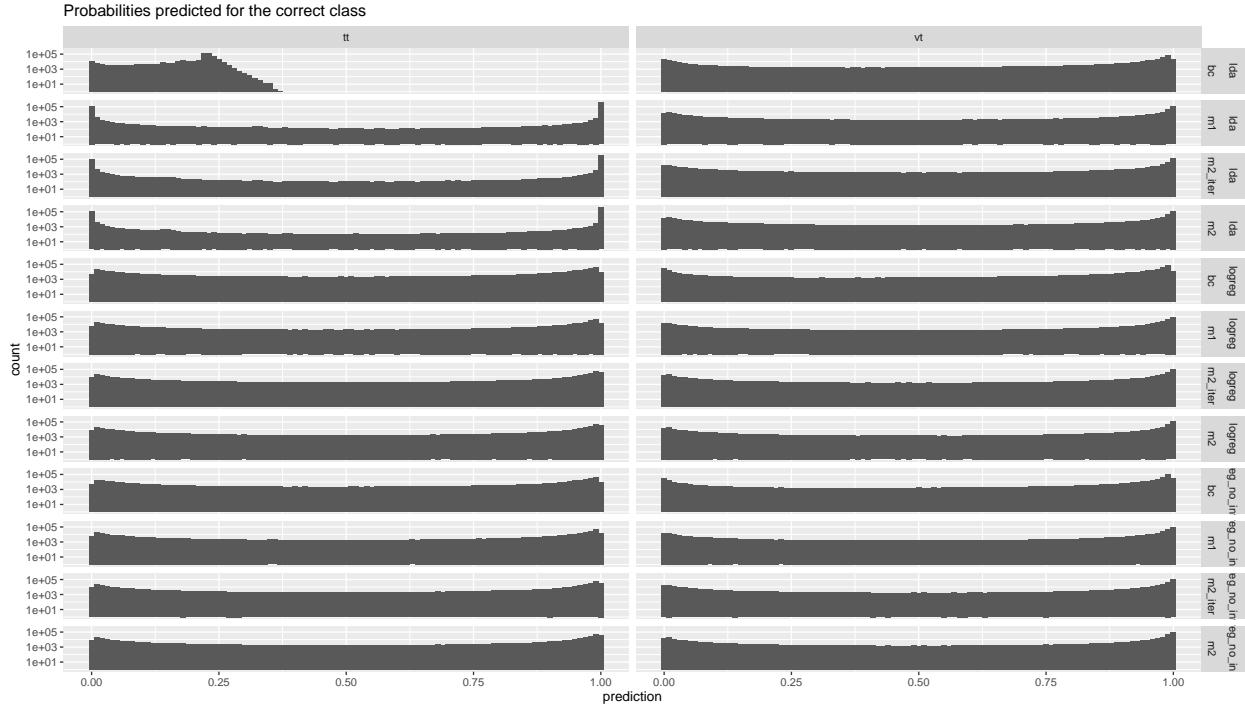


```
val_ens_cor_probs$train_type <- "vt"
train_ens_cor_probs$train_type <- "tt"
ens_cor_probs <- rbind(val_ens_cor_probs, train_ens_cor_probs)
```

```
ens_cor_preds_histo <- ggplot(data=ens_cor_probs) + geom_histogram(mapping=aes(x=prediction), binwidth=
```

```
## Warning: Transformation introduced infinite values in continuous y-axis
```

```
## Warning: Removed 63 rows containing missing values (geom_bar).
```



In case of logreg combining method there are fewer values in the extreme bins than for lda. Also, there are fewer of these extreme values for tt than for vt.

```

df_val_Rs <- melt(np$load(file.path(base_dir, "val_training_class_aggr_R.npy")))
df_train_Rs <- melt(np$load(file.path(base_dir, "train_training_class_aggr_R.npy")))
co_m_R <- read.csv(file.path(base_dir, "R_mat_co_m_names.csv"), header=FALSE)

names(df_val_Rs) <- c("combining_method", "precision", "class", "class1", "class2", "prob")
names(df_train_Rs) <- c("combining_method", "precision", "class", "class1", "class2", "prob")

df_val_Rs[,c("class", "class1", "class2", "combining_method")] <- lapply(df_val_Rs[,c("class", "class1",
df_train_Rs[,c("class", "class1", "class2", "combining_method")] <- lapply(df_train_Rs[,c("class", "class1",
levels(df_val_Rs$combining_method) <- co_m_R$V1
levels(df_train_Rs$combining_method) <- co_m_R$V1

df_val_Rs$train_type <- "vt"
df_train_Rs$train_type <- "tt"
class_mean_Rs <- rbind(df_val_Rs, df_train_Rs)

df_aggr_Rs_diff <- class_mean_Rs %>% pivot_wider(names_from = train_type, values_from = prob) %>% mutate

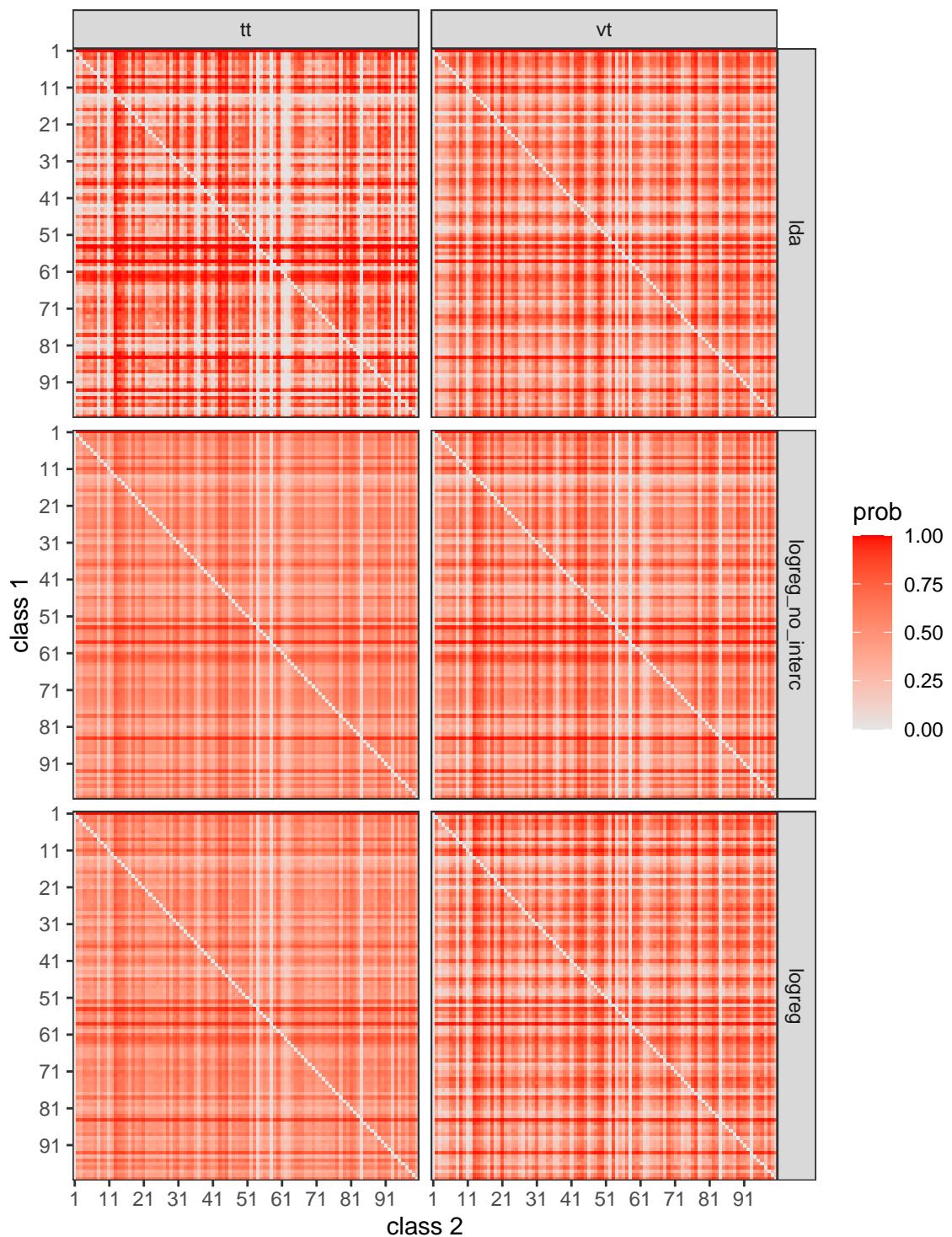
for (cls in 1:classes)
{
  cur_class_Rs <- class_mean_Rs %>% filter(class == cls)
  plot_cls <- ggplot(cur_class_Rs, aes(x = class2, y = class1)) +
    geom_raster(aes(fill=prob)) +
    facet_grid(rows=vars(combining_method), cols=vars(train_type)) +
    scale_fill_gradient(low="grey90", high="red", limits=c(0, 1)) +
    scale_y_discrete(limits=rev, breaks=seq(1, 100, 10)) +
}

```

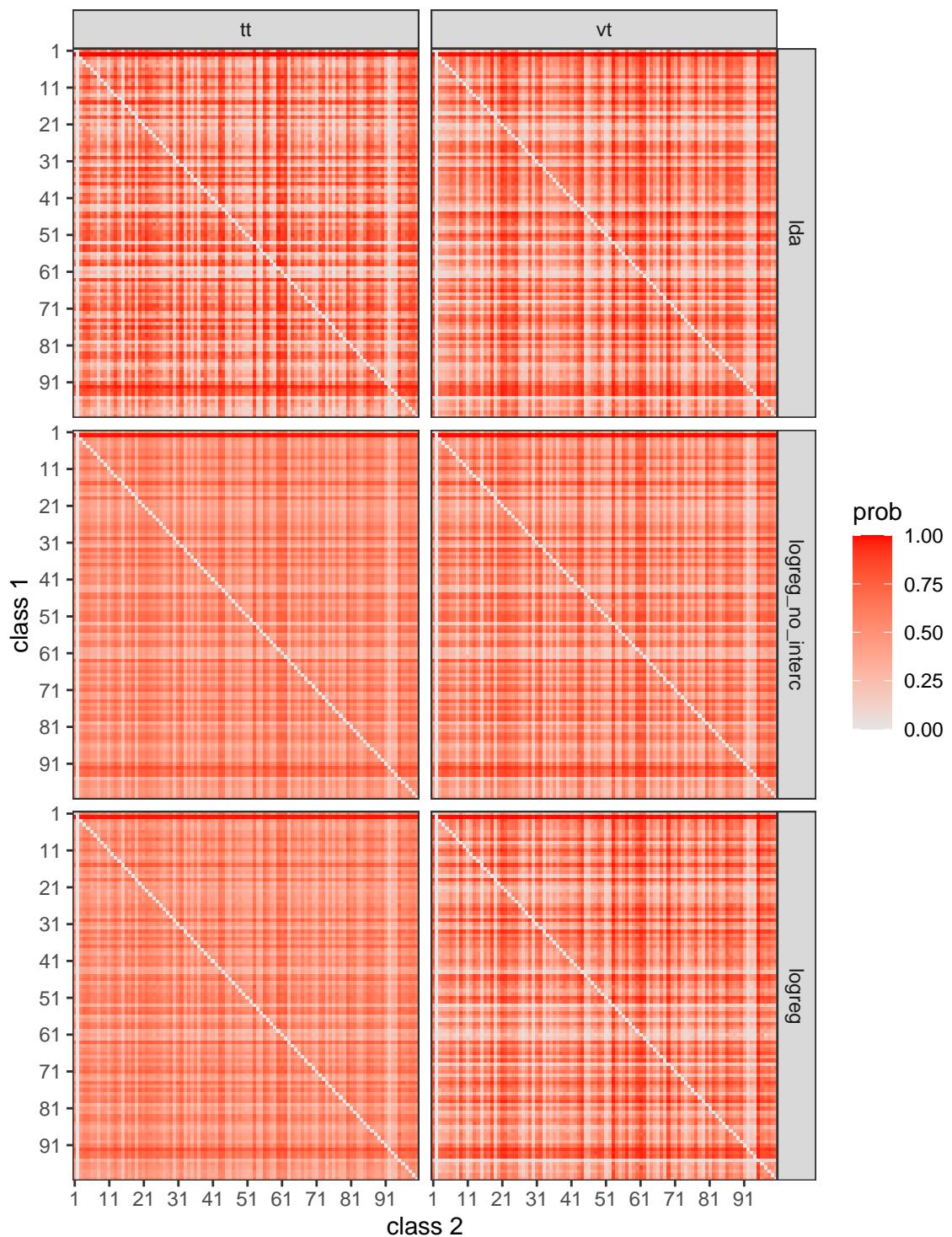
```
scale_x_discrete(breaks=seq(1, 100, 10)) +
  labs(x="class 2", y="class 1", title=paste("Average pairwise probabilities - class ", cls)) +
  theme_bw()

print(plot_cls)
}
```

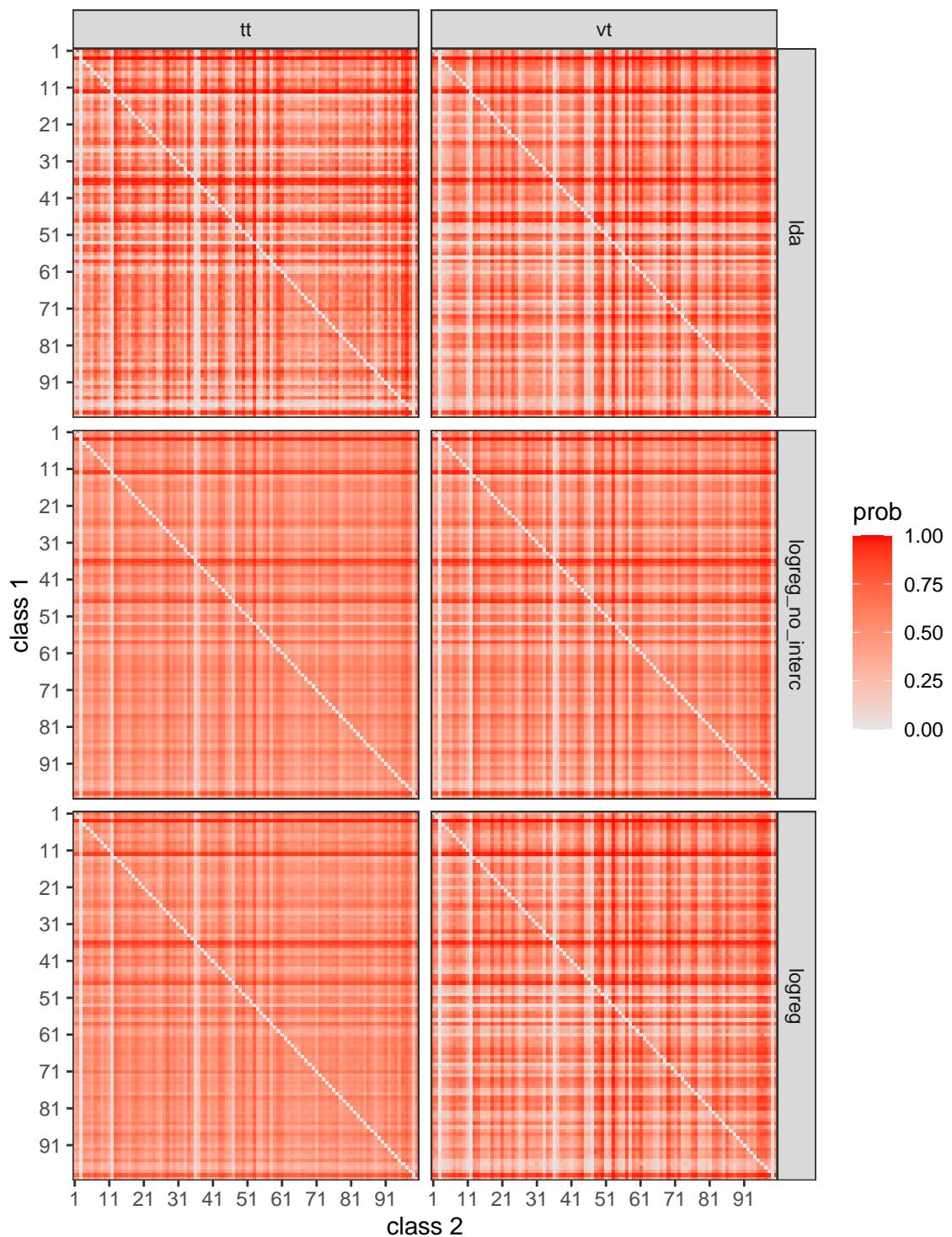
Average pairwise probabilities – class 1



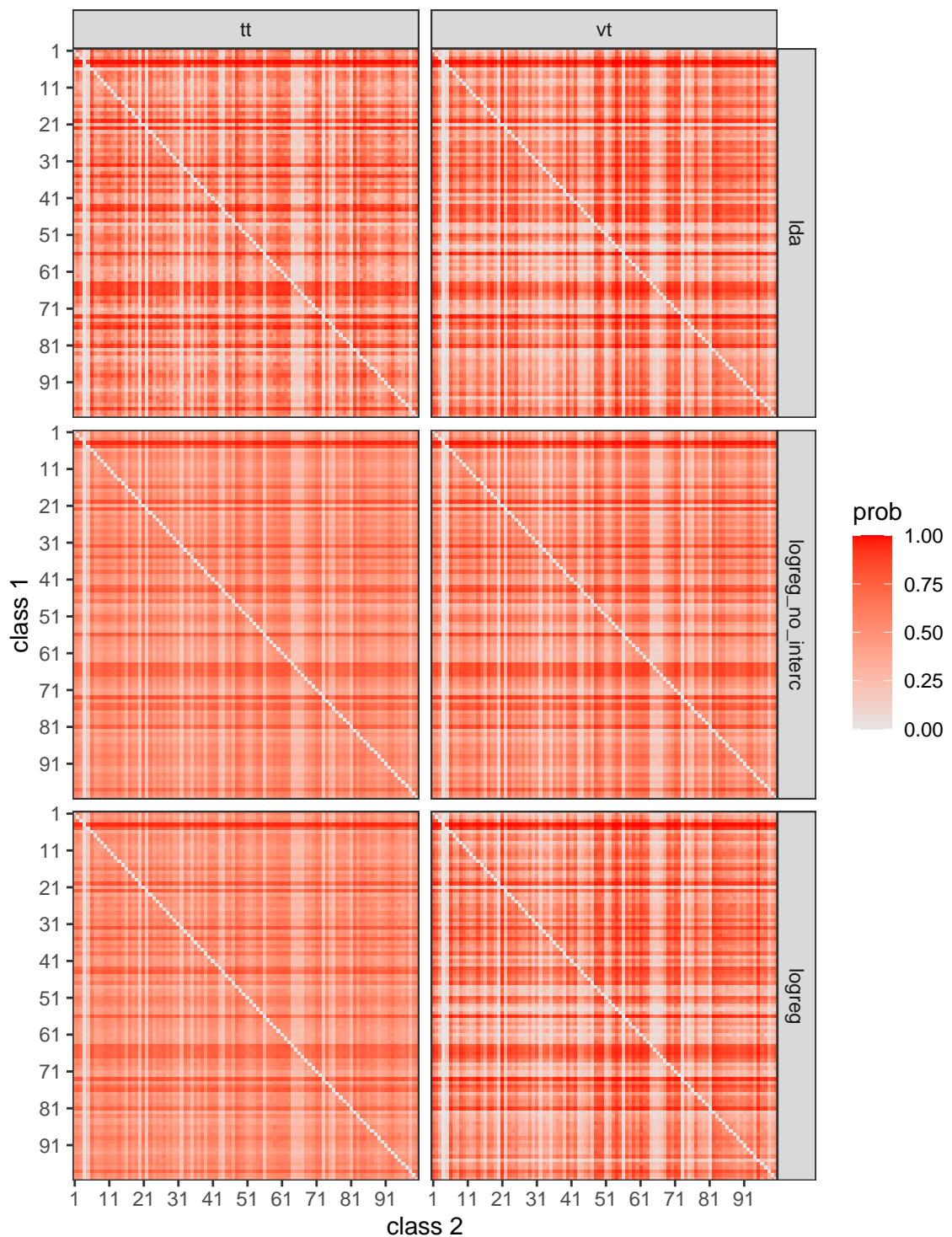
Average pairwise probabilities – class 2



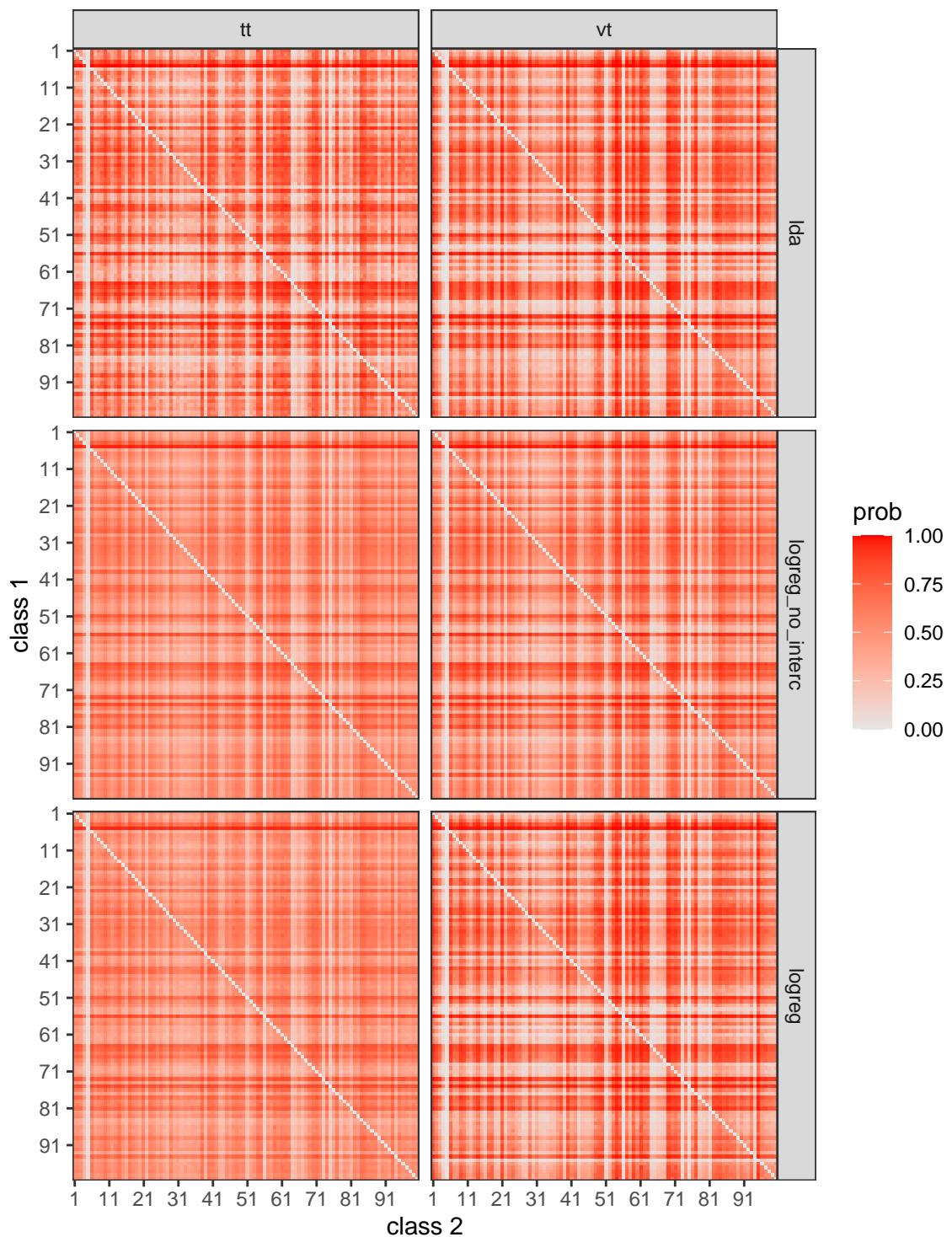
Average pairwise probabilities – class 3



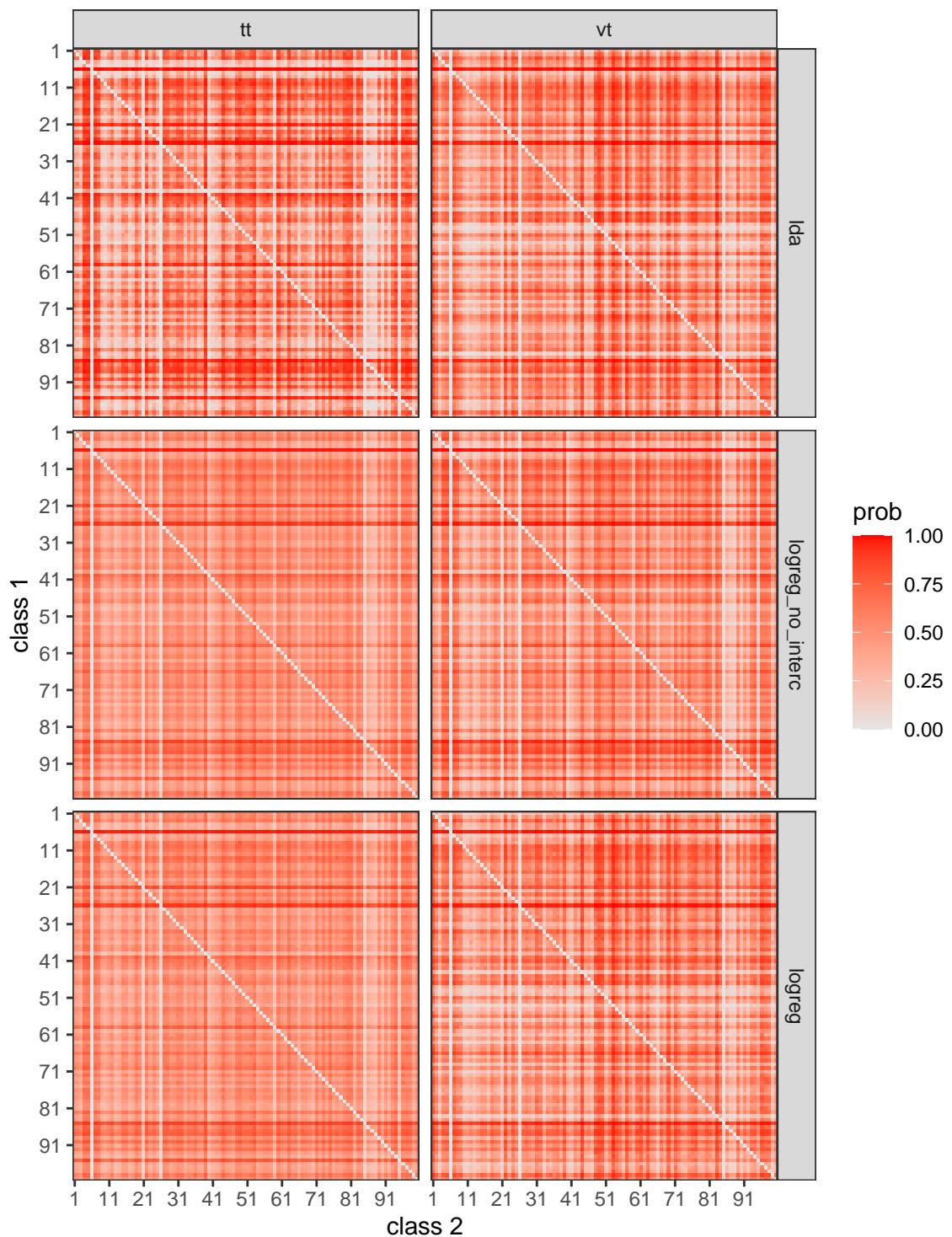
Average pairwise probabilities – class 4



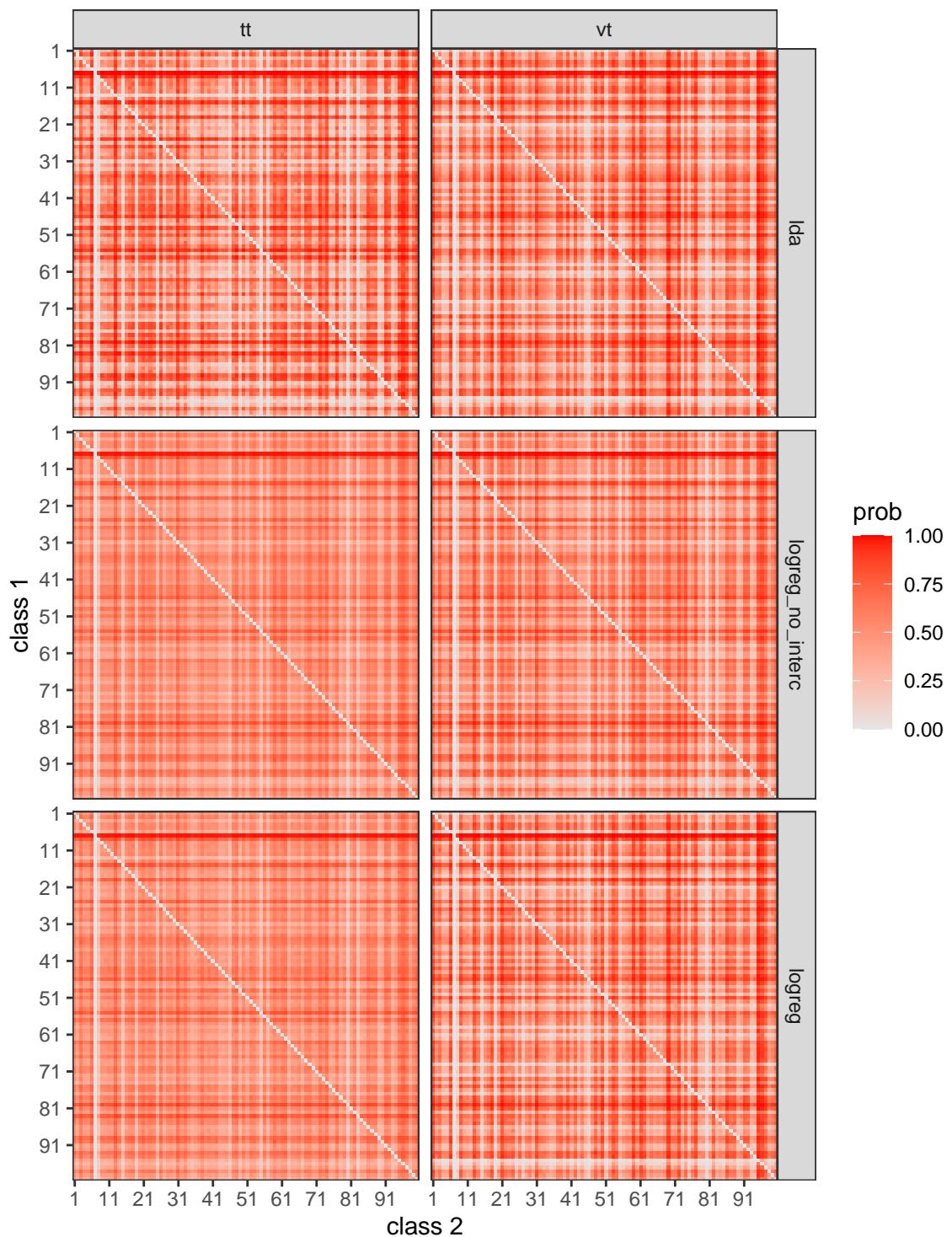
Average pairwise probabilities – class 5



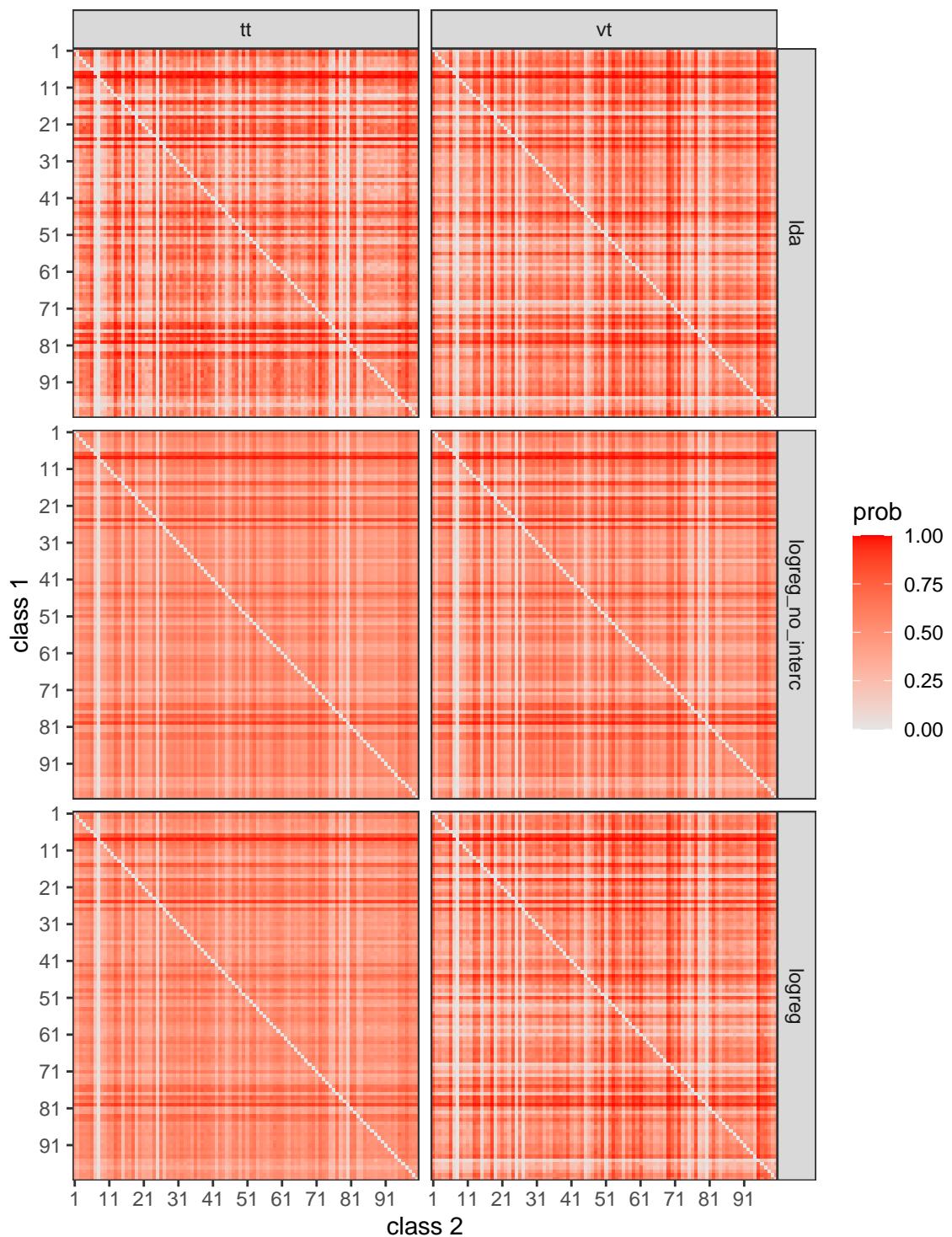
Average pairwise probabilities – class 6



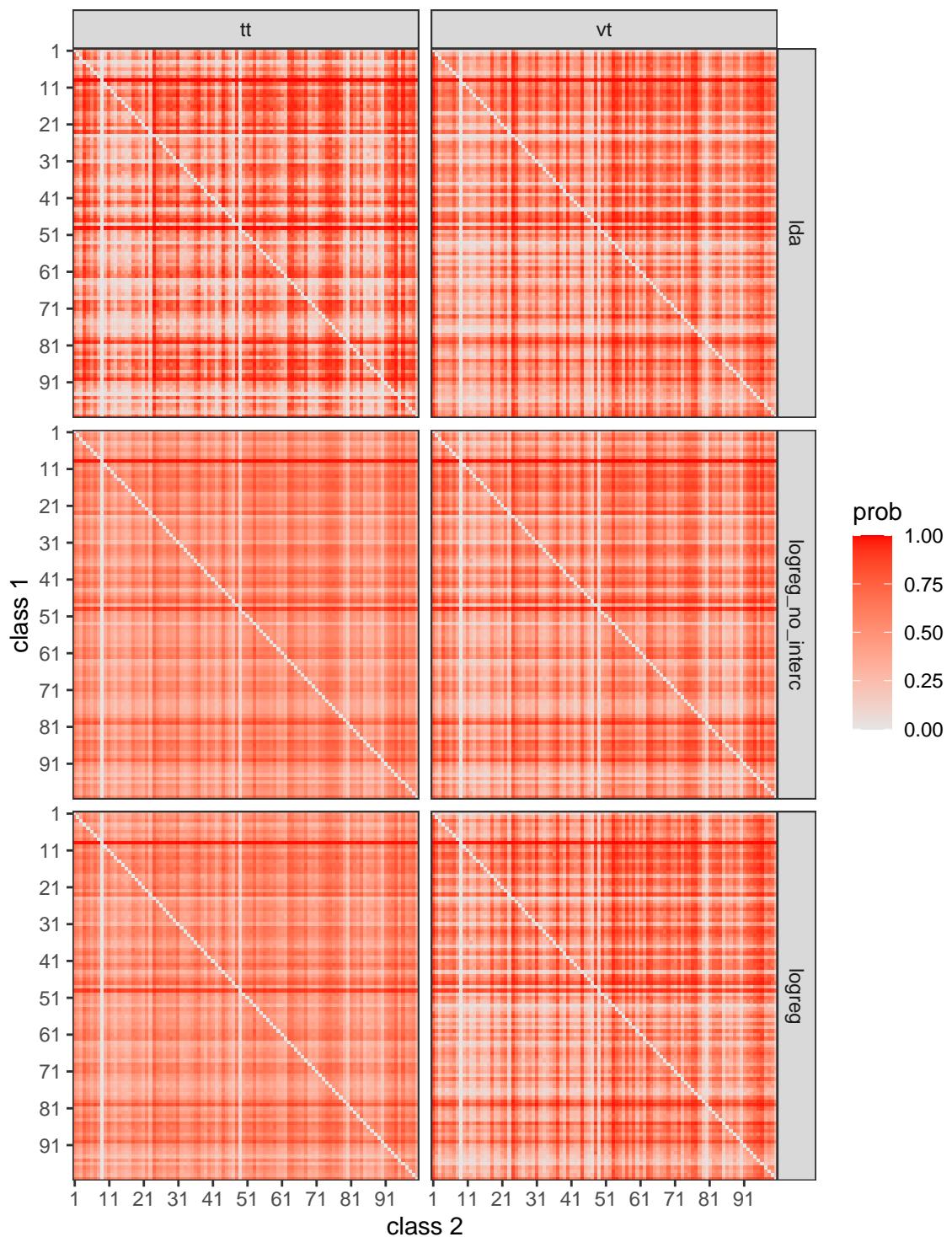
Average pairwise probabilities – class 7



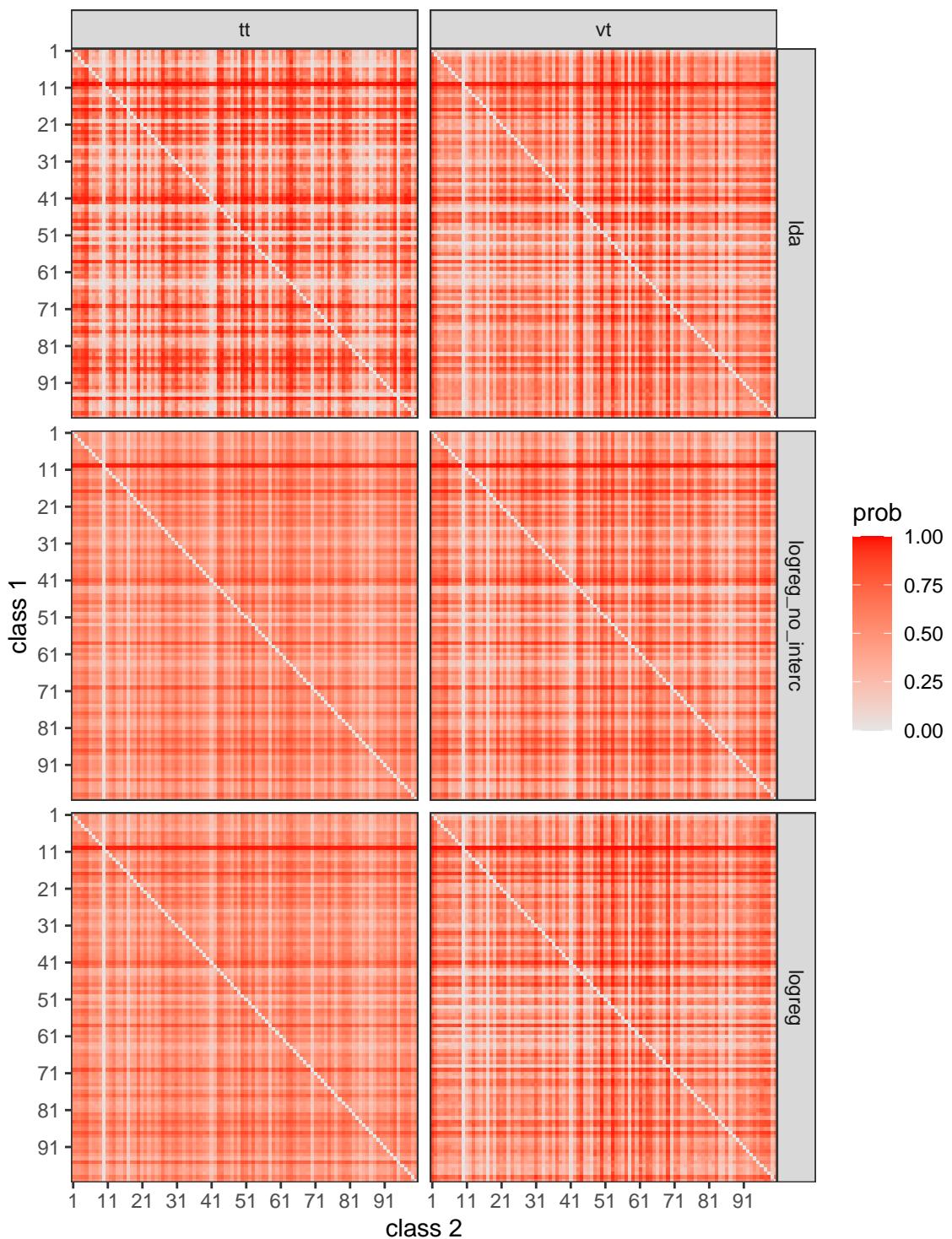
Average pairwise probabilities – class 8



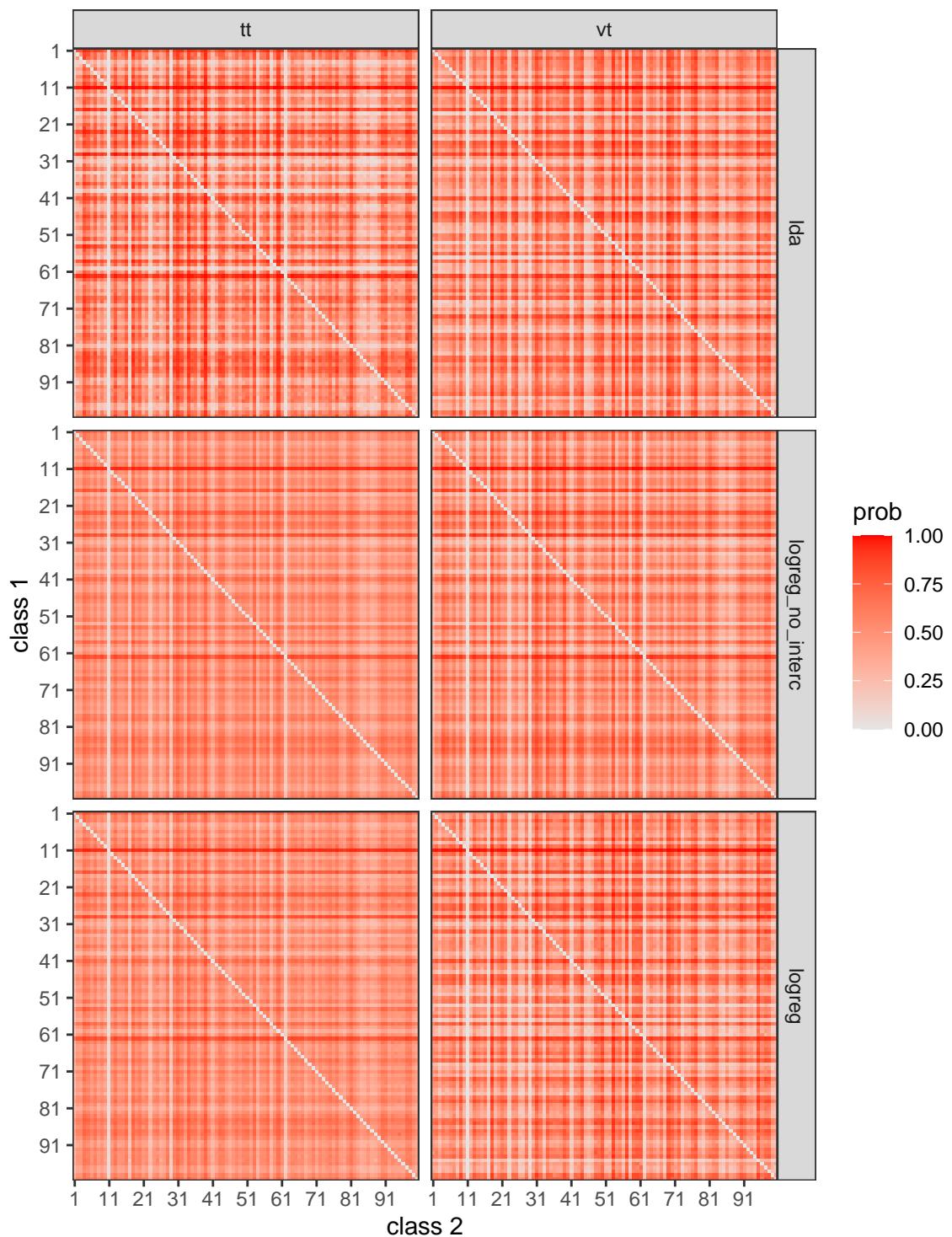
Average pairwise probabilities – class 9



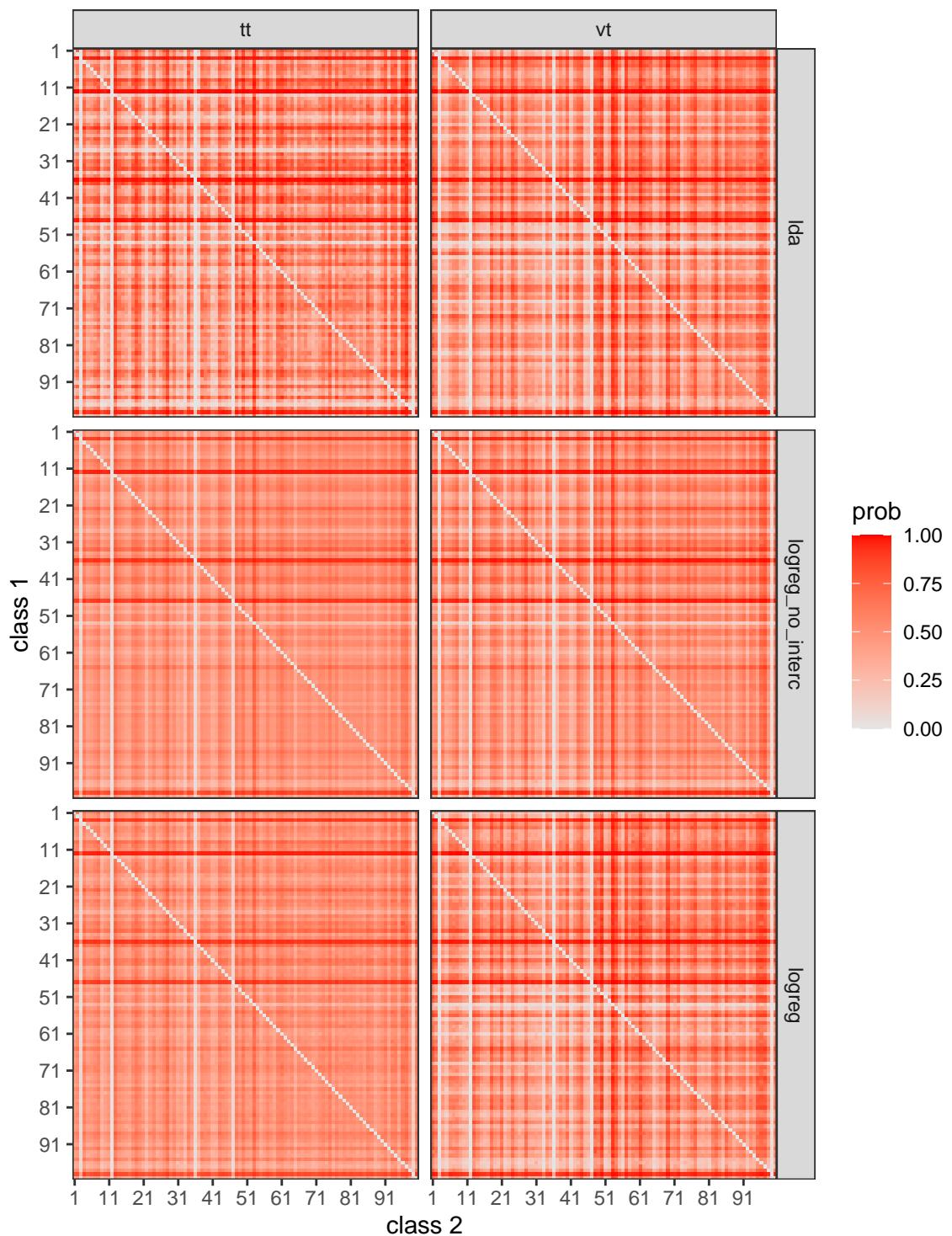
Average pairwise probabilities – class 10



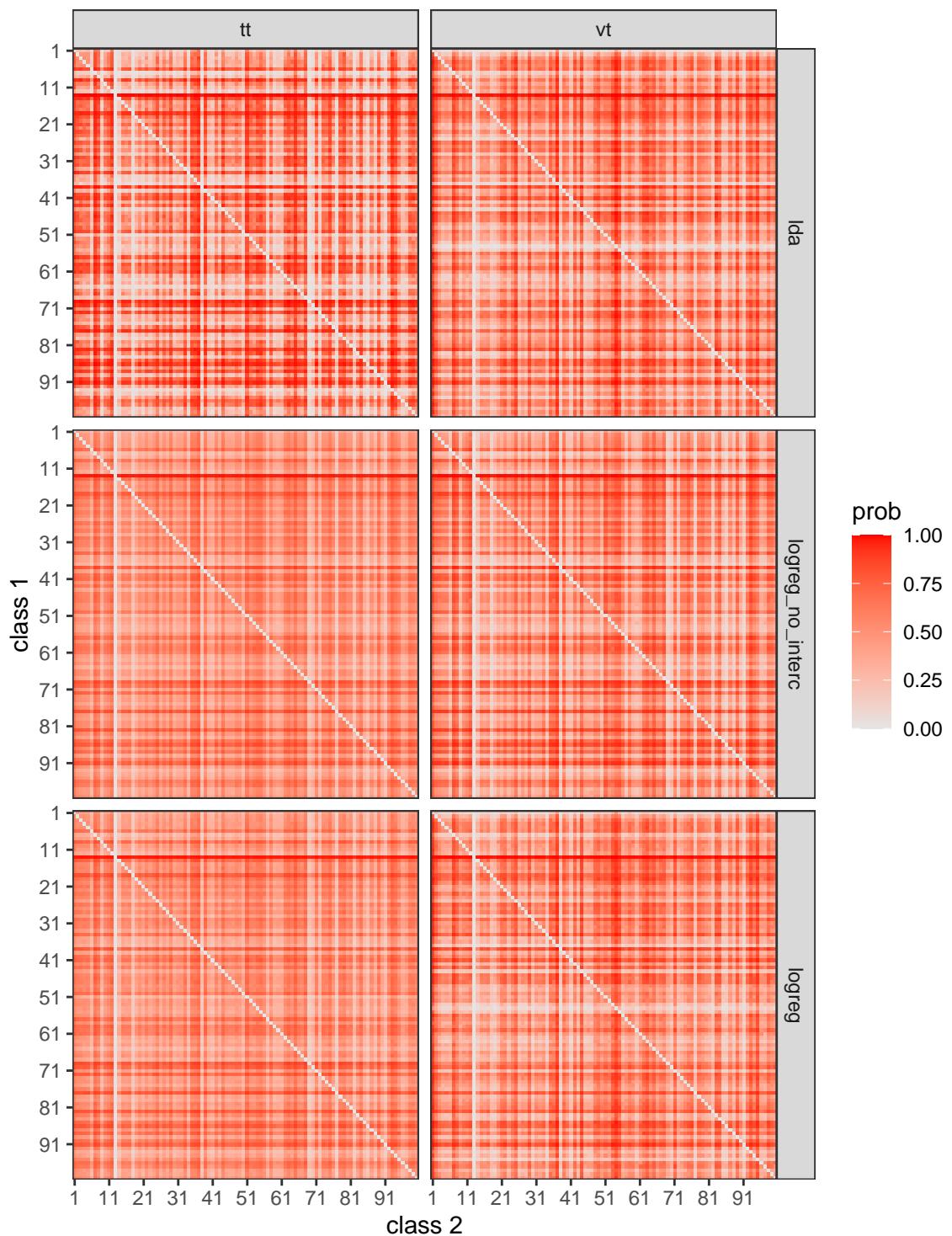
Average pairwise probabilities – class 11



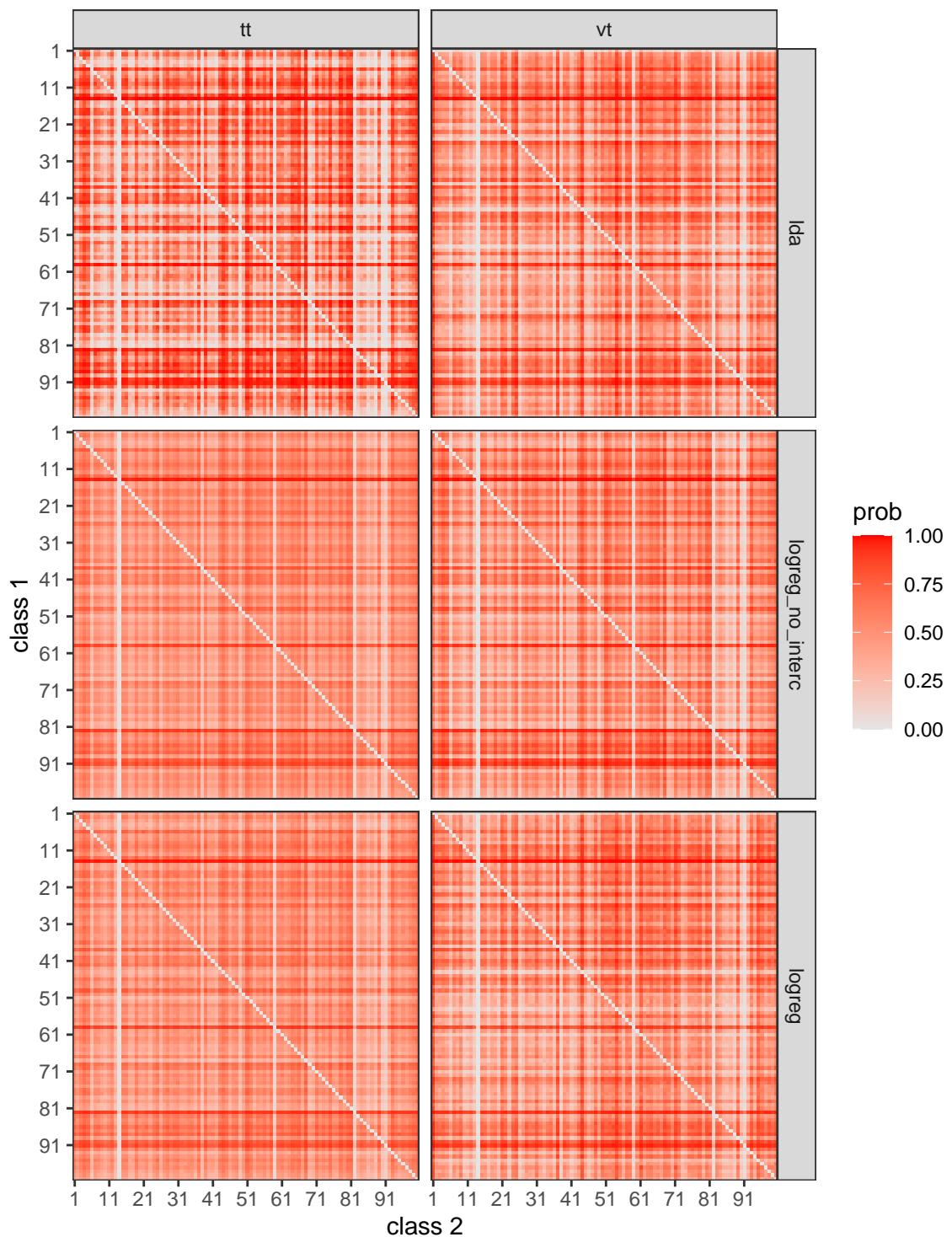
Average pairwise probabilities – class 12



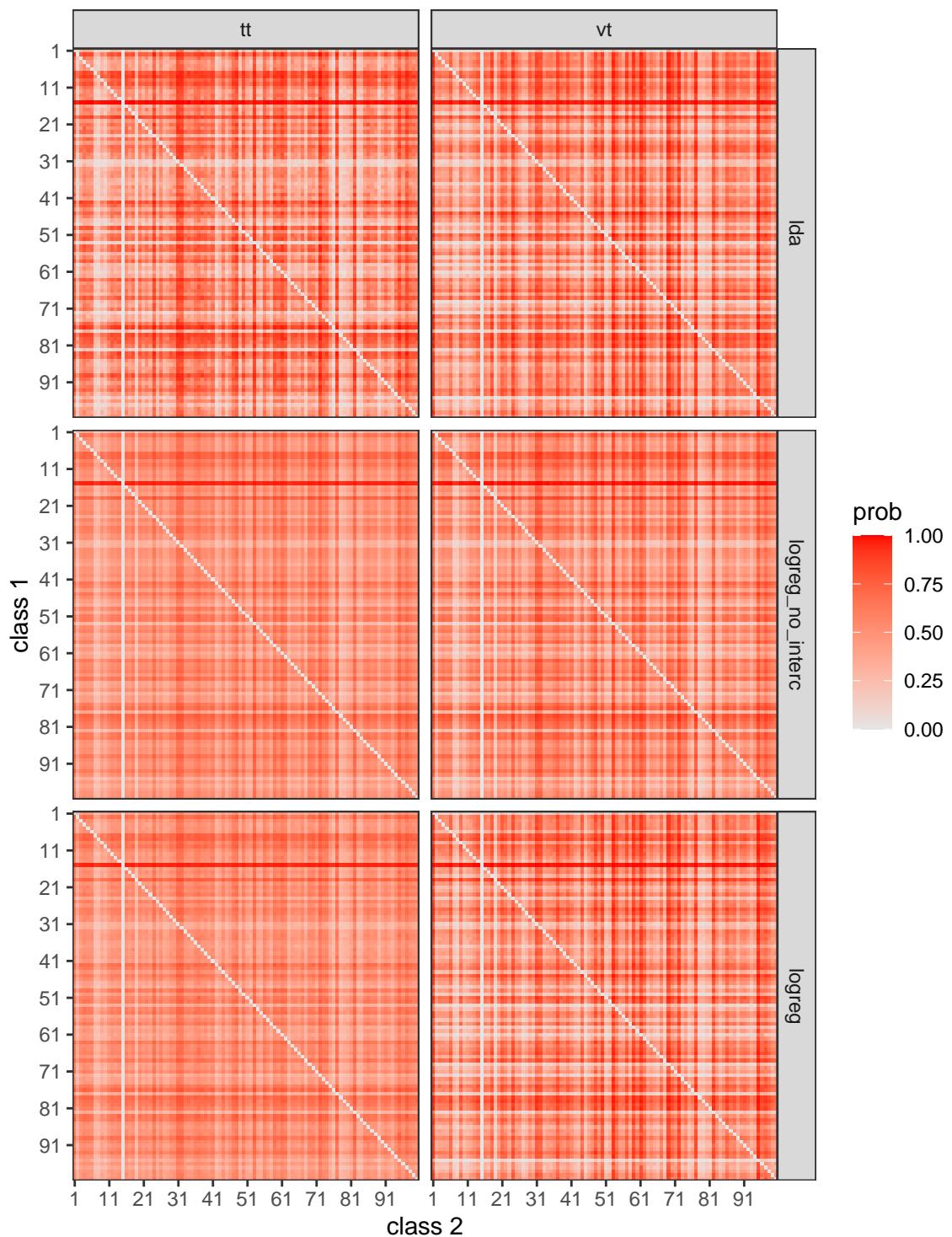
Average pairwise probabilities – class 13



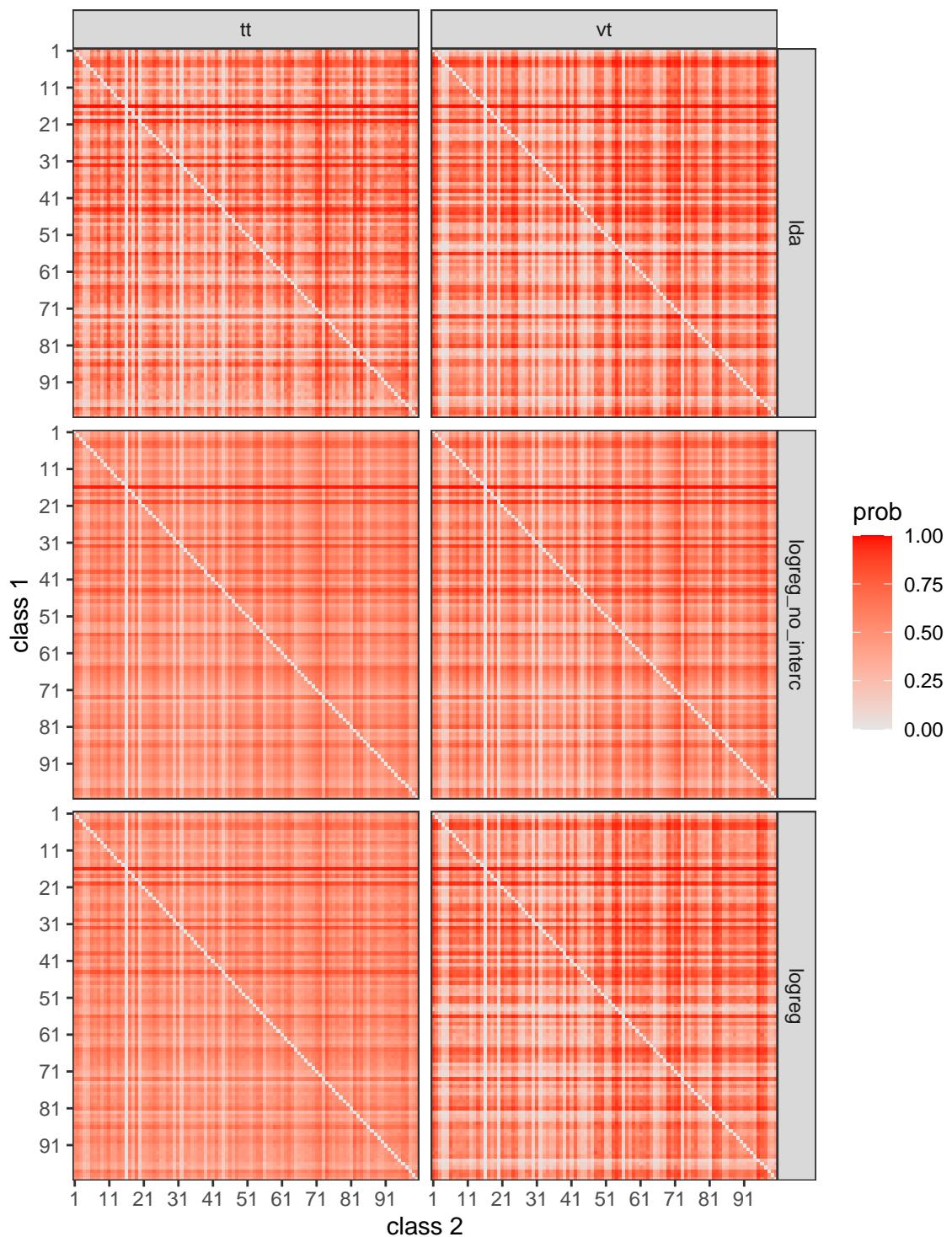
Average pairwise probabilities – class 14



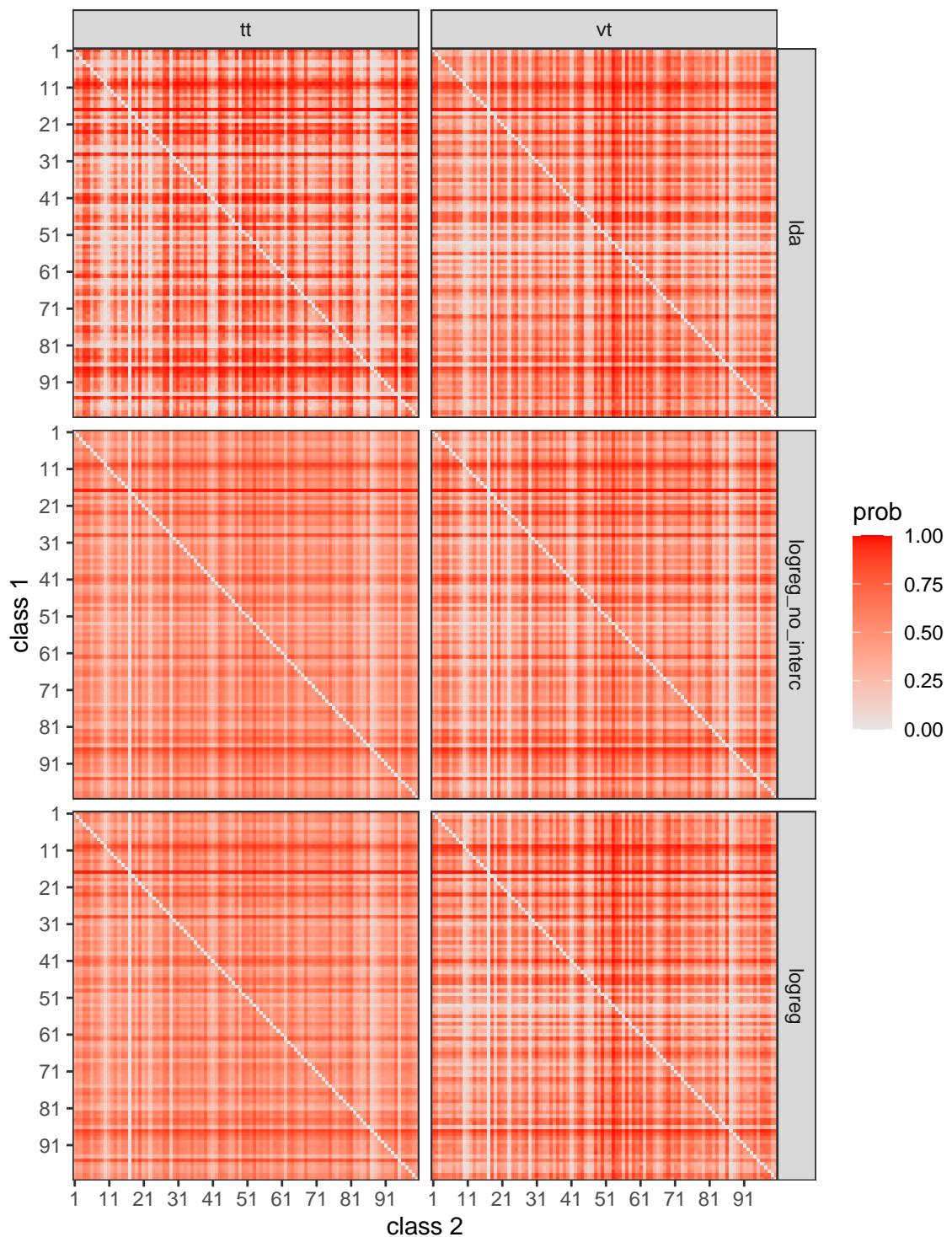
Average pairwise probabilities – class 15



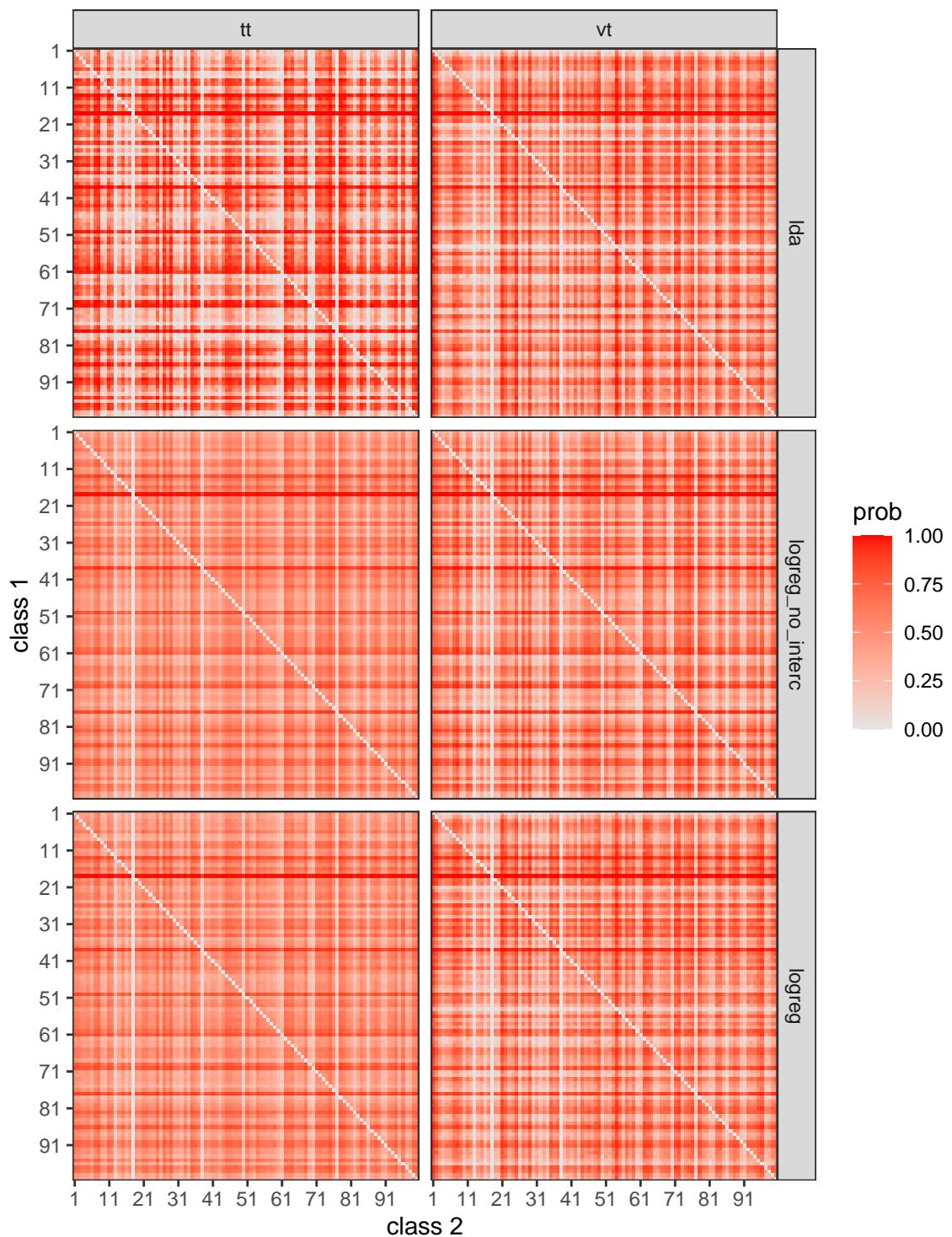
Average pairwise probabilities – class 16



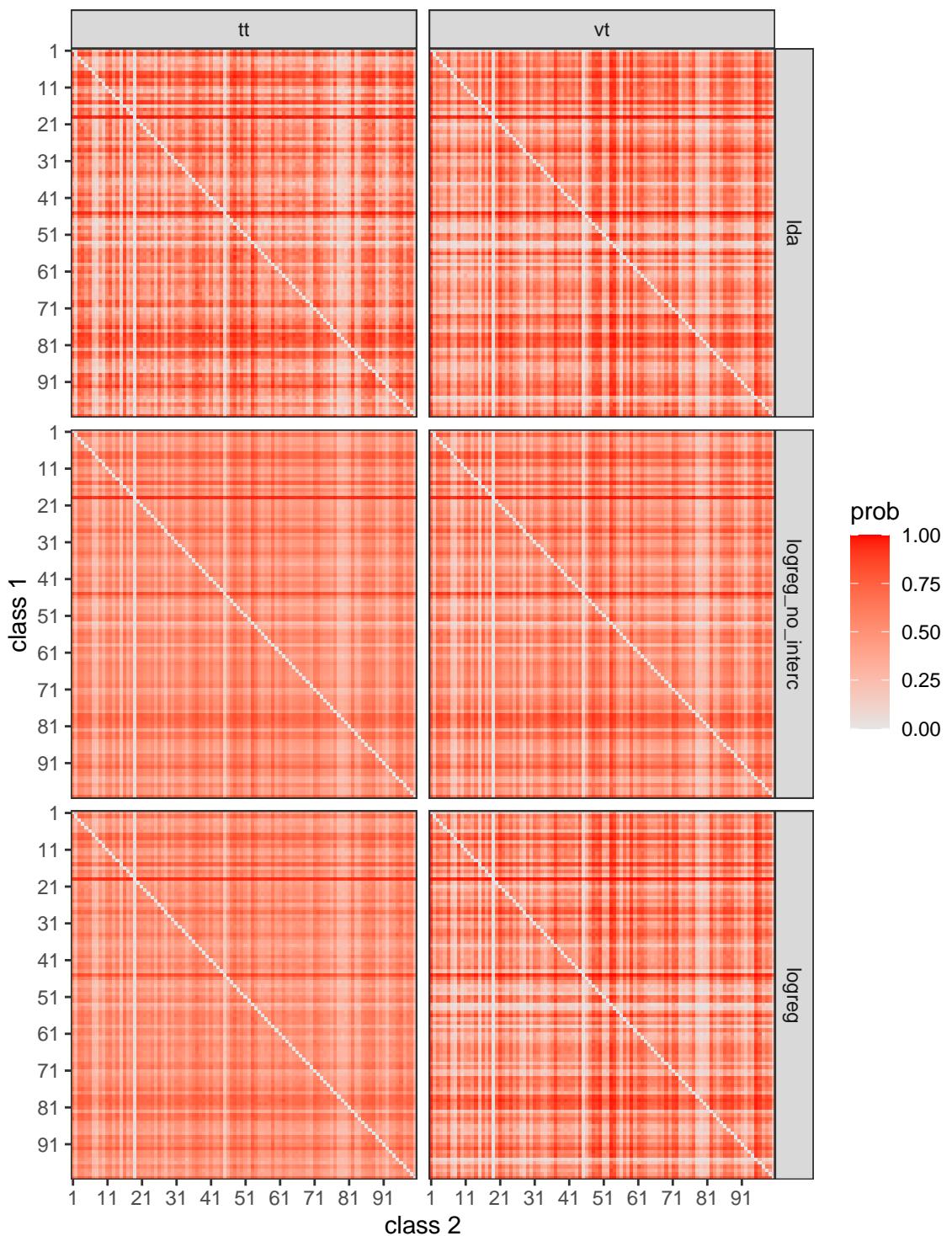
Average pairwise probabilities – class 17



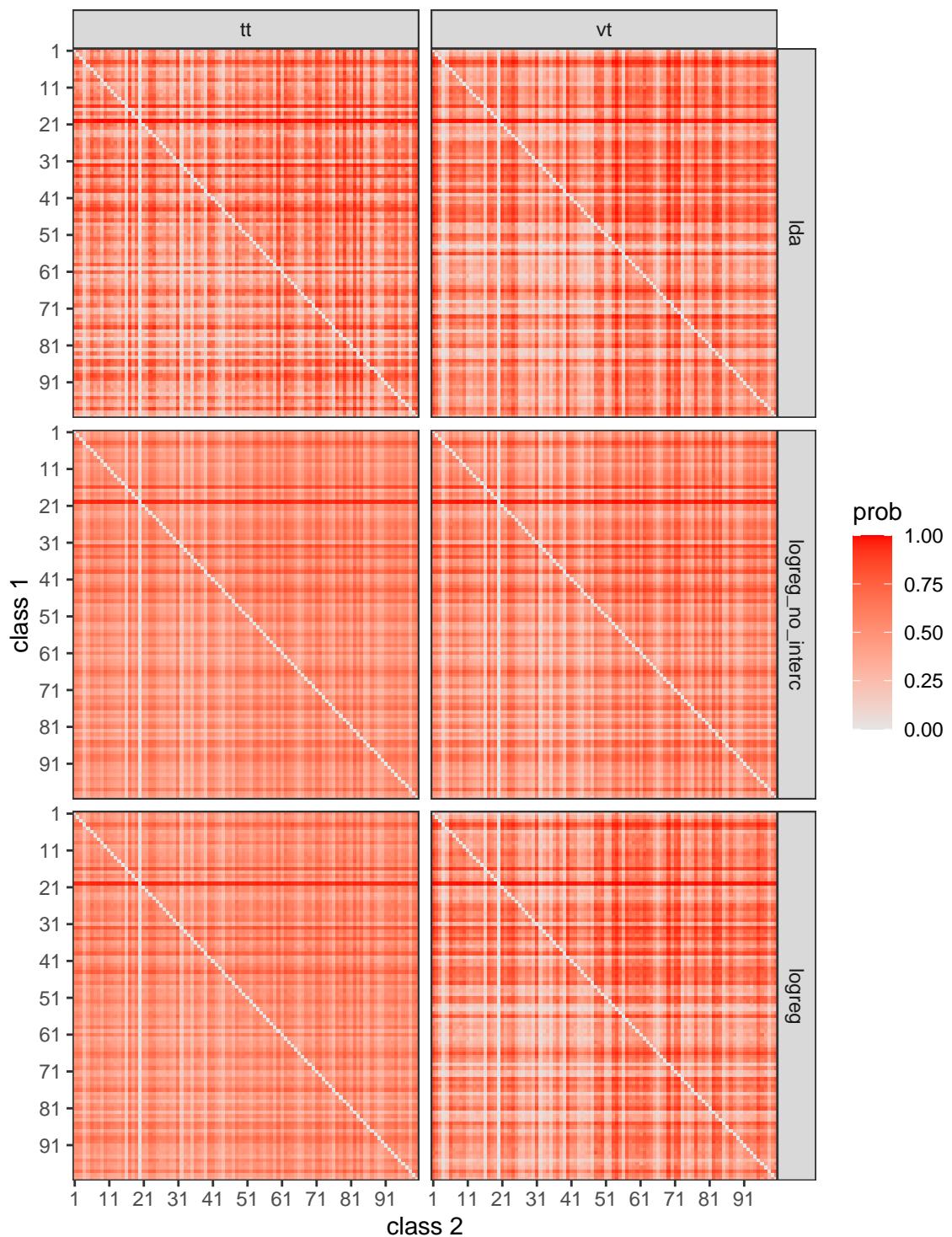
Average pairwise probabilities – class 18



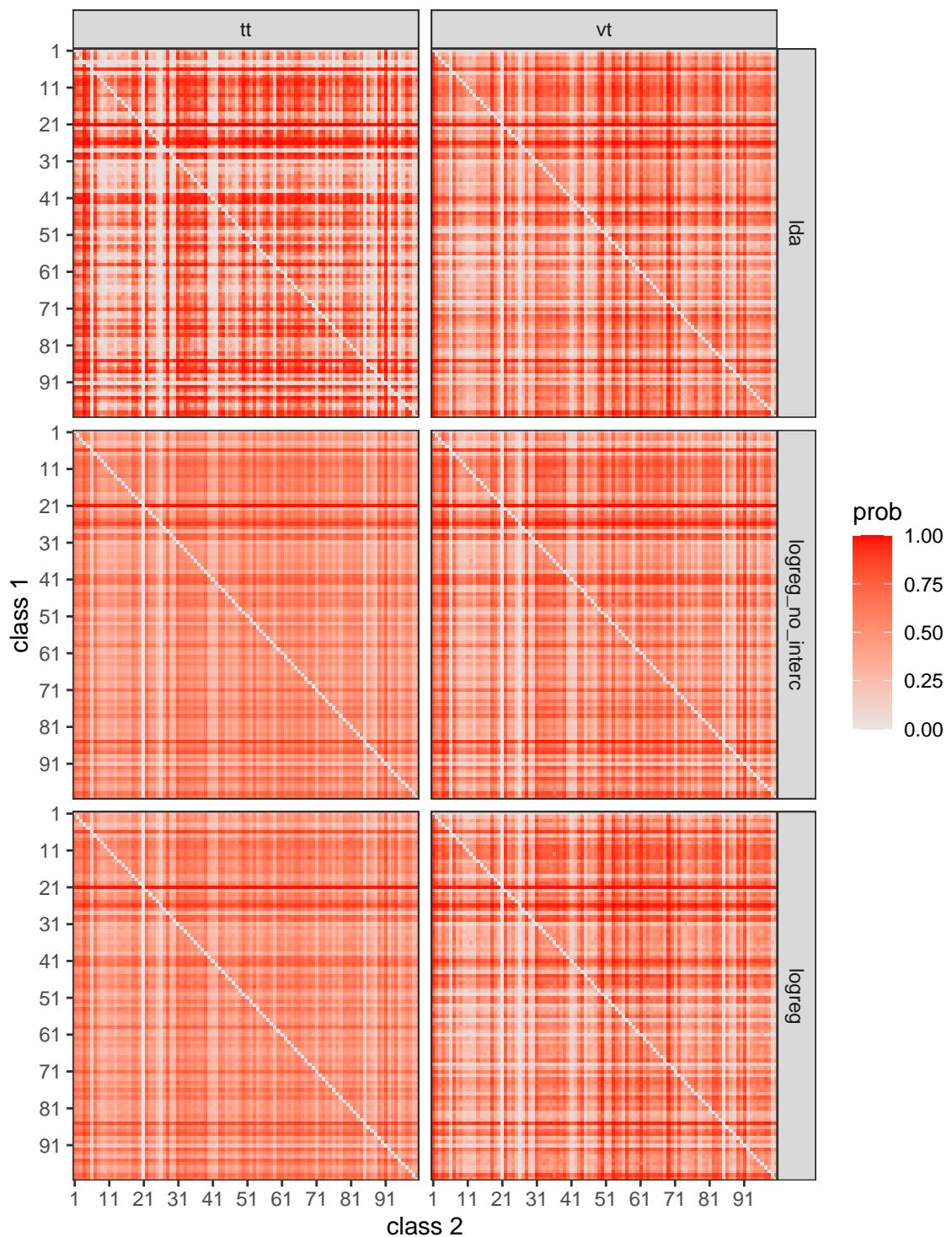
Average pairwise probabilities – class 19



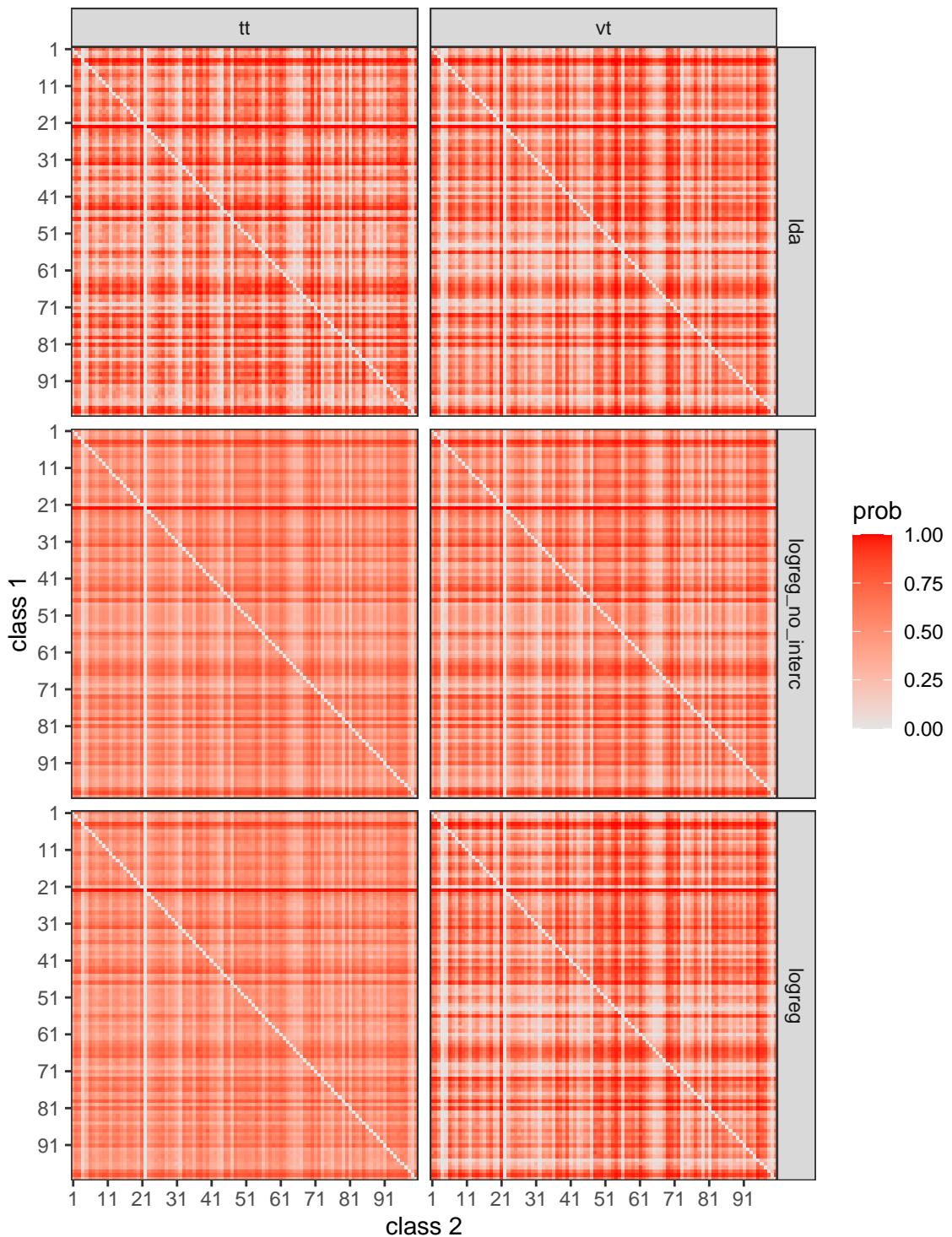
Average pairwise probabilities – class 20



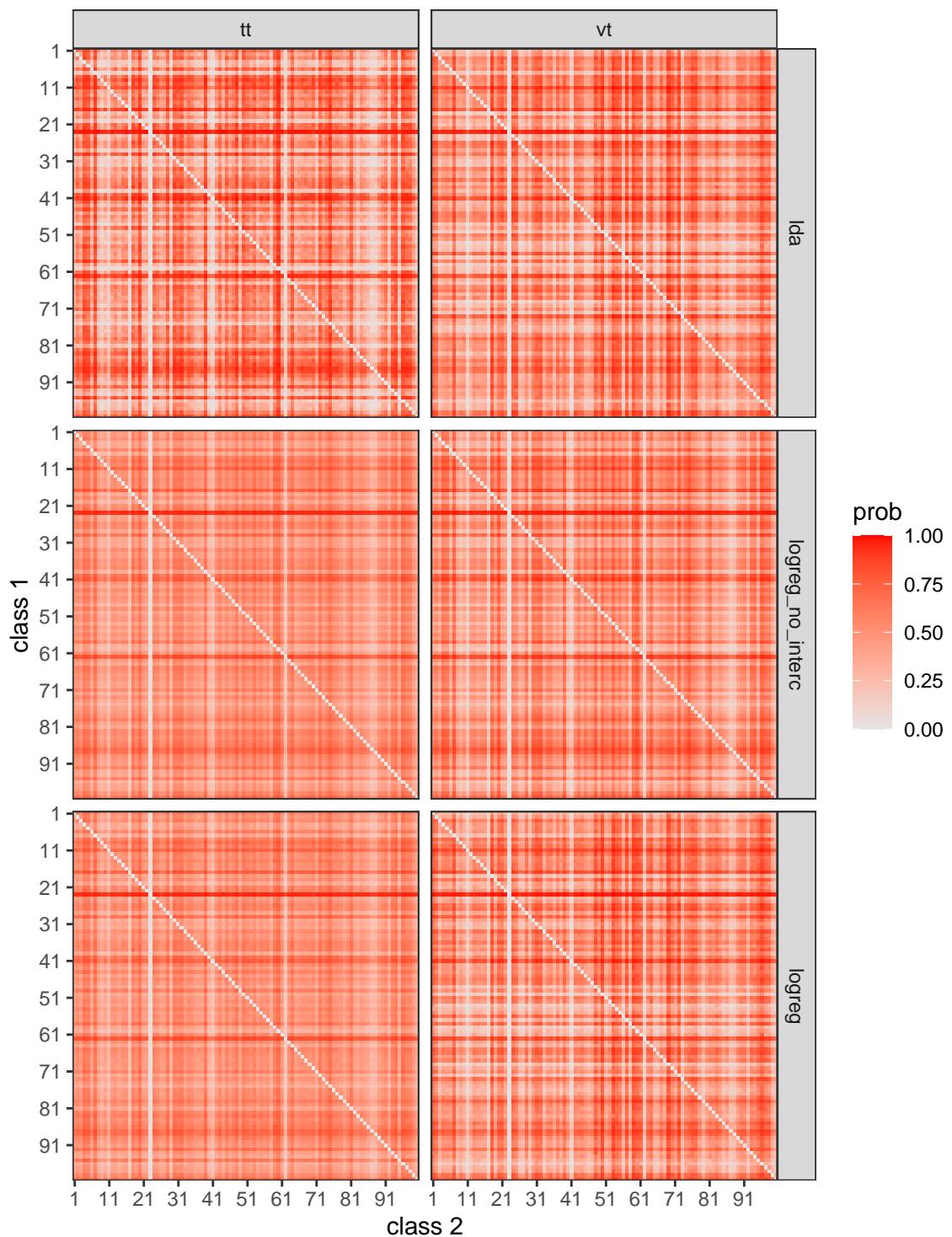
Average pairwise probabilities – class 21



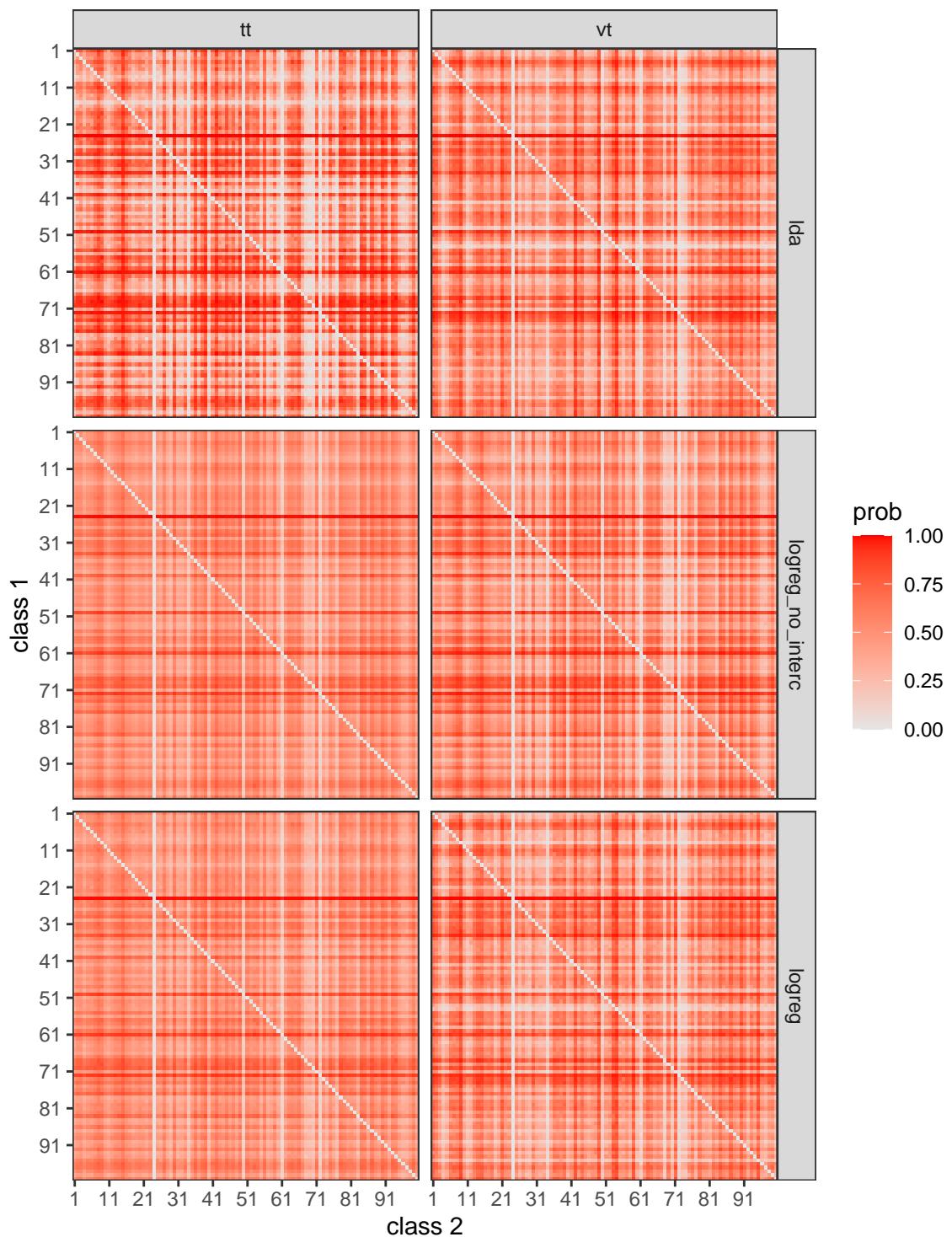
## Average pairwise probabilities – class 22



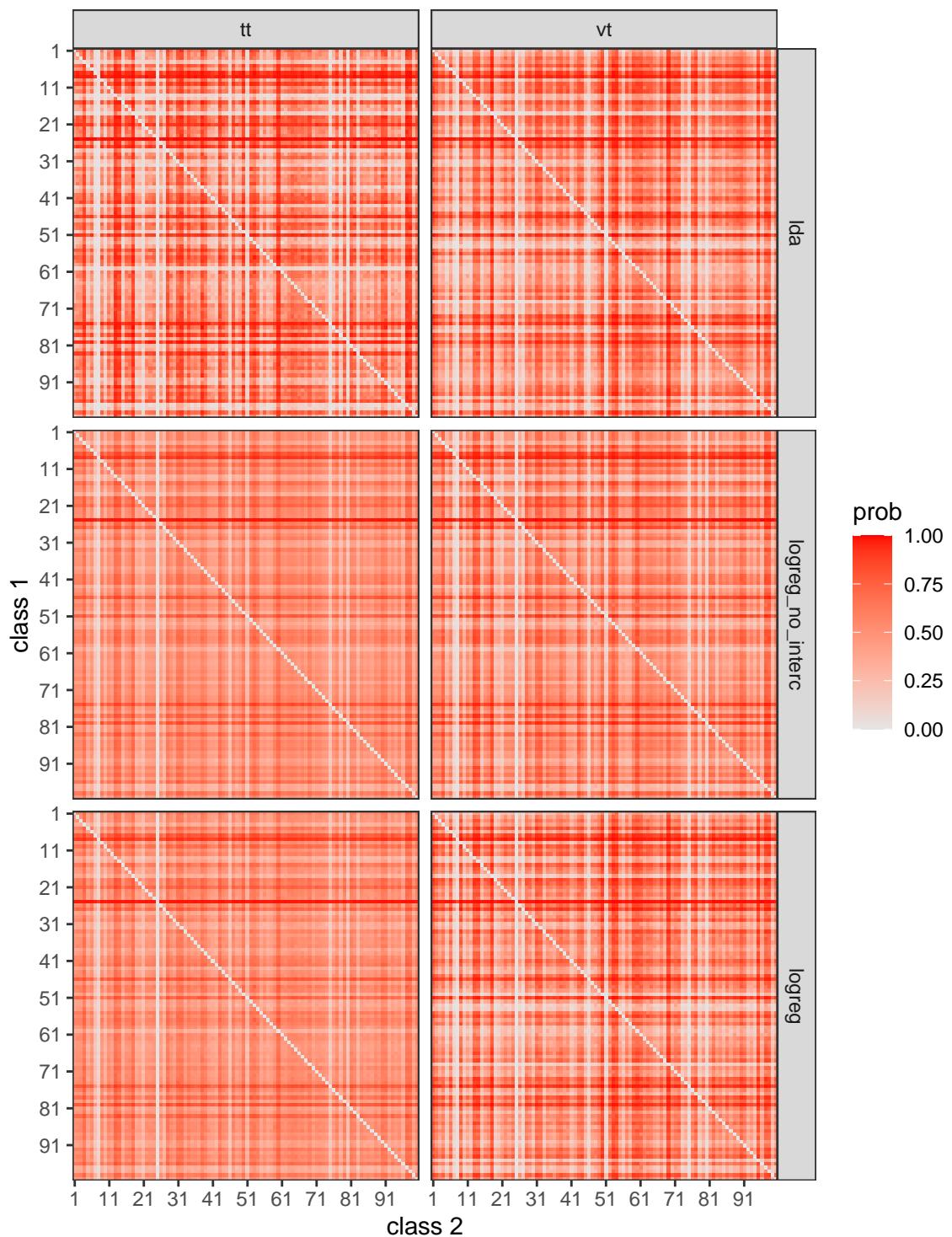
Average pairwise probabilities – class 23



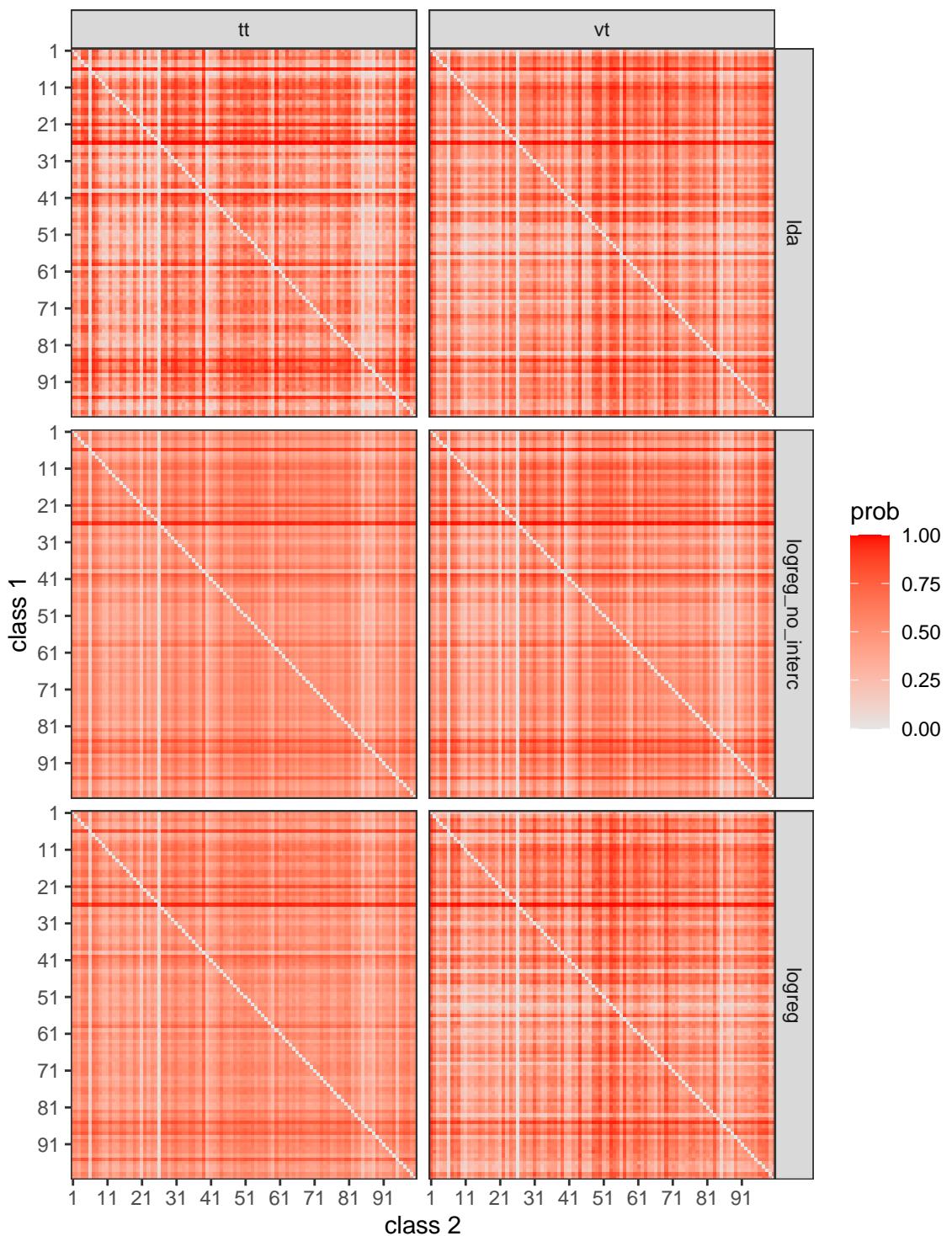
Average pairwise probabilities – class 24



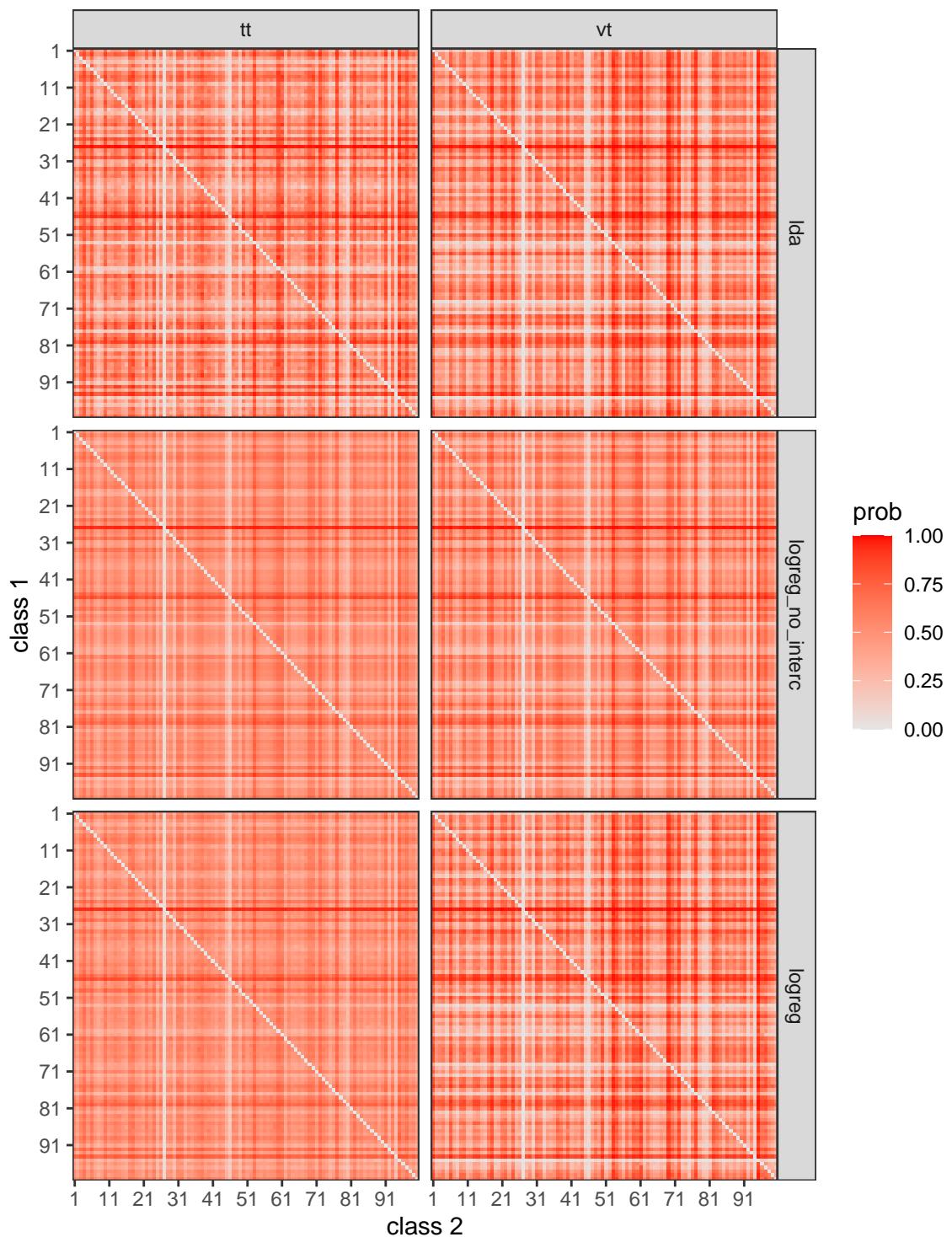
Average pairwise probabilities – class 25



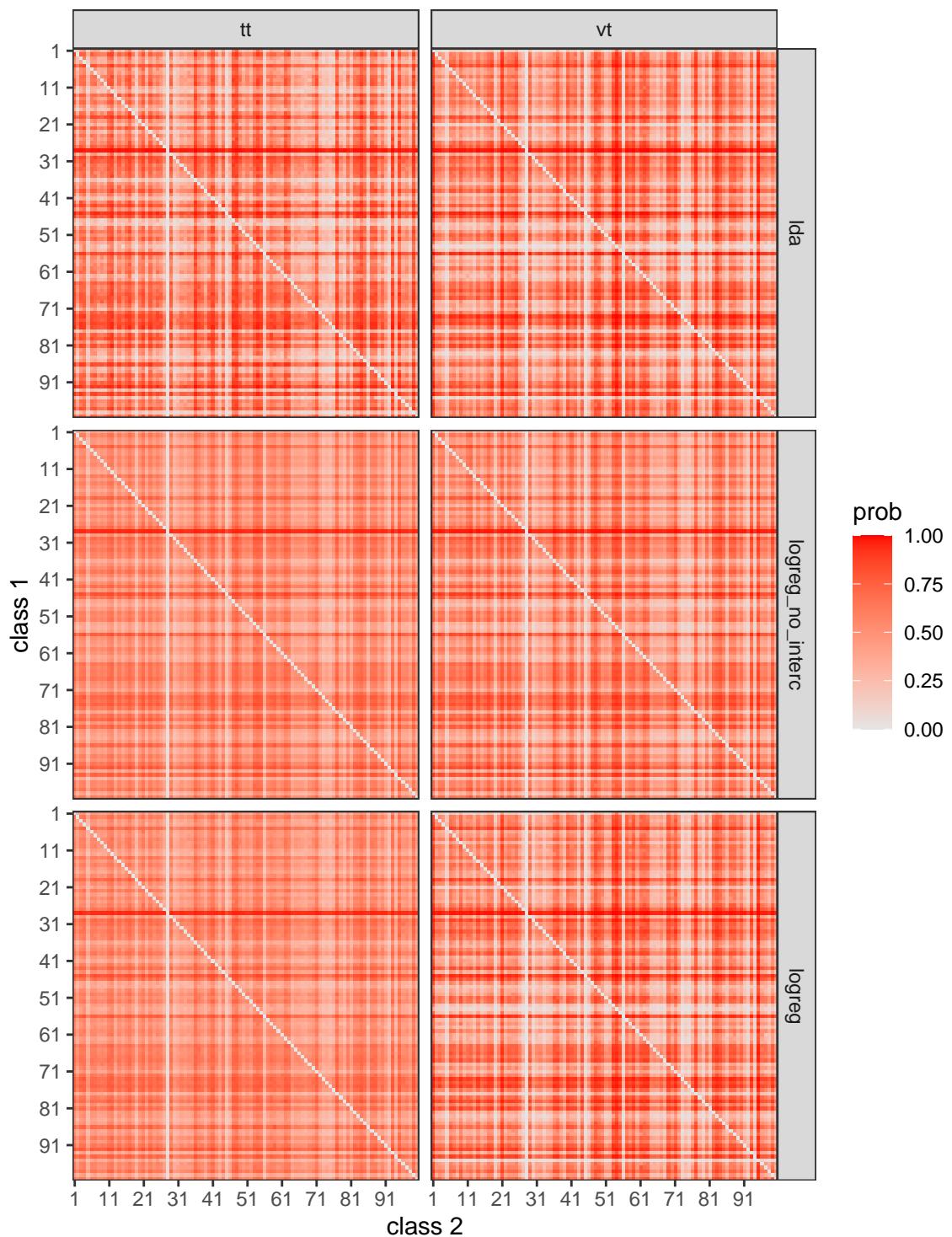
Average pairwise probabilities – class 26



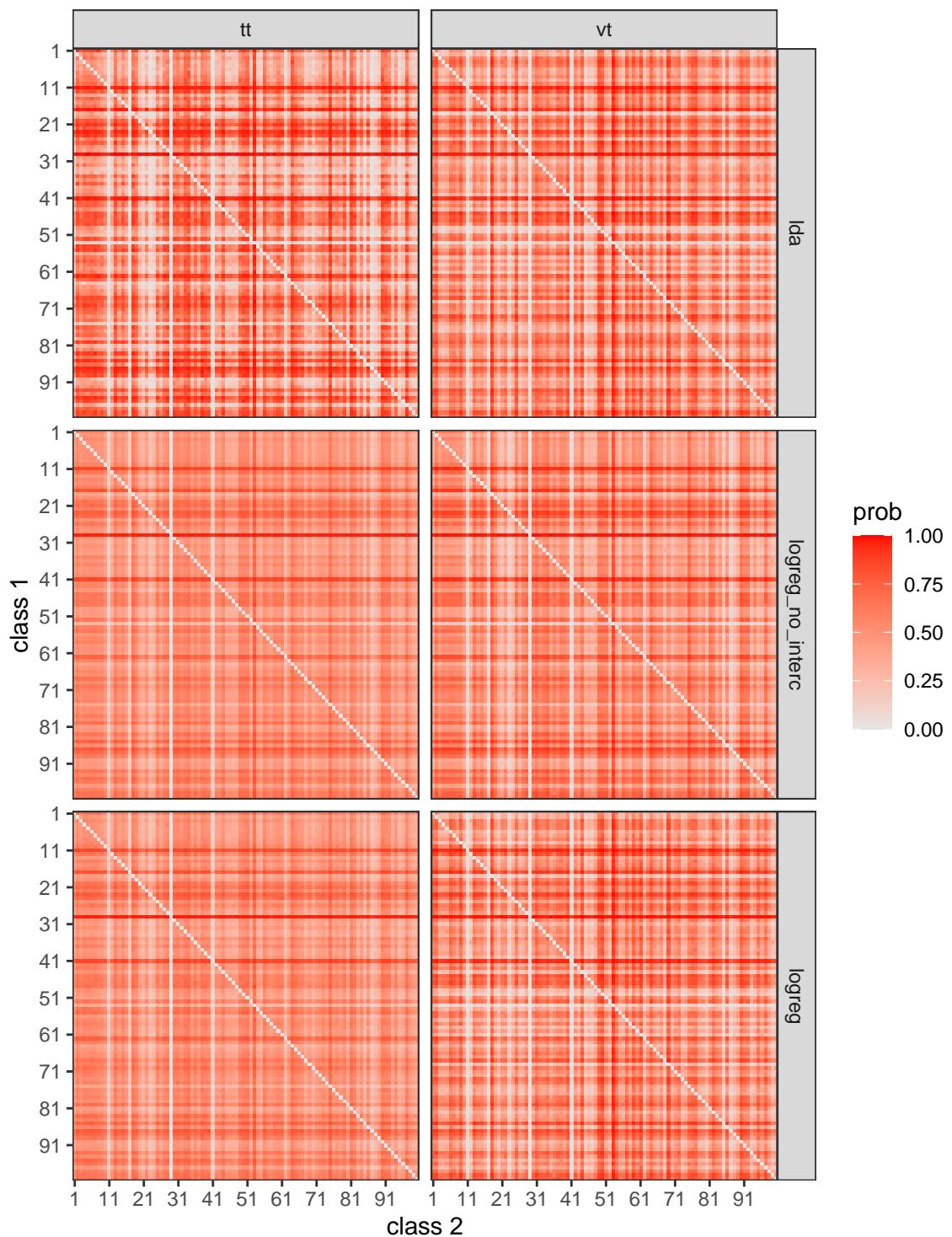
Average pairwise probabilities – class 27



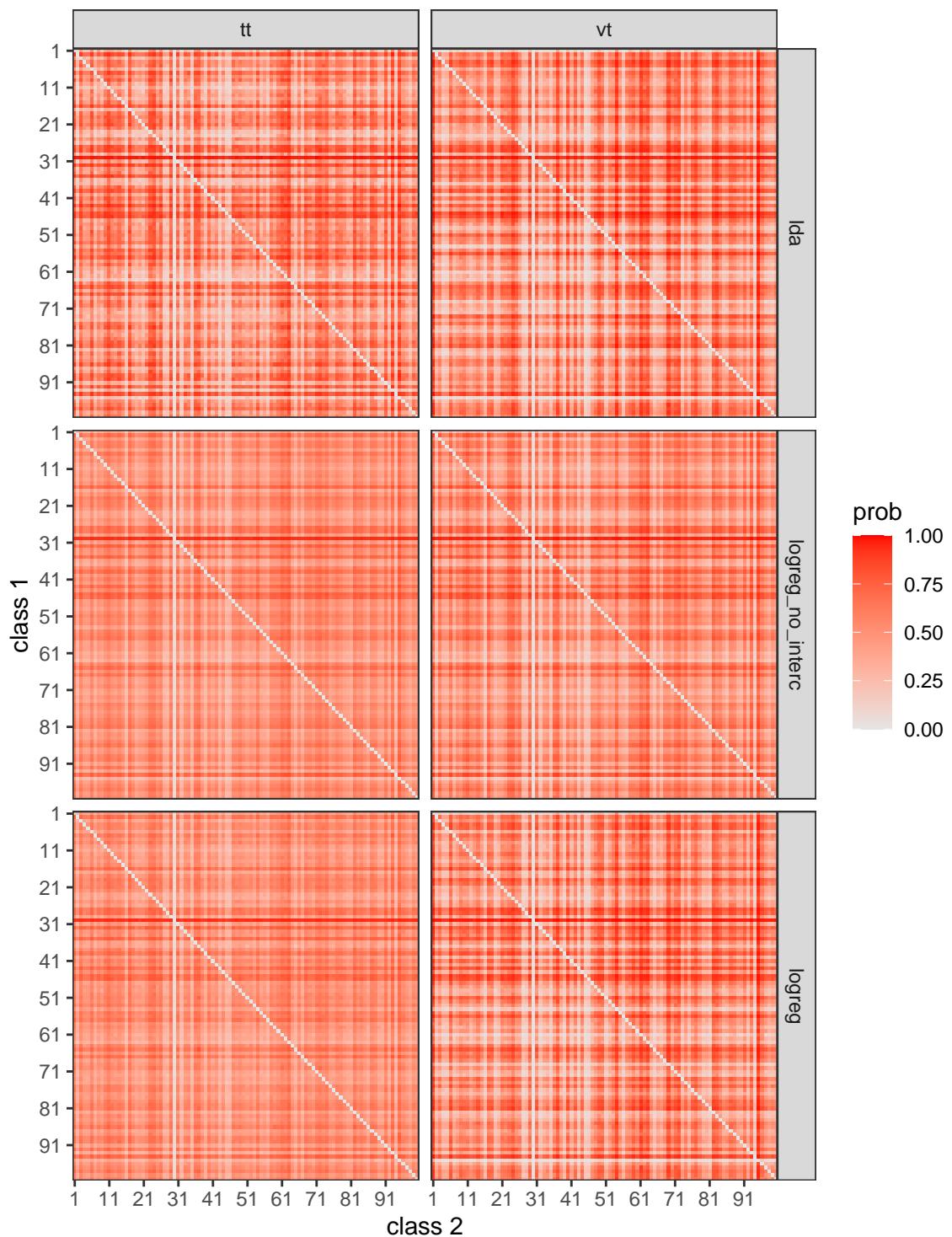
Average pairwise probabilities – class 28



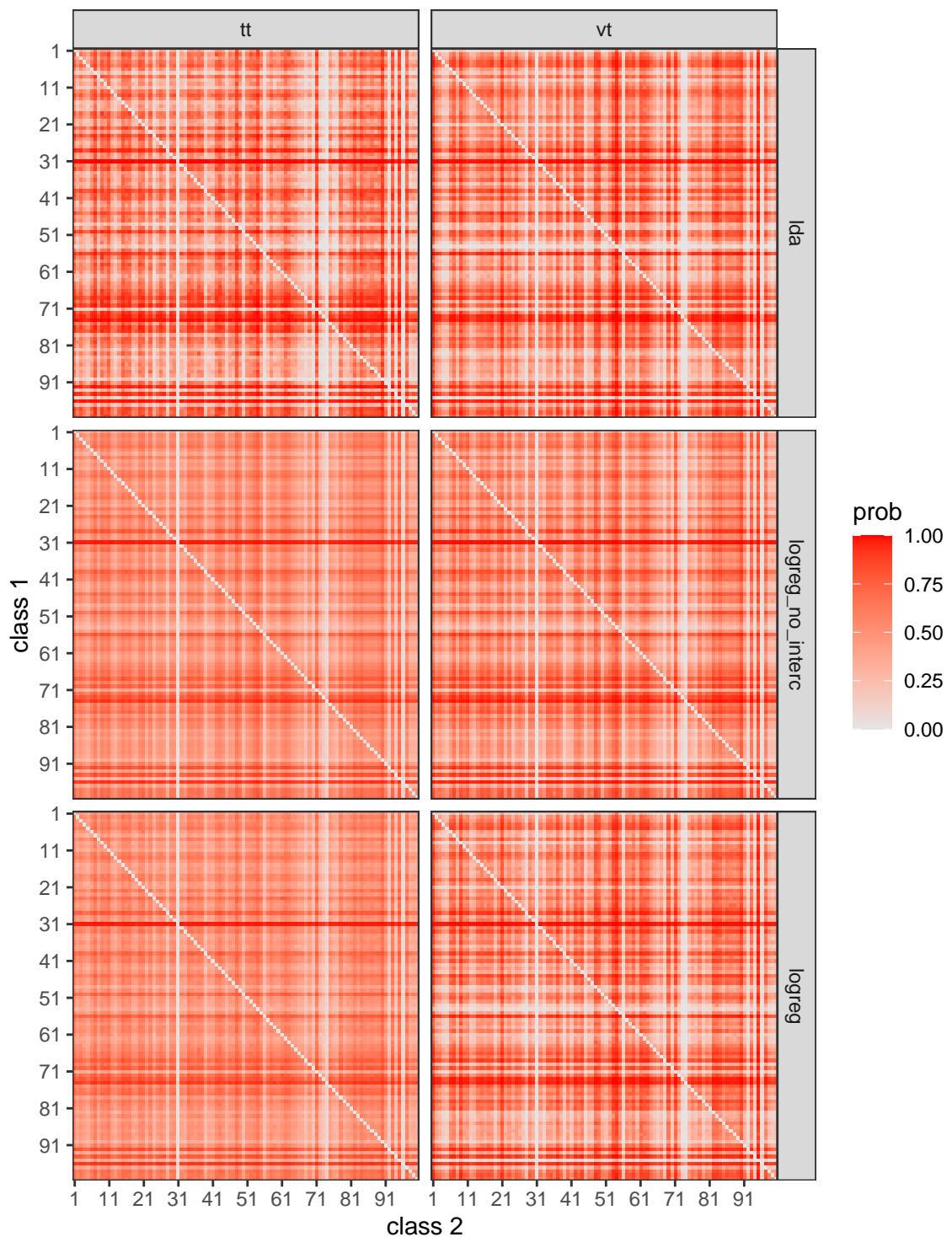
Average pairwise probabilities – class 29



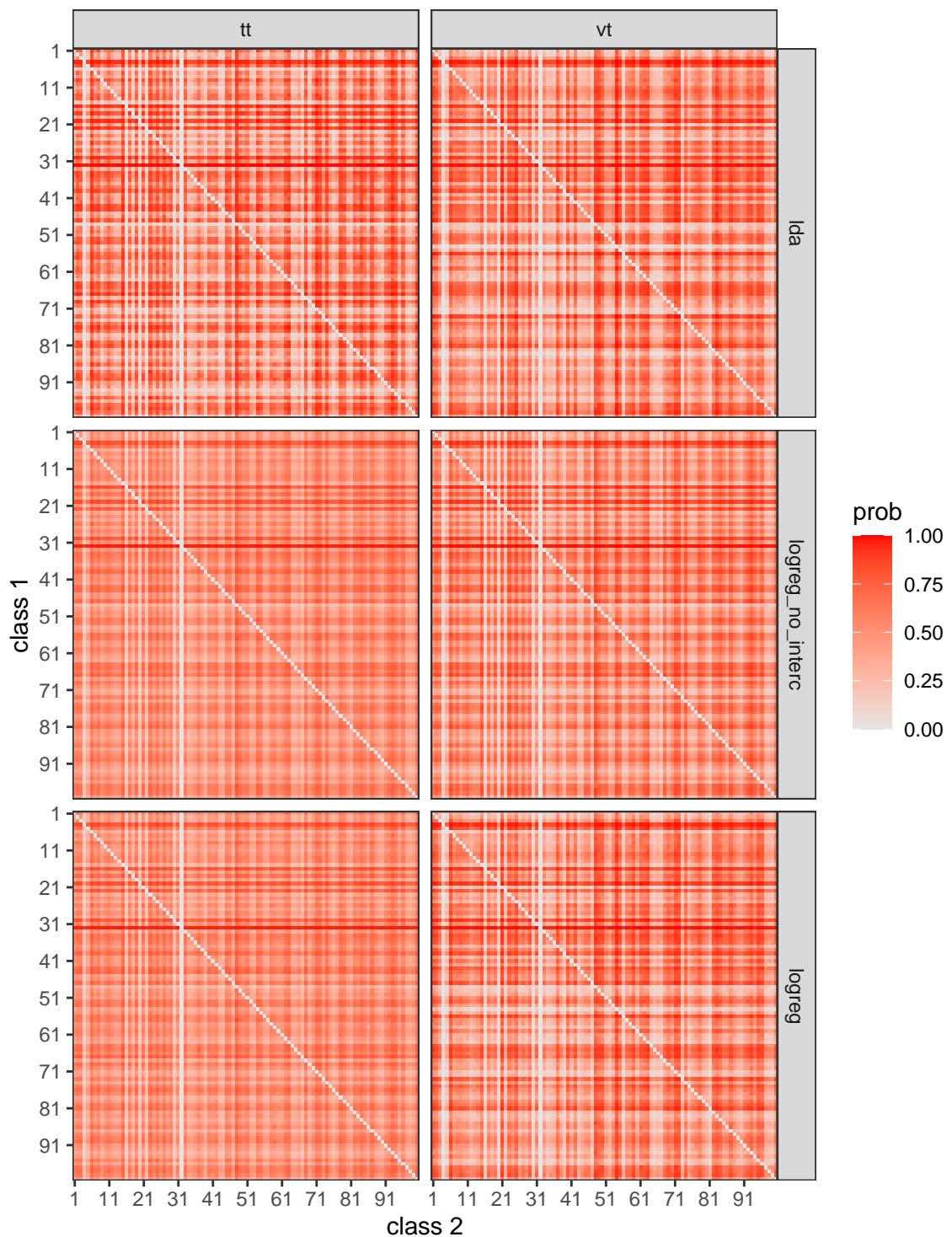
Average pairwise probabilities – class 30



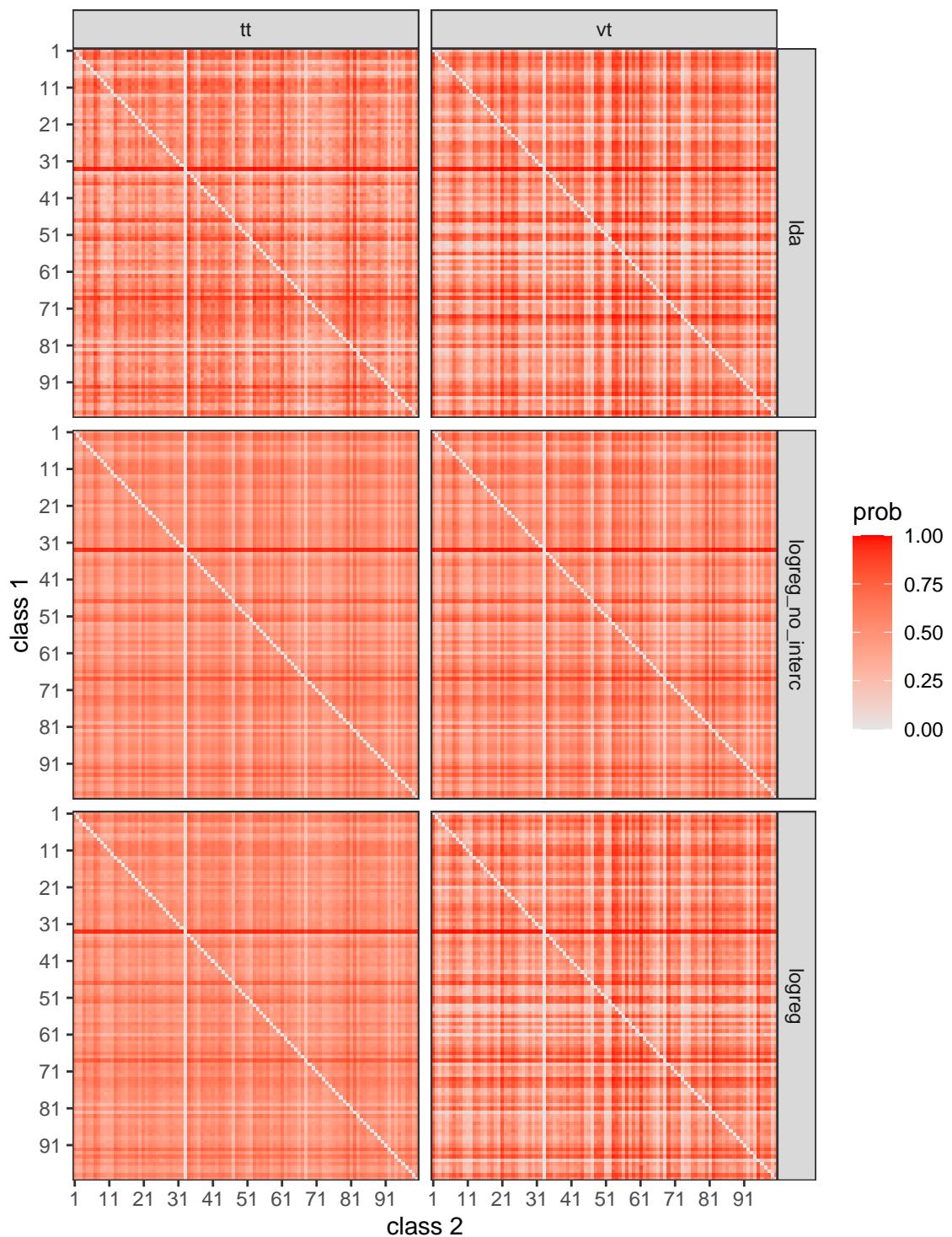
Average pairwise probabilities – class 31



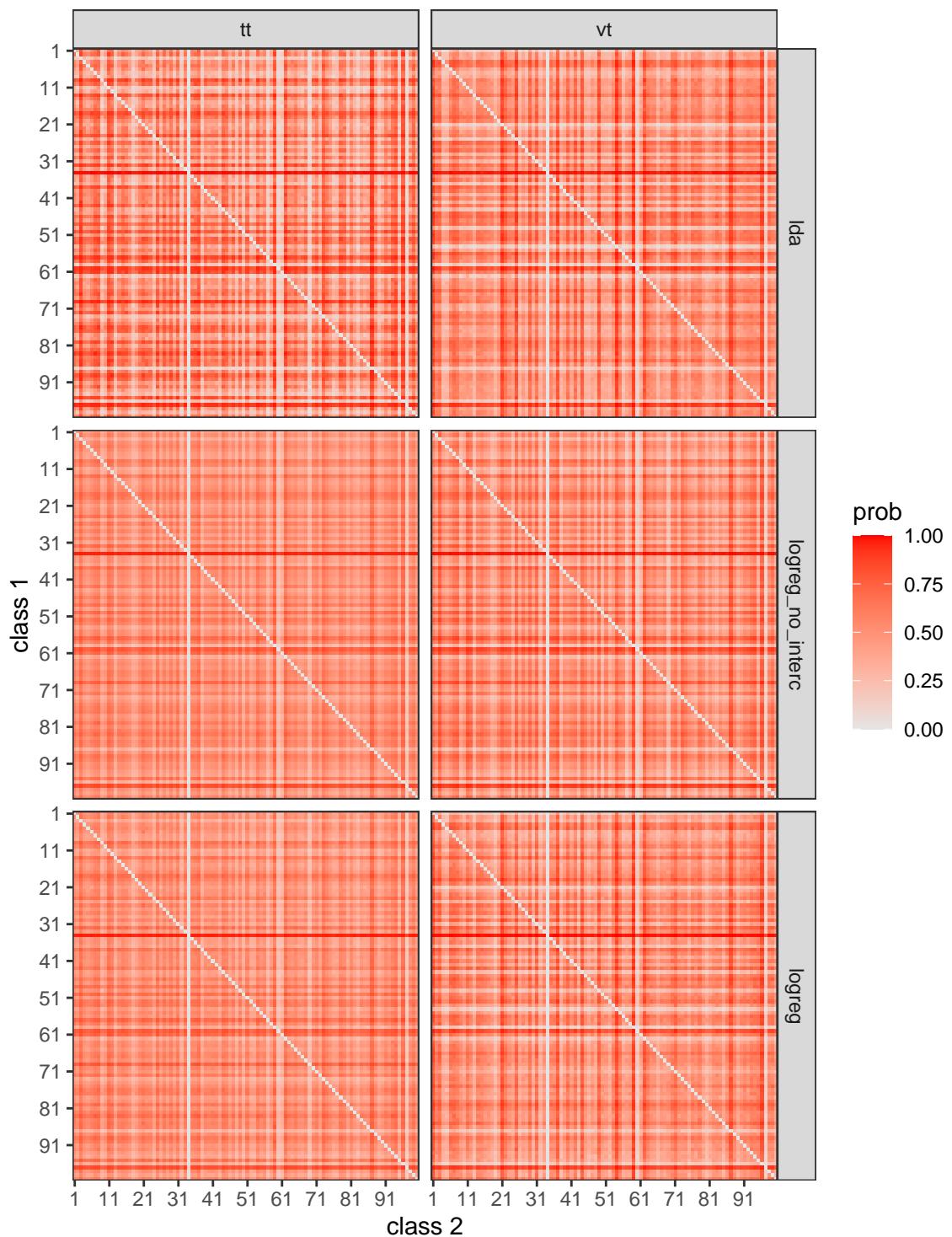
Average pairwise probabilities – class 32



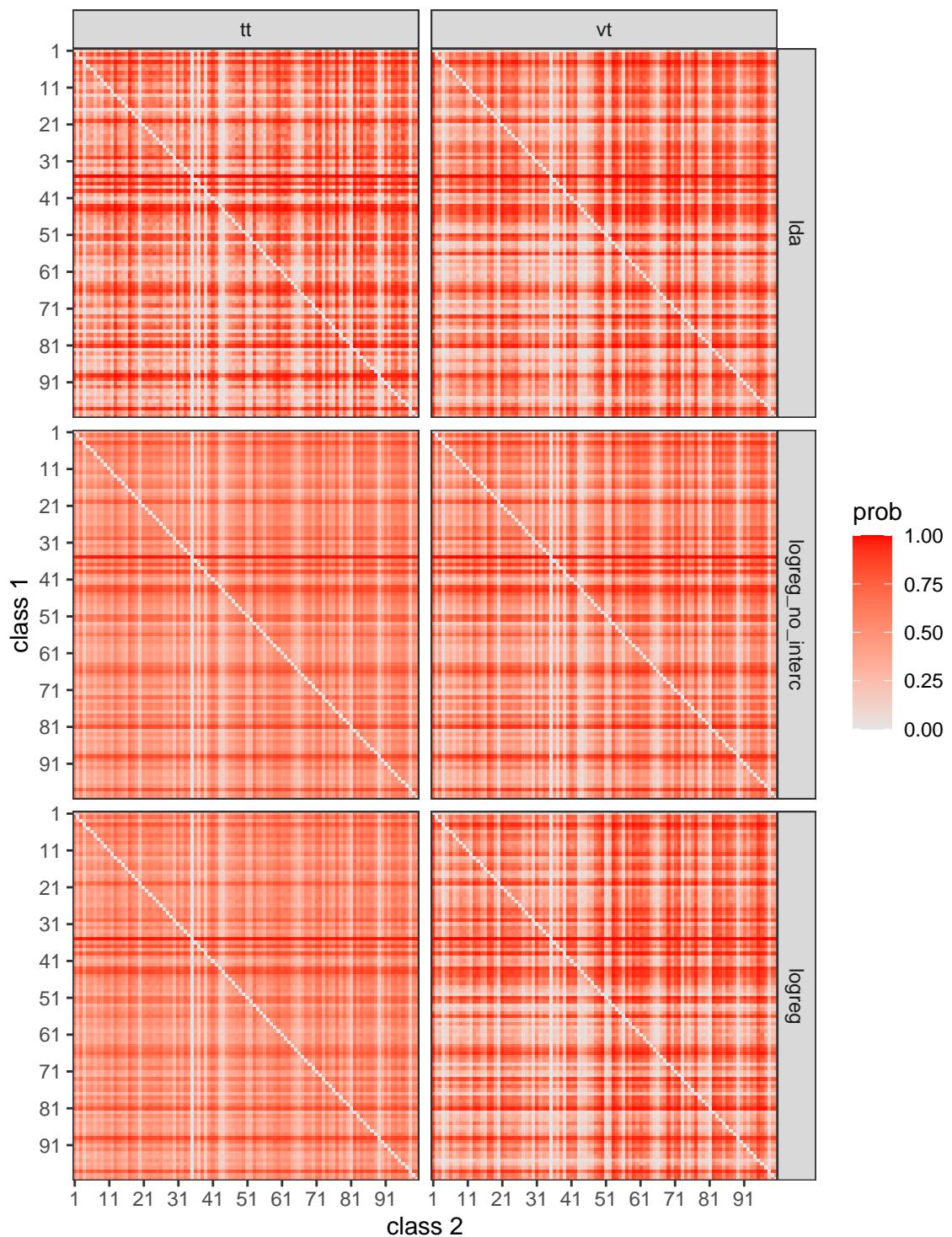
Average pairwise probabilities – class 33



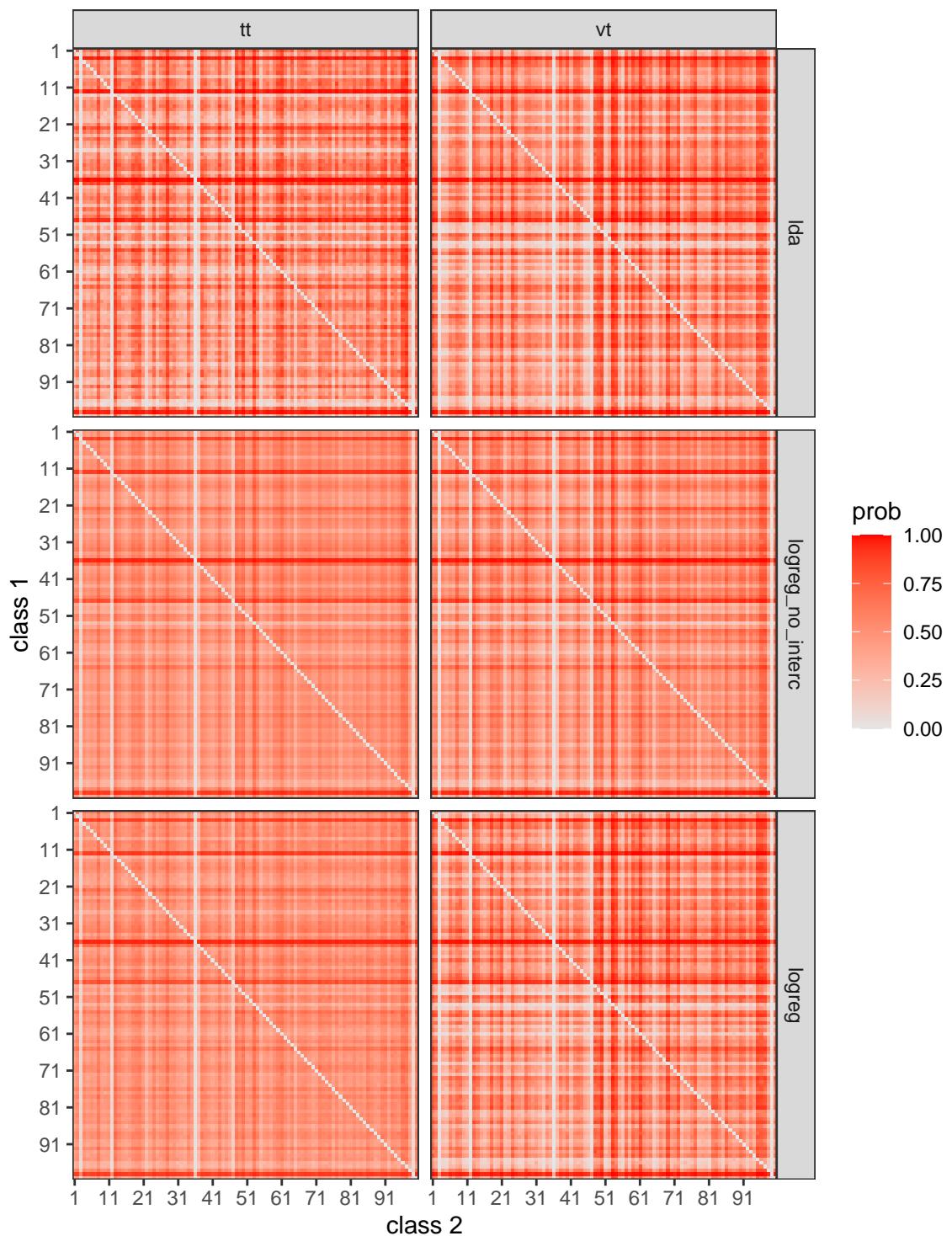
Average pairwise probabilities – class 34



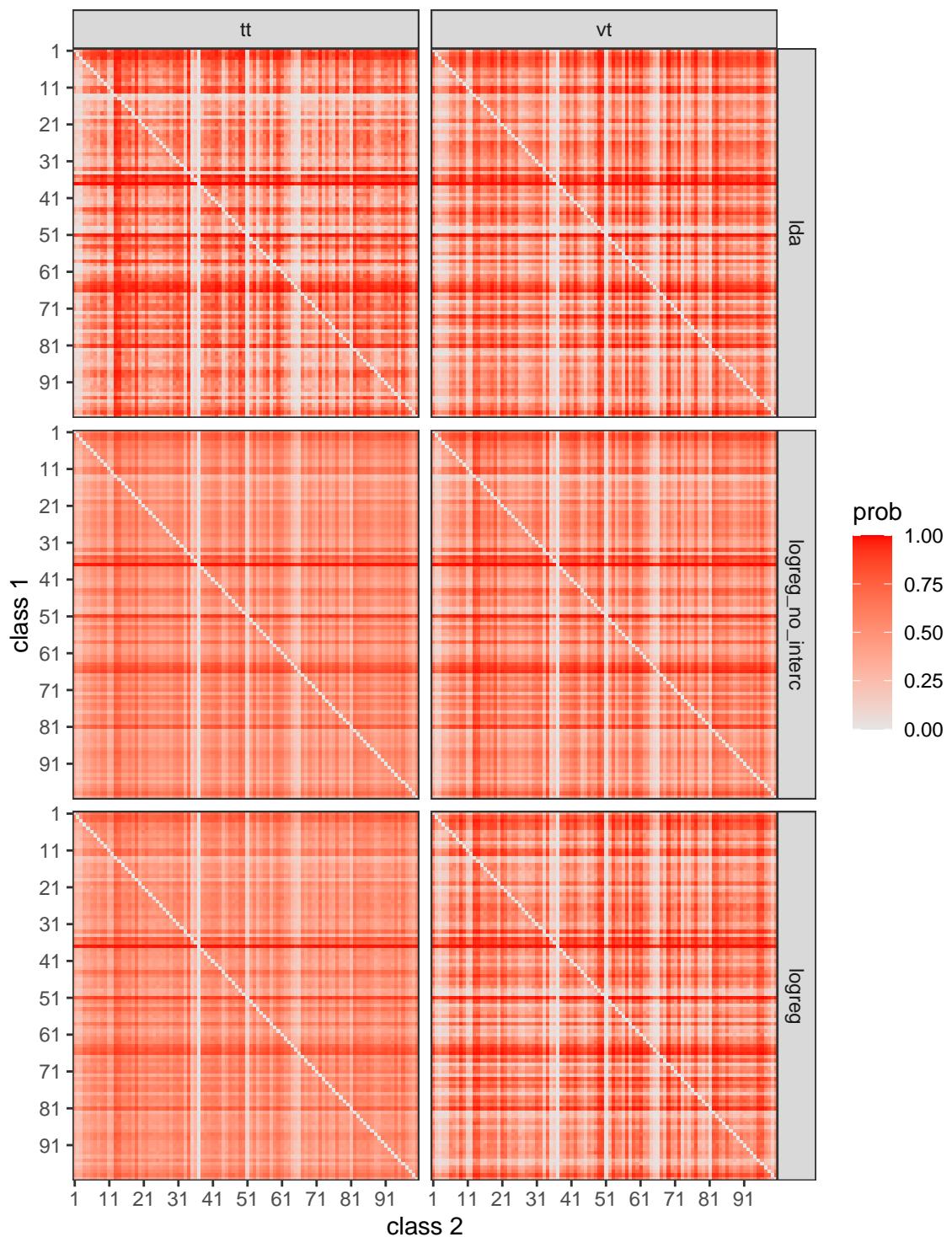
Average pairwise probabilities – class 35



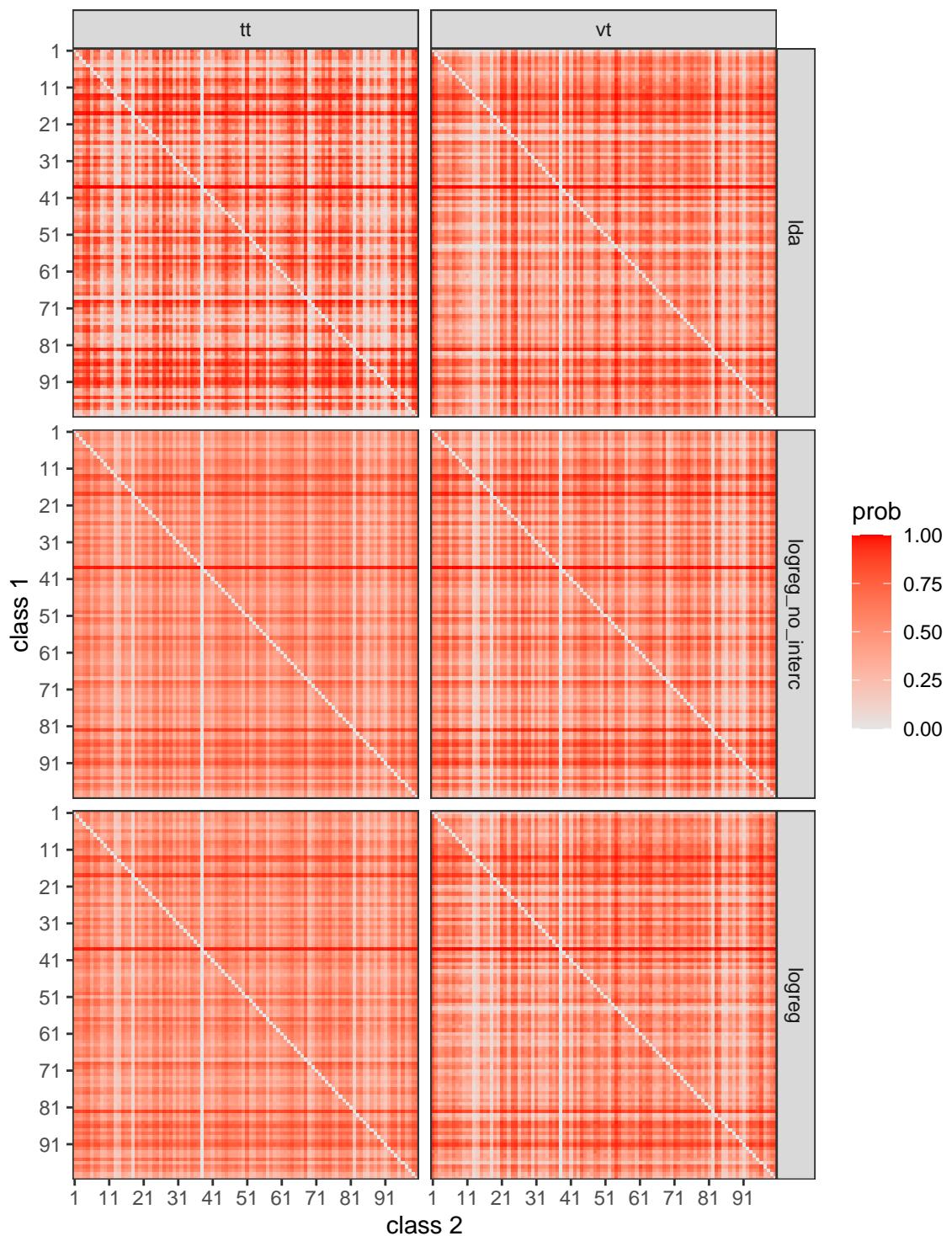
Average pairwise probabilities – class 36



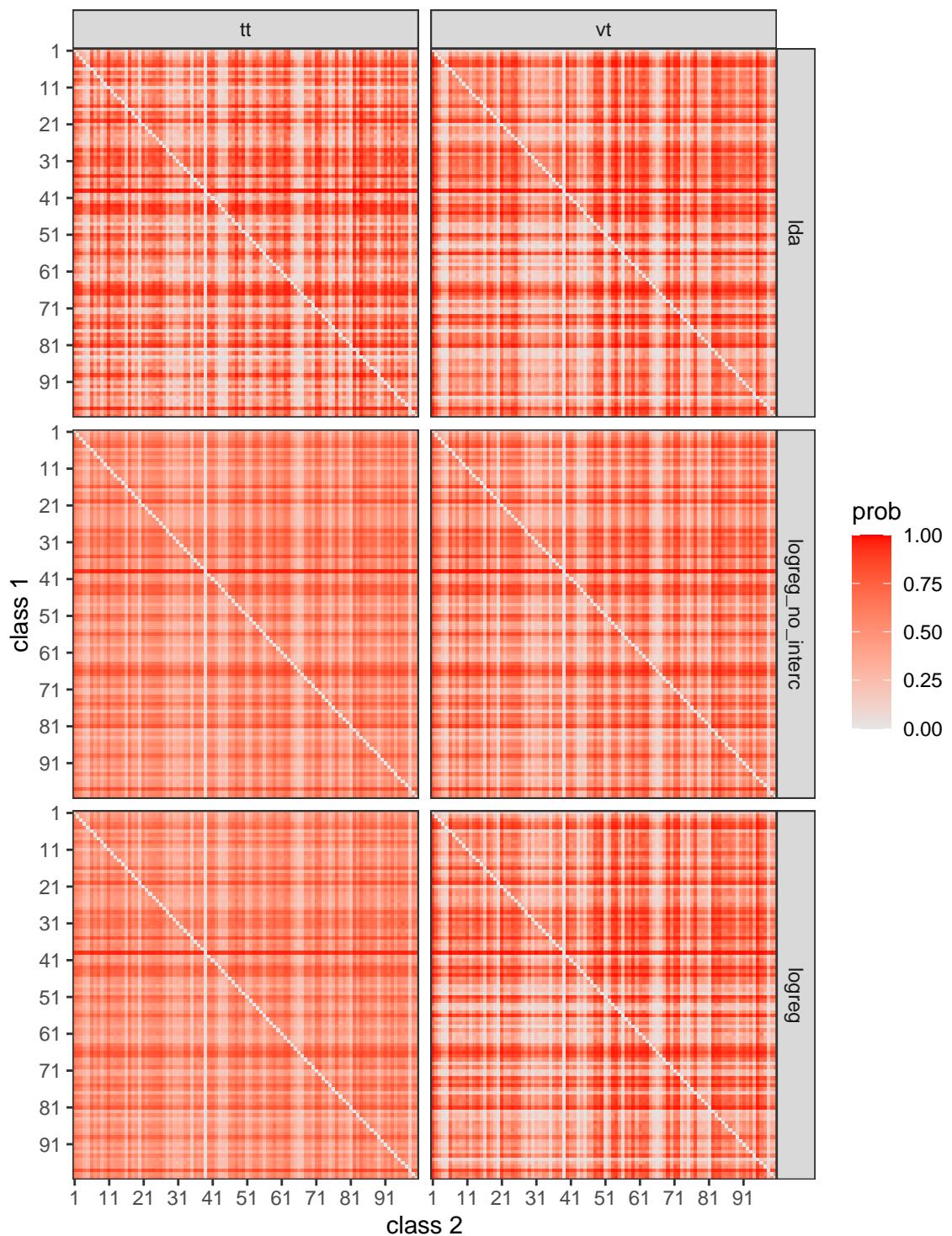
Average pairwise probabilities – class 37



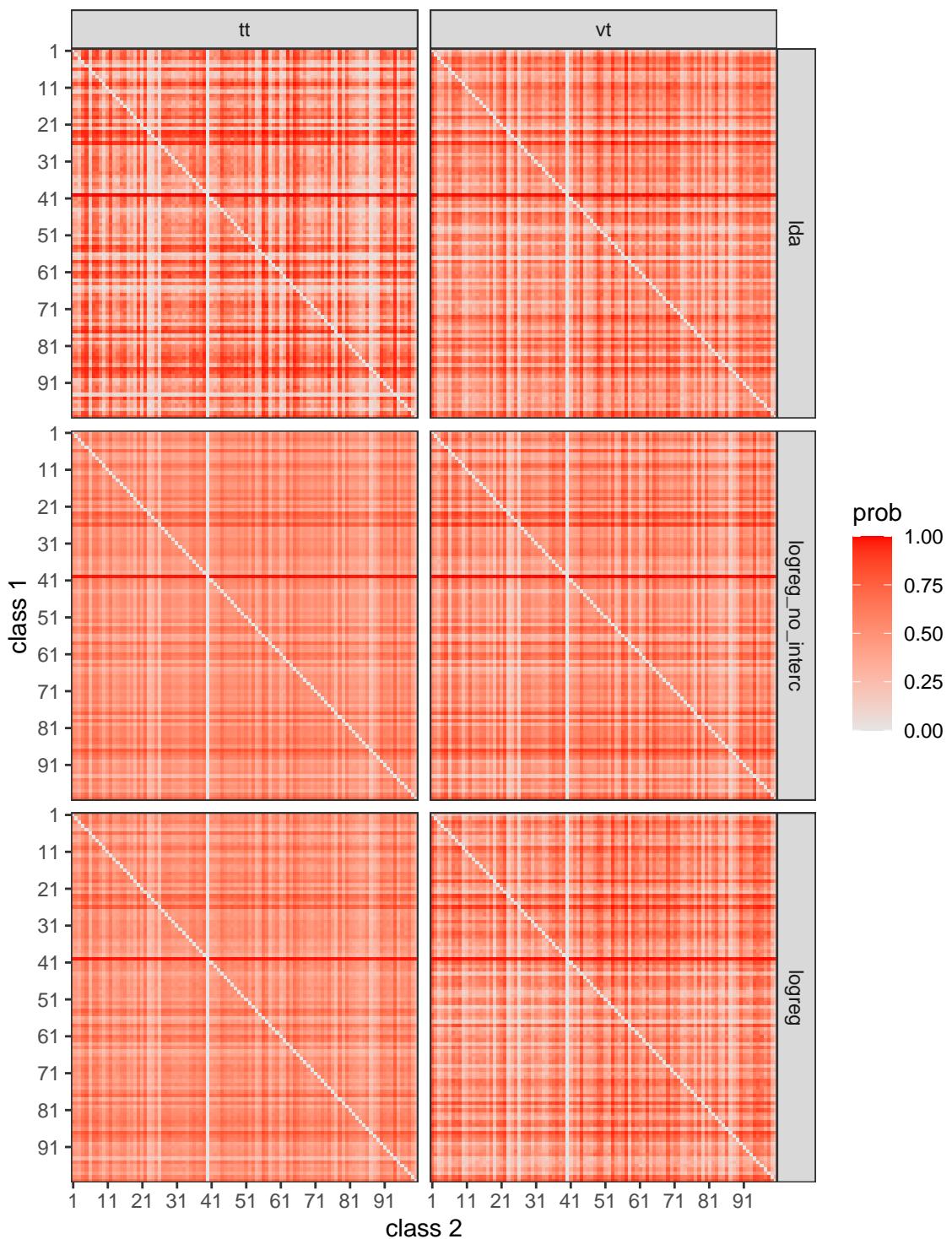
Average pairwise probabilities – class 38



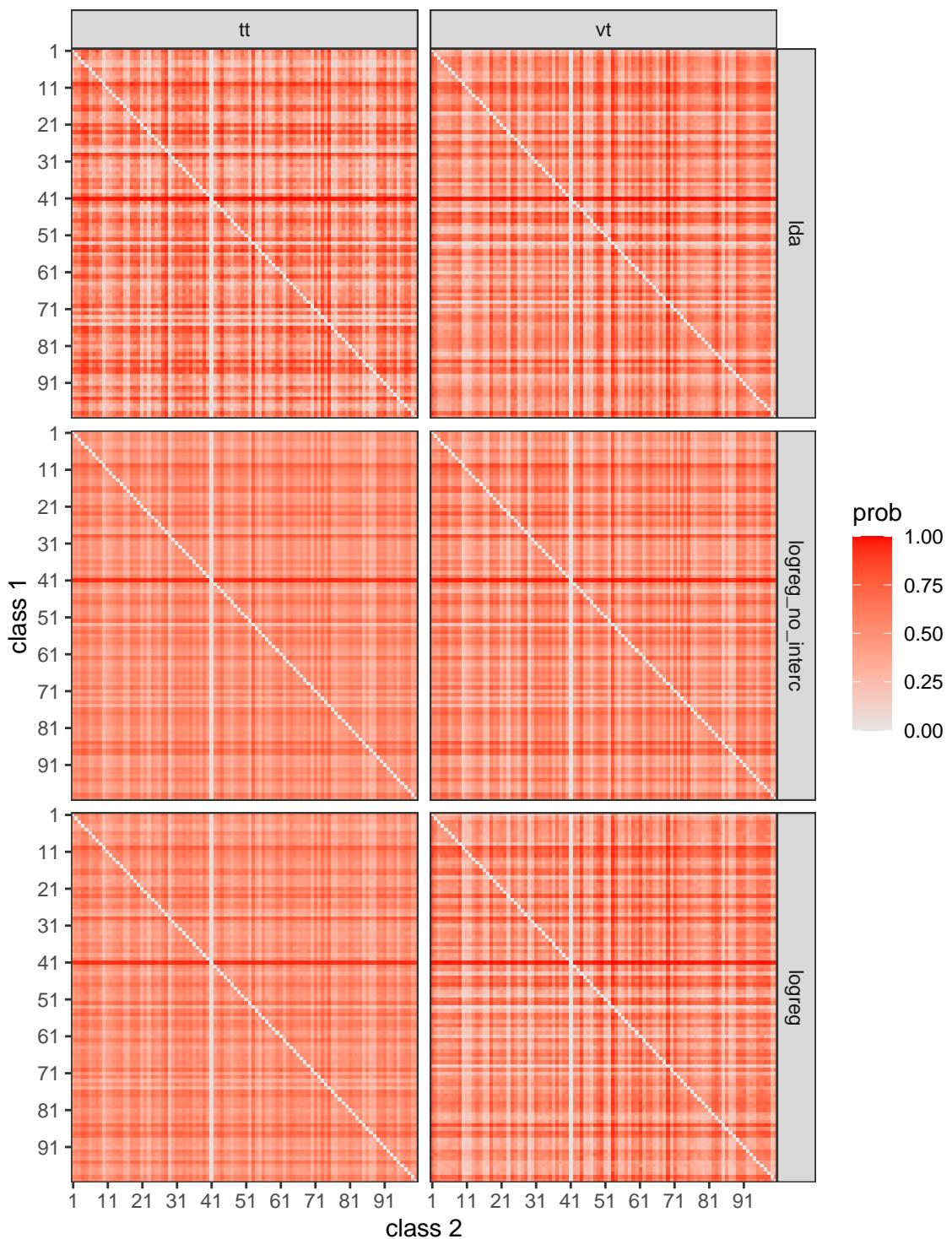
Average pairwise probabilities – class 39



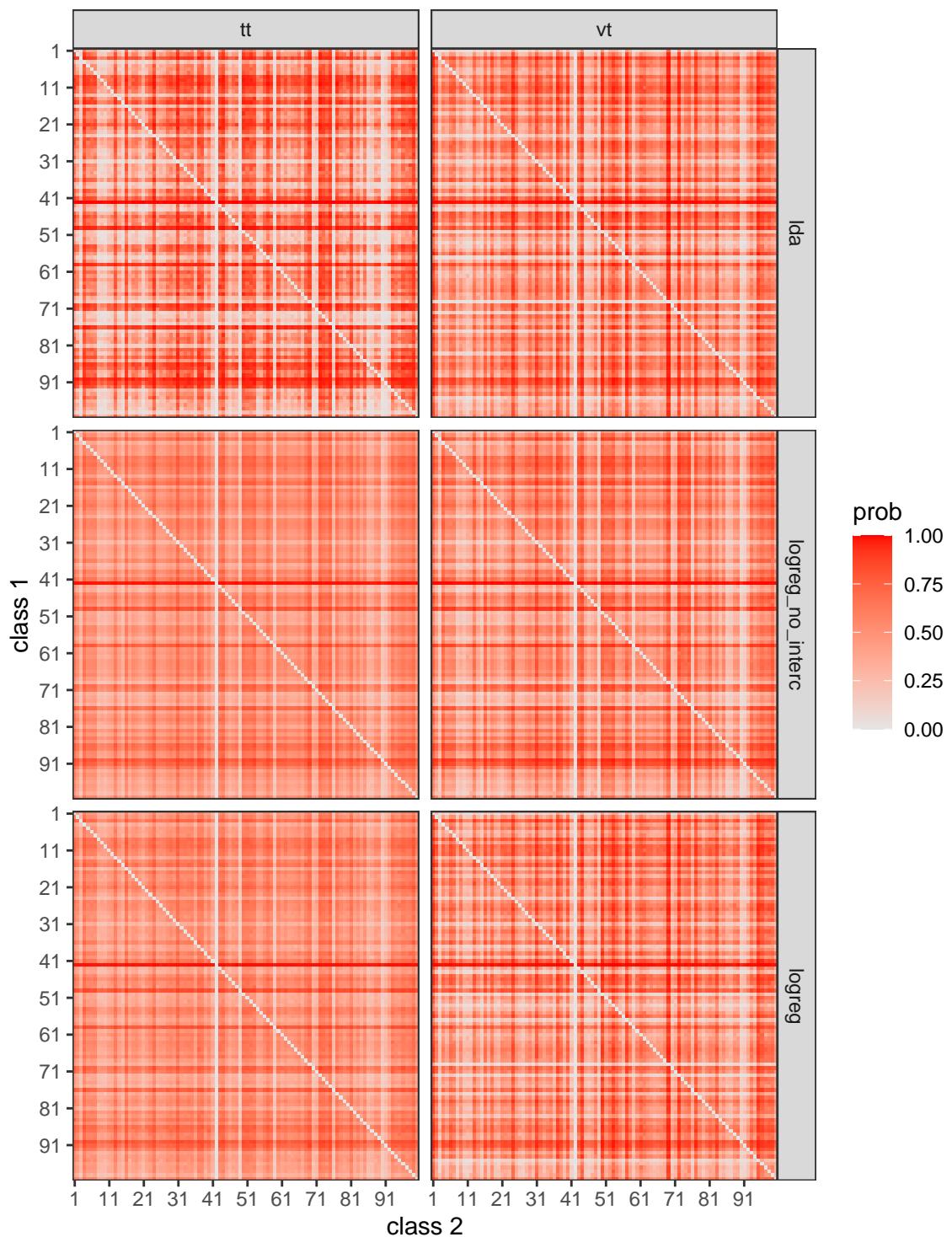
Average pairwise probabilities – class 40



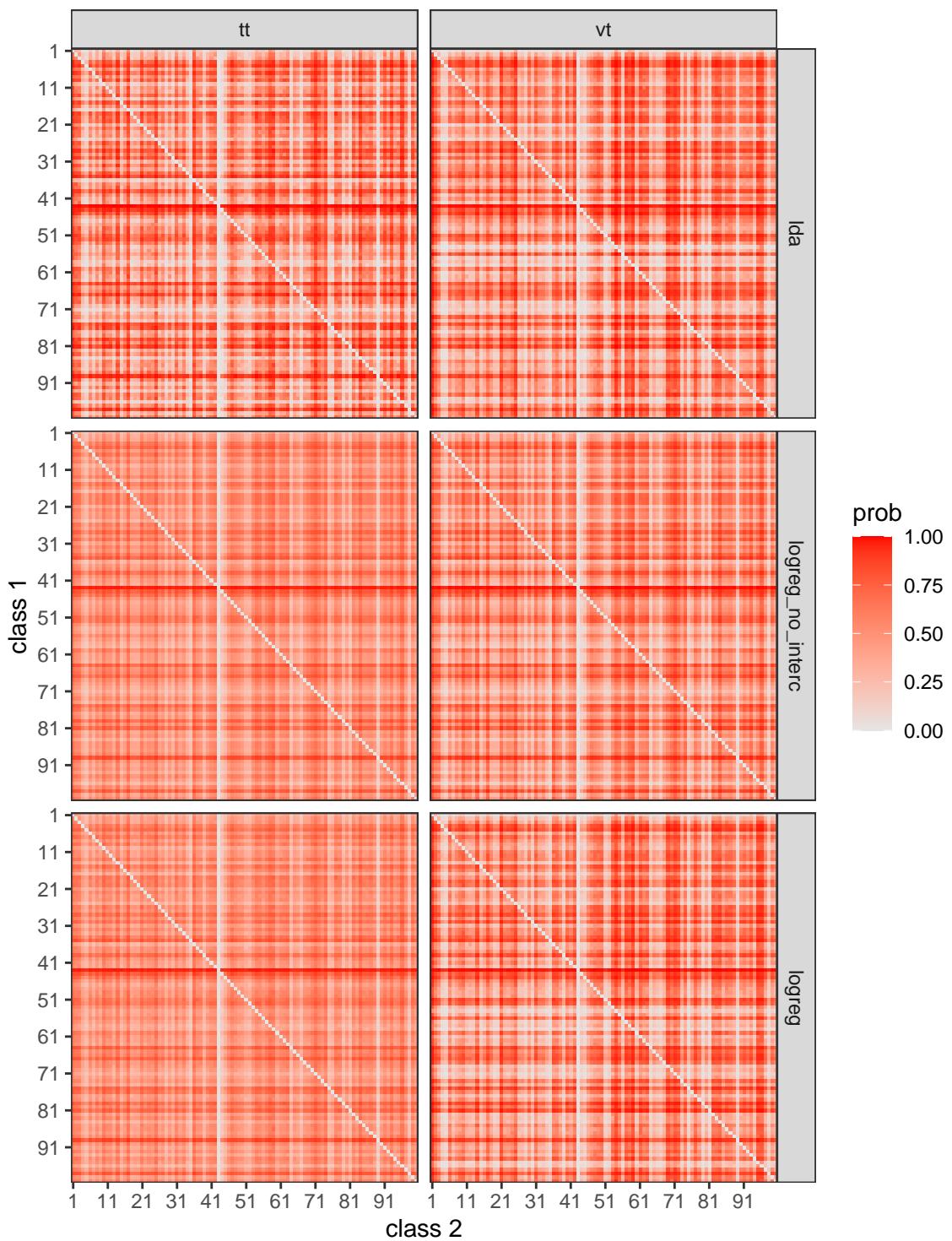
Average pairwise probabilities – class 41



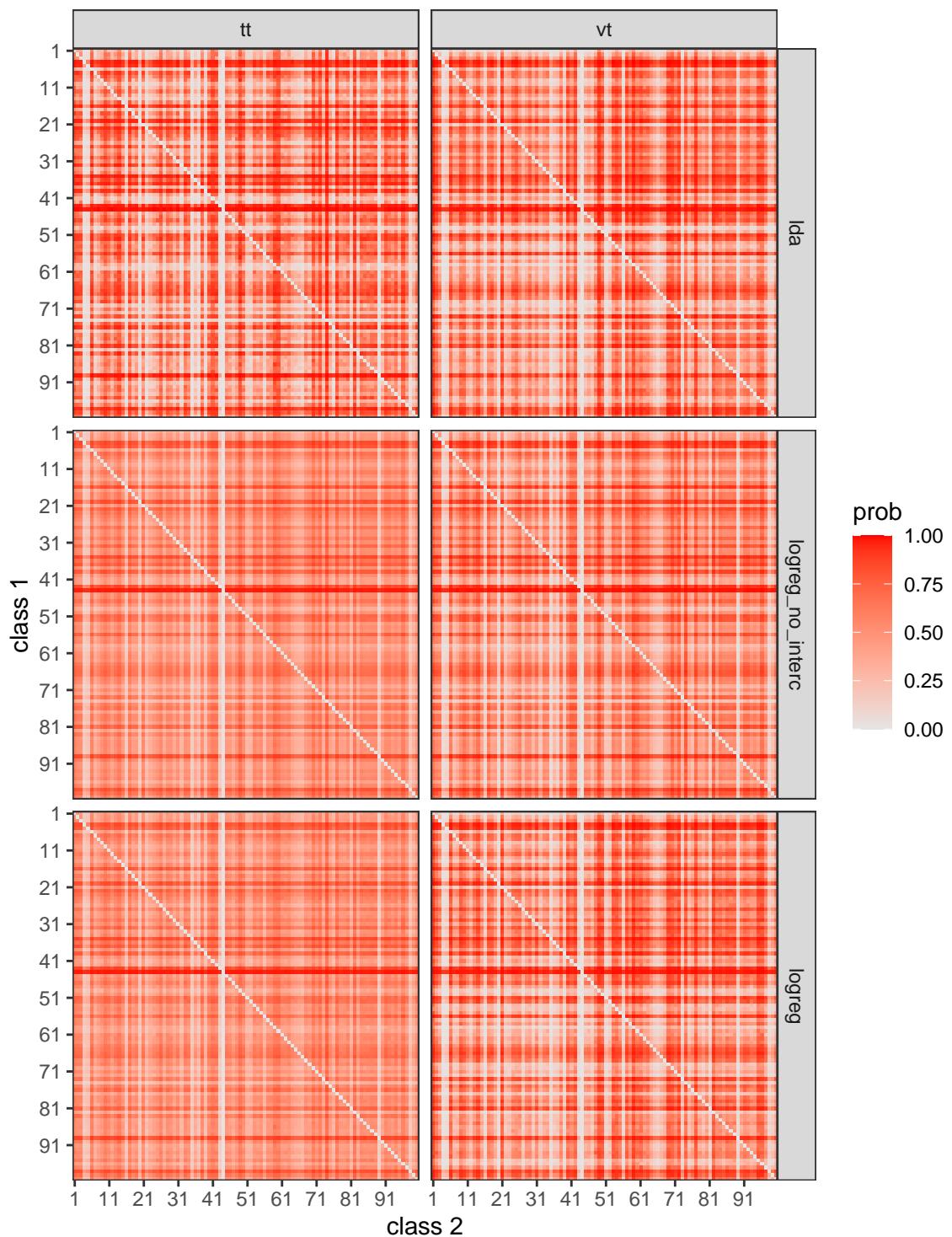
Average pairwise probabilities – class 42



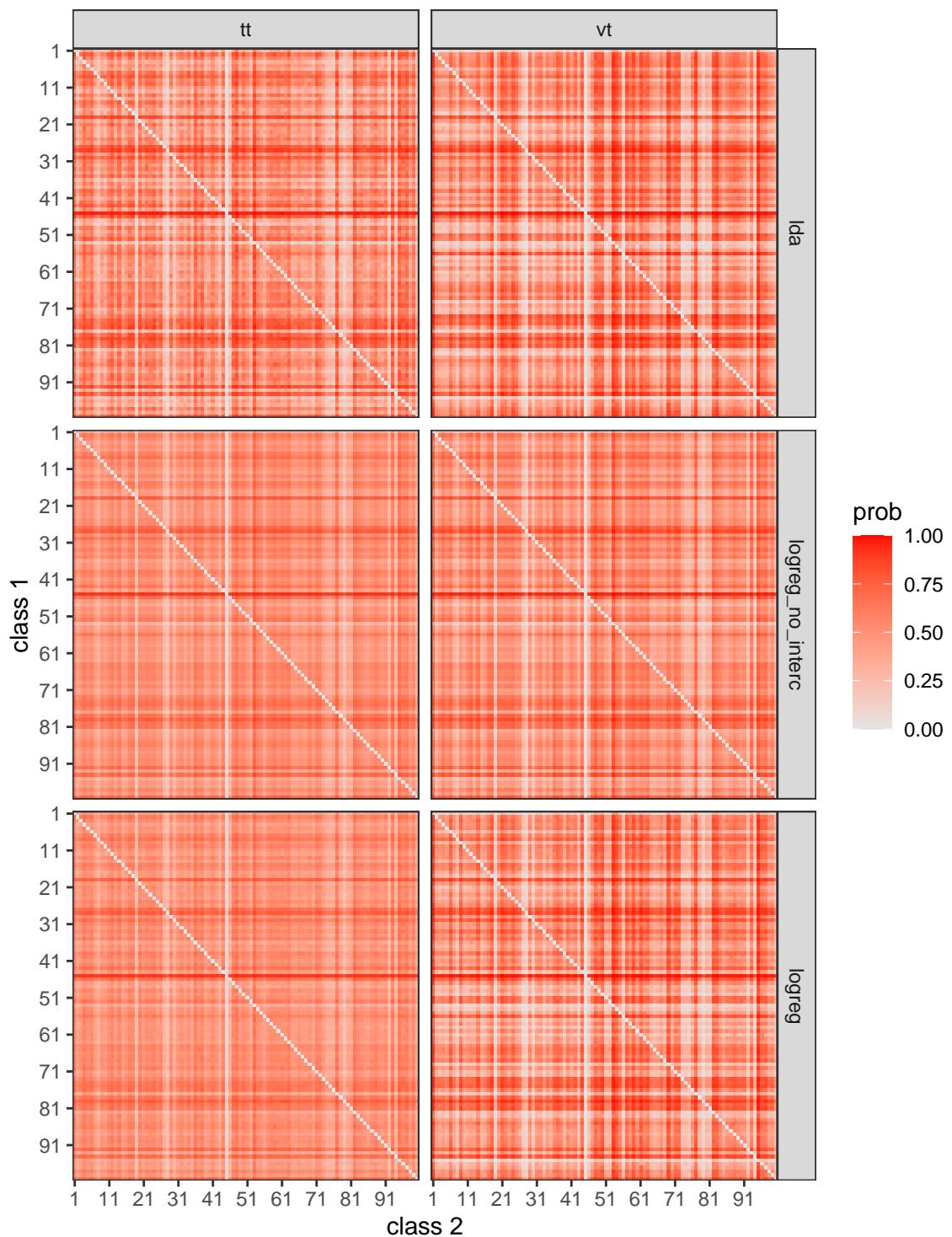
Average pairwise probabilities – class 43



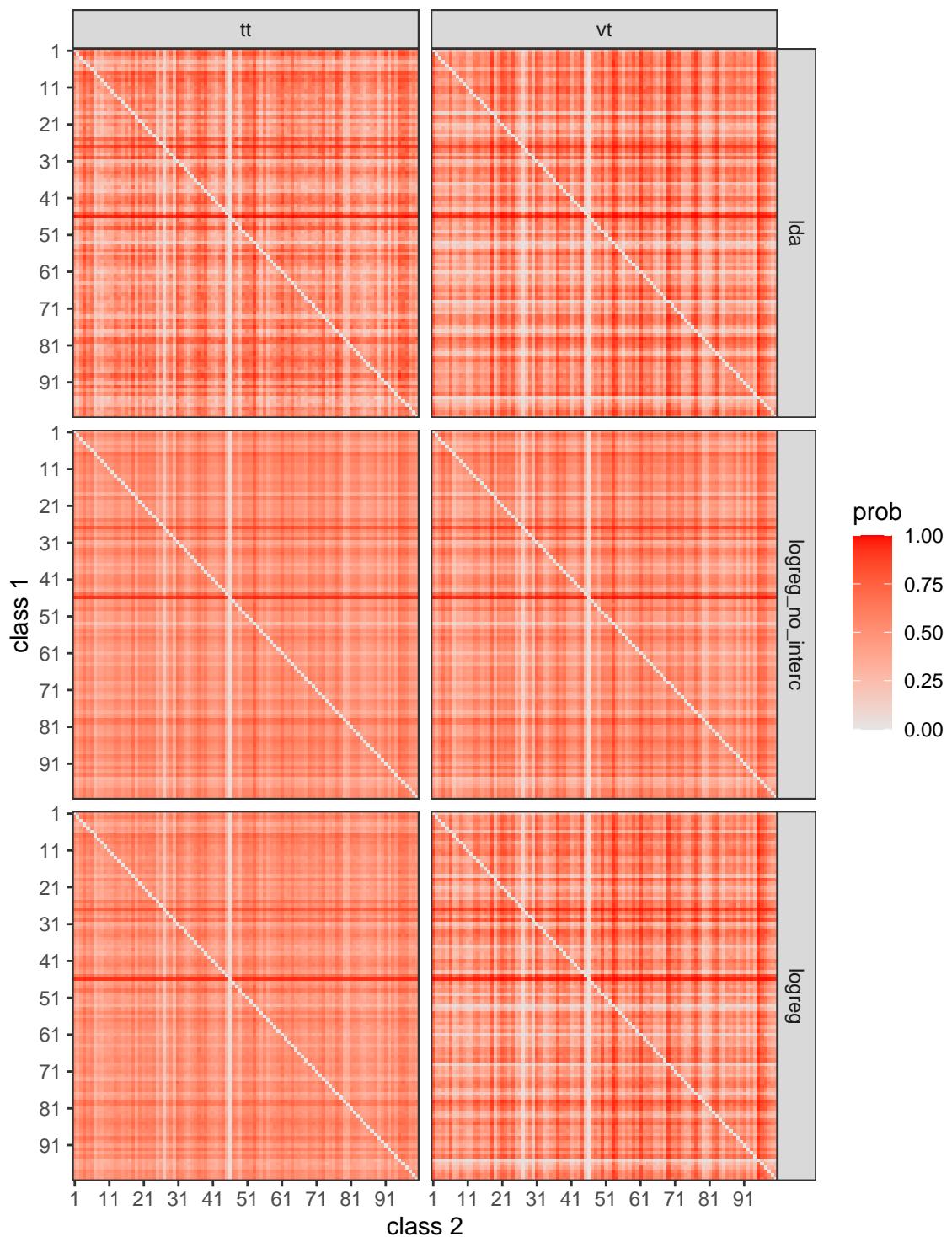
Average pairwise probabilities – class 44



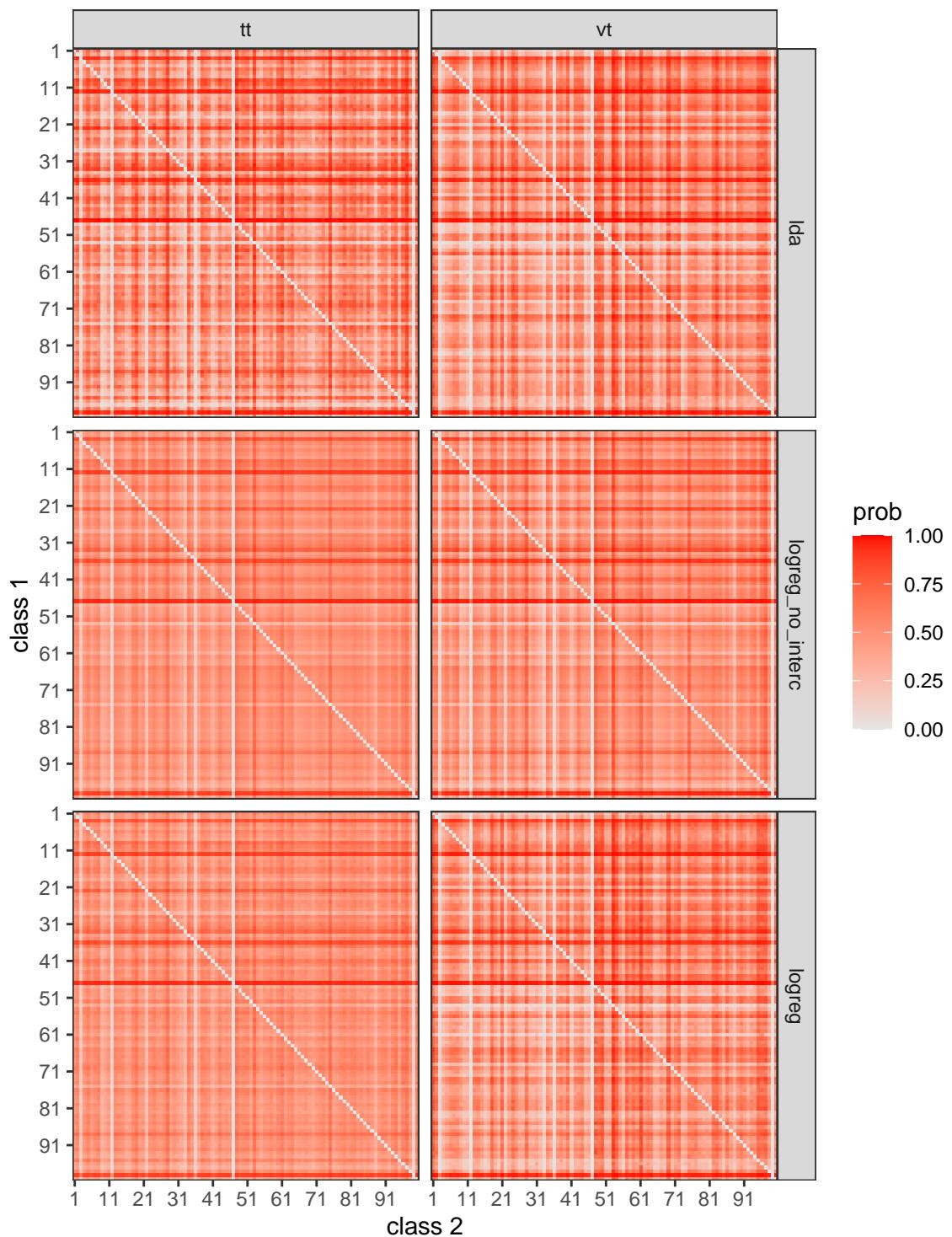
Average pairwise probabilities – class 45



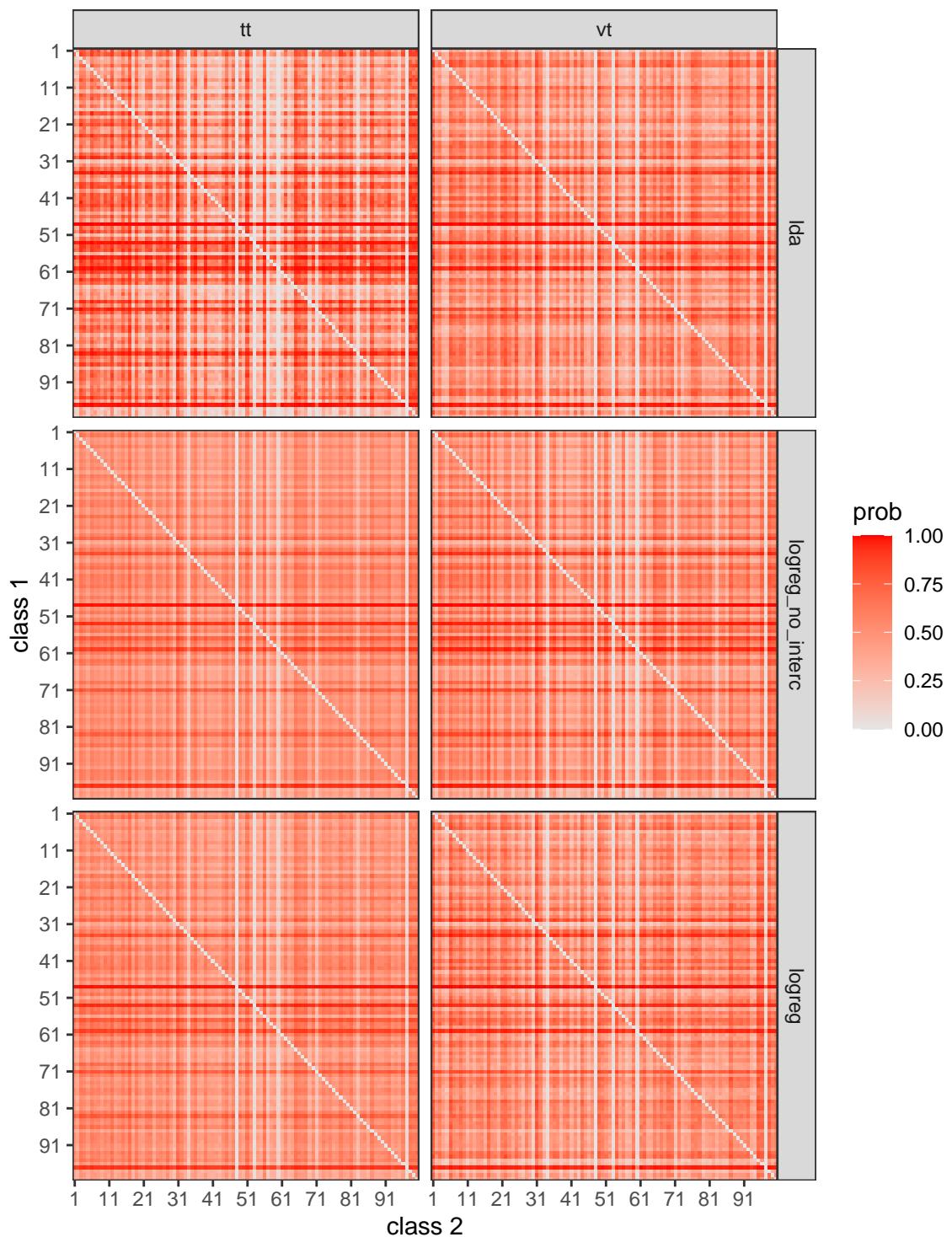
Average pairwise probabilities – class 46



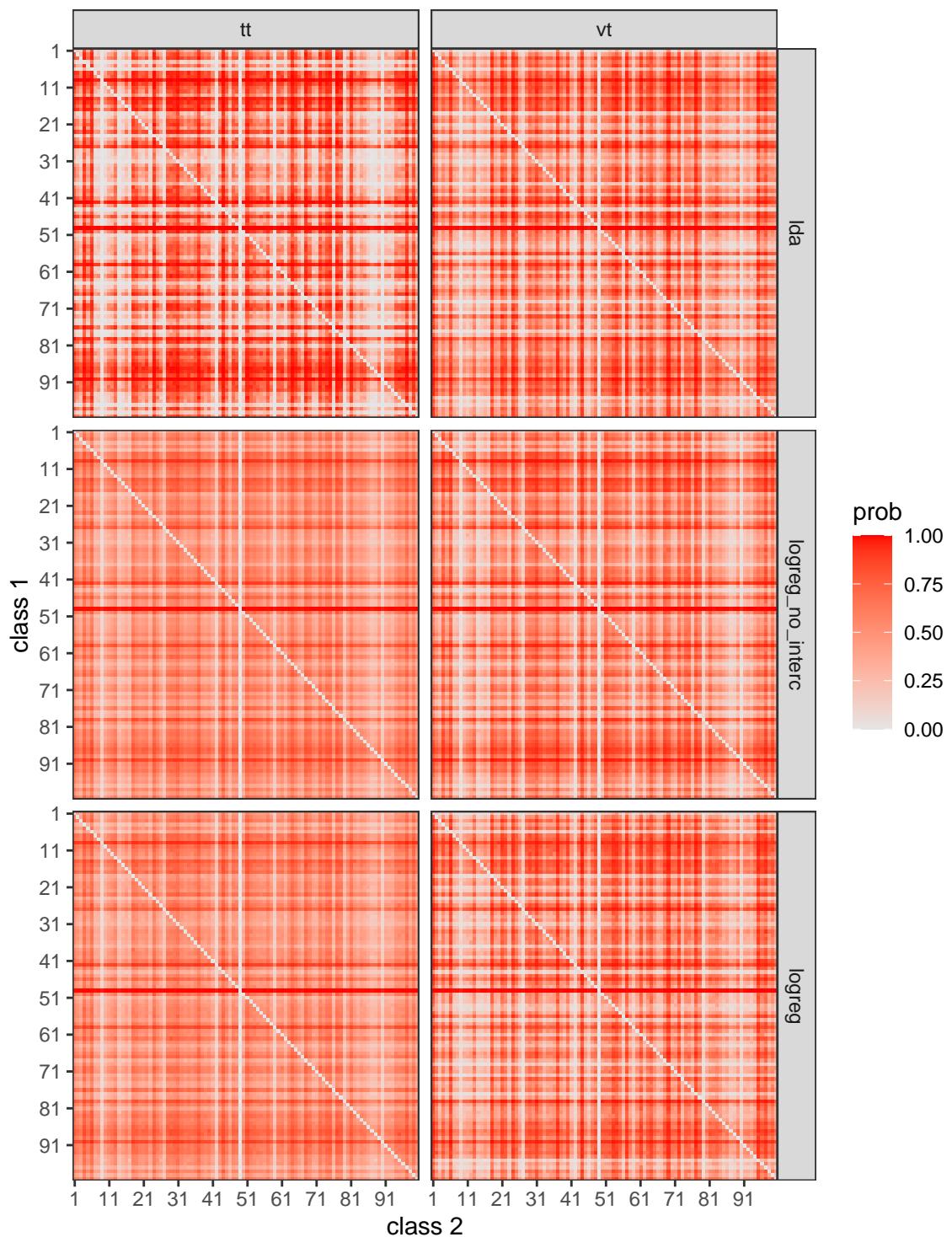
Average pairwise probabilities – class 47



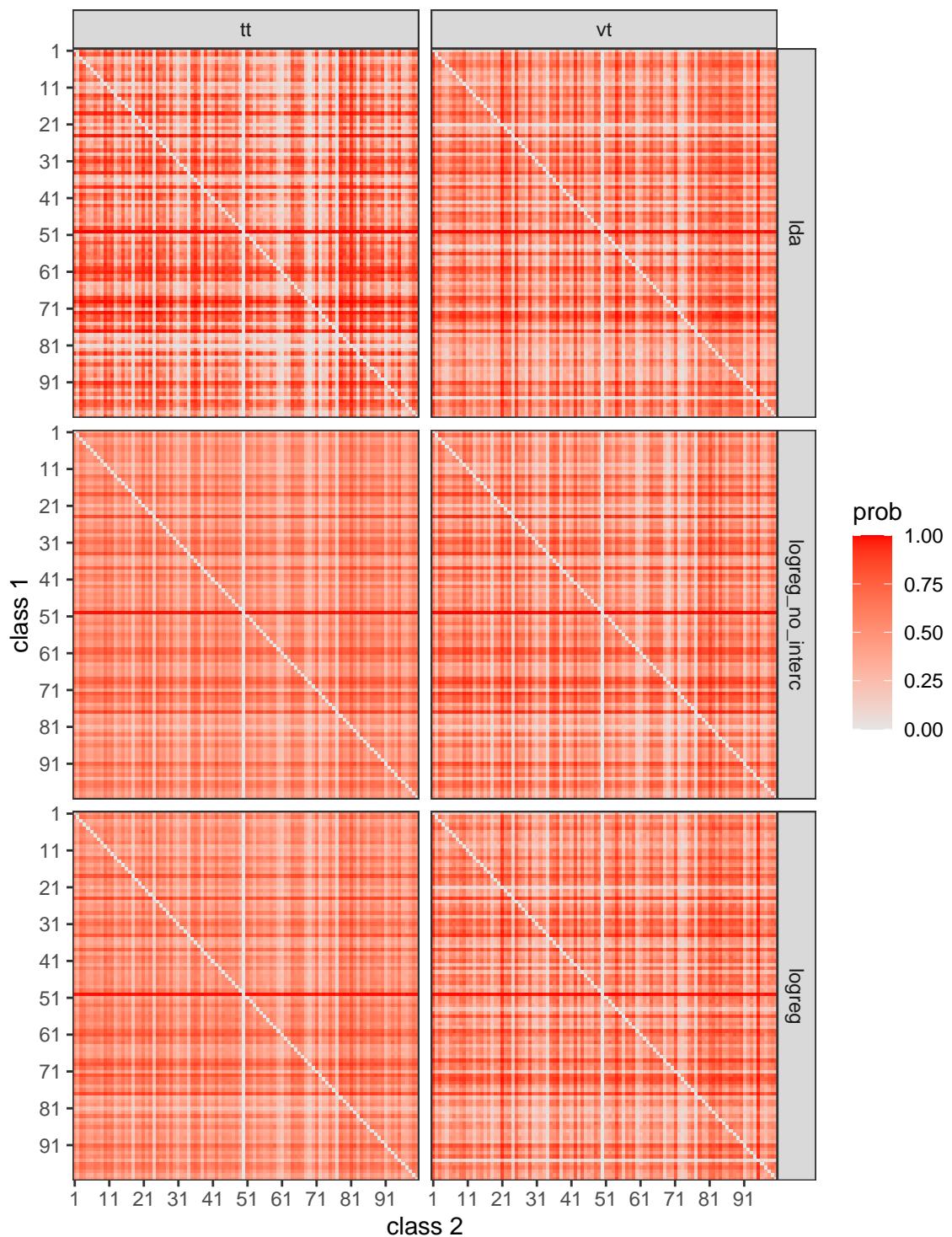
Average pairwise probabilities – class 48



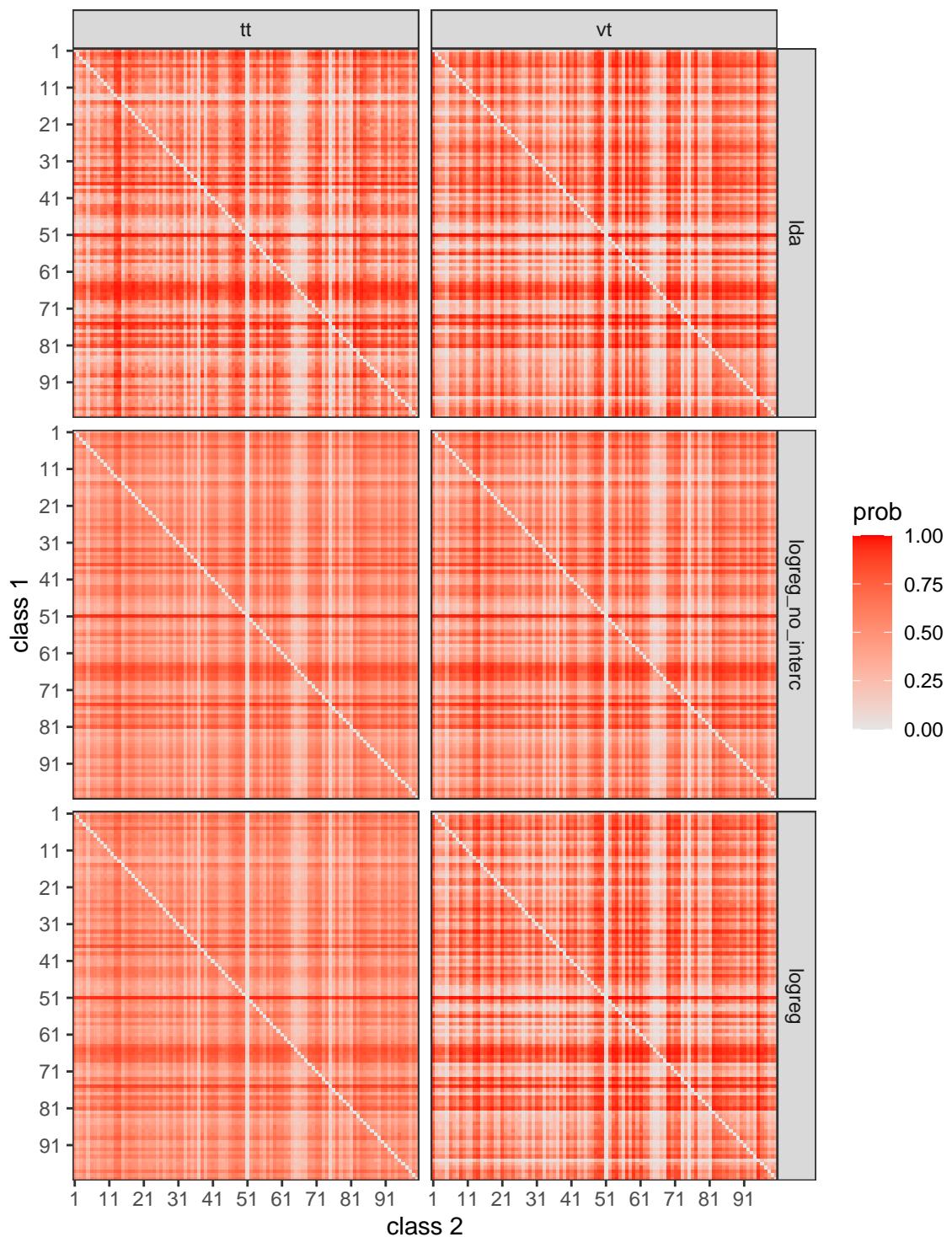
Average pairwise probabilities – class 49



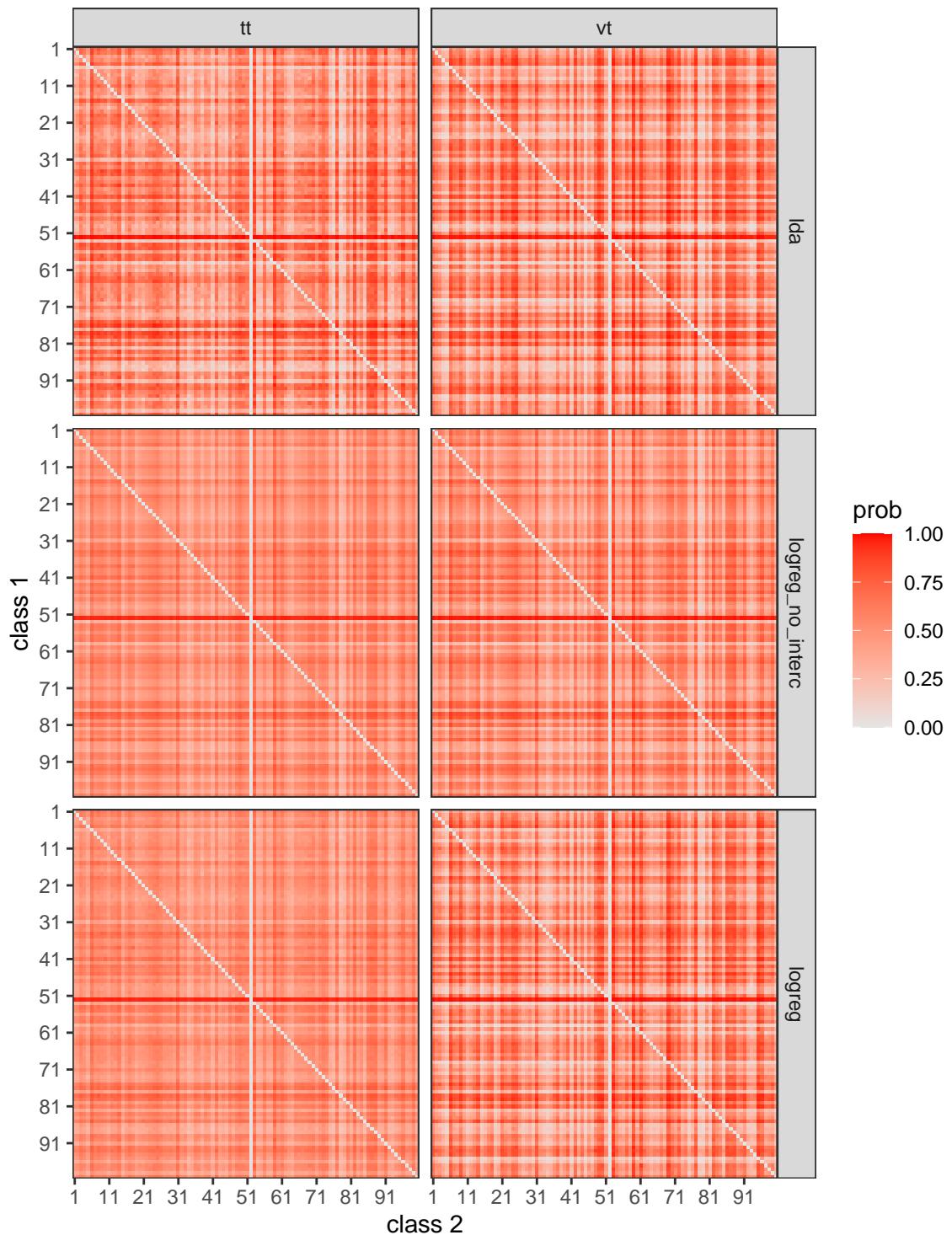
Average pairwise probabilities – class 50



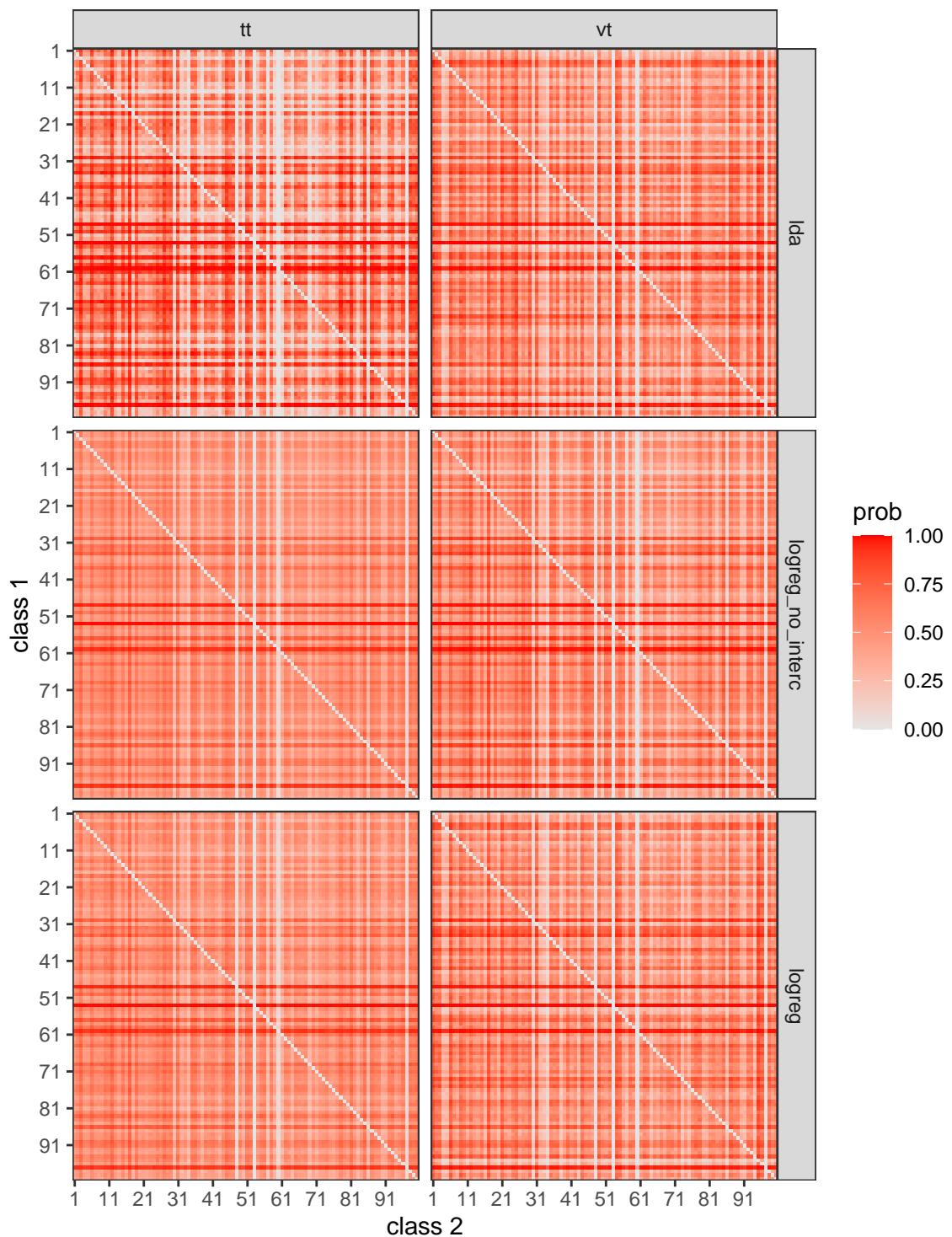
Average pairwise probabilities – class 51



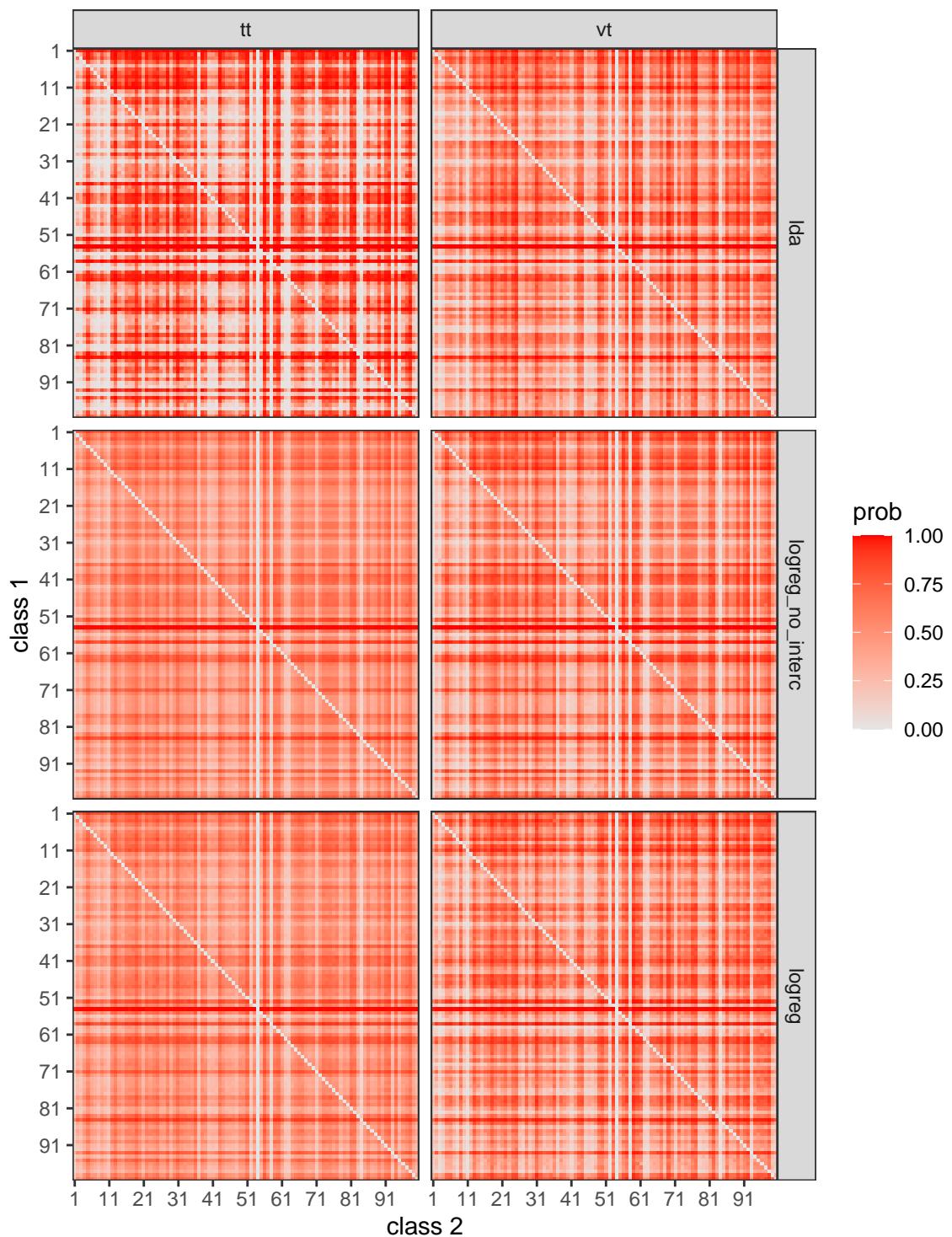
## Average pairwise probabilities – class 52



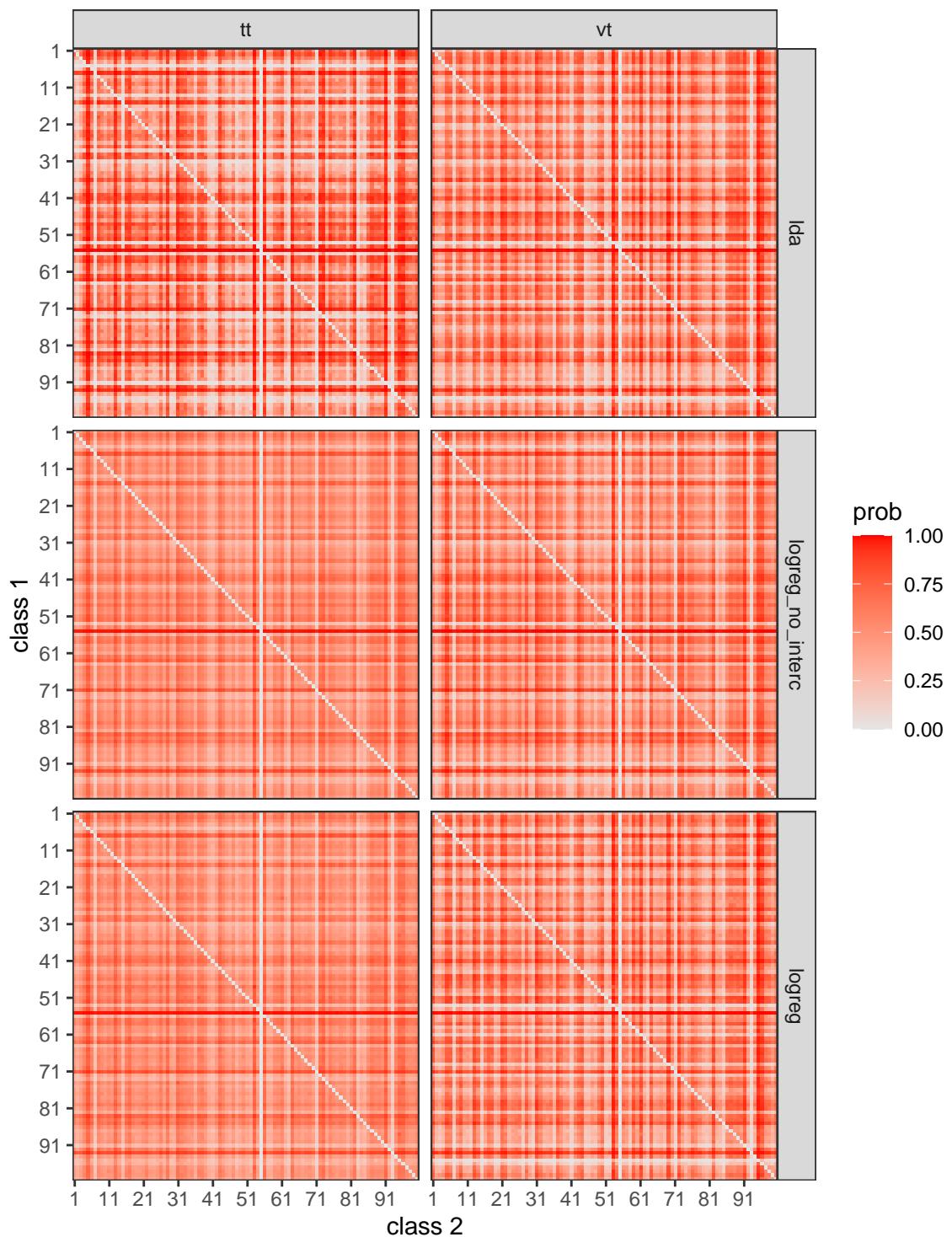
Average pairwise probabilities – class 53



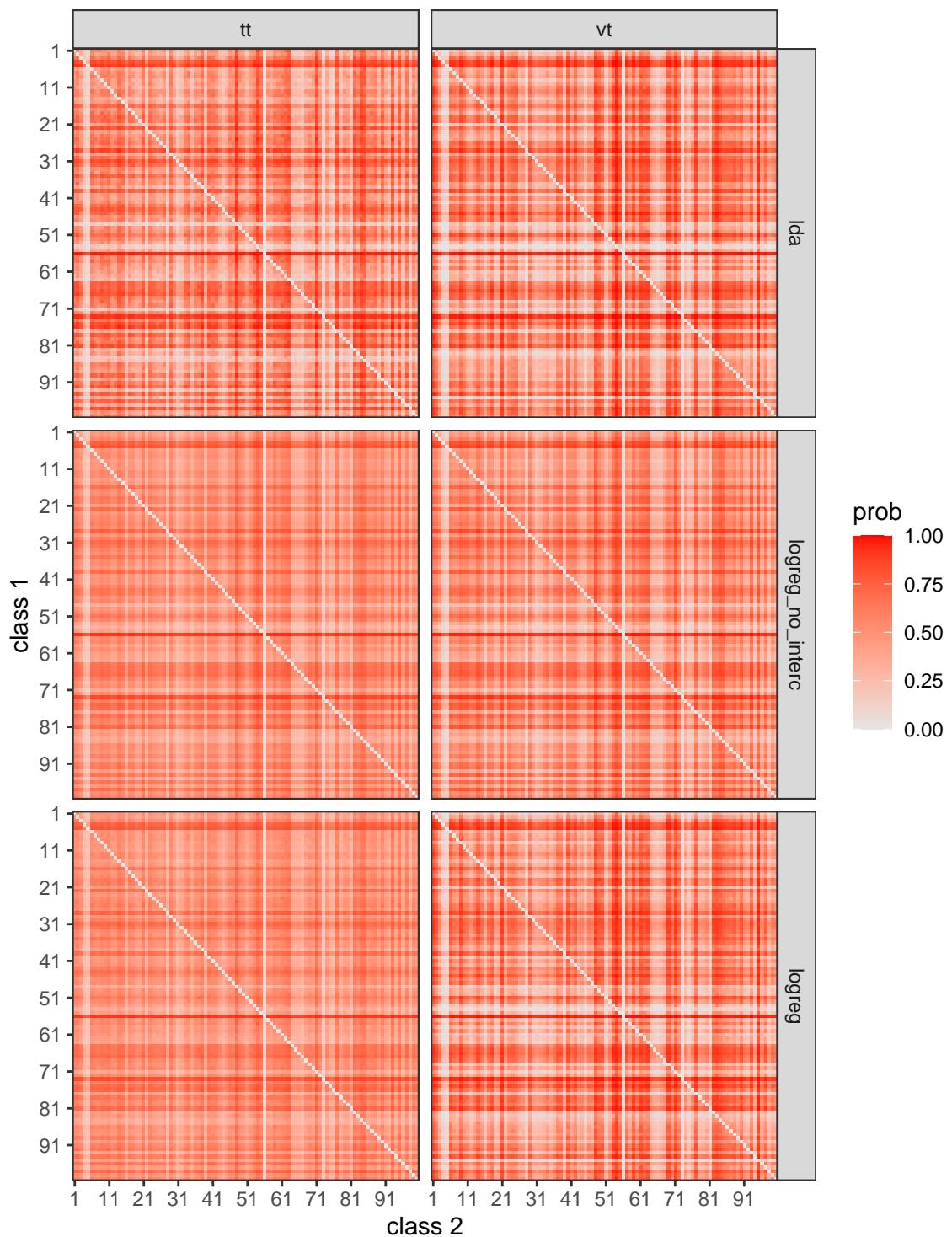
Average pairwise probabilities – class 54



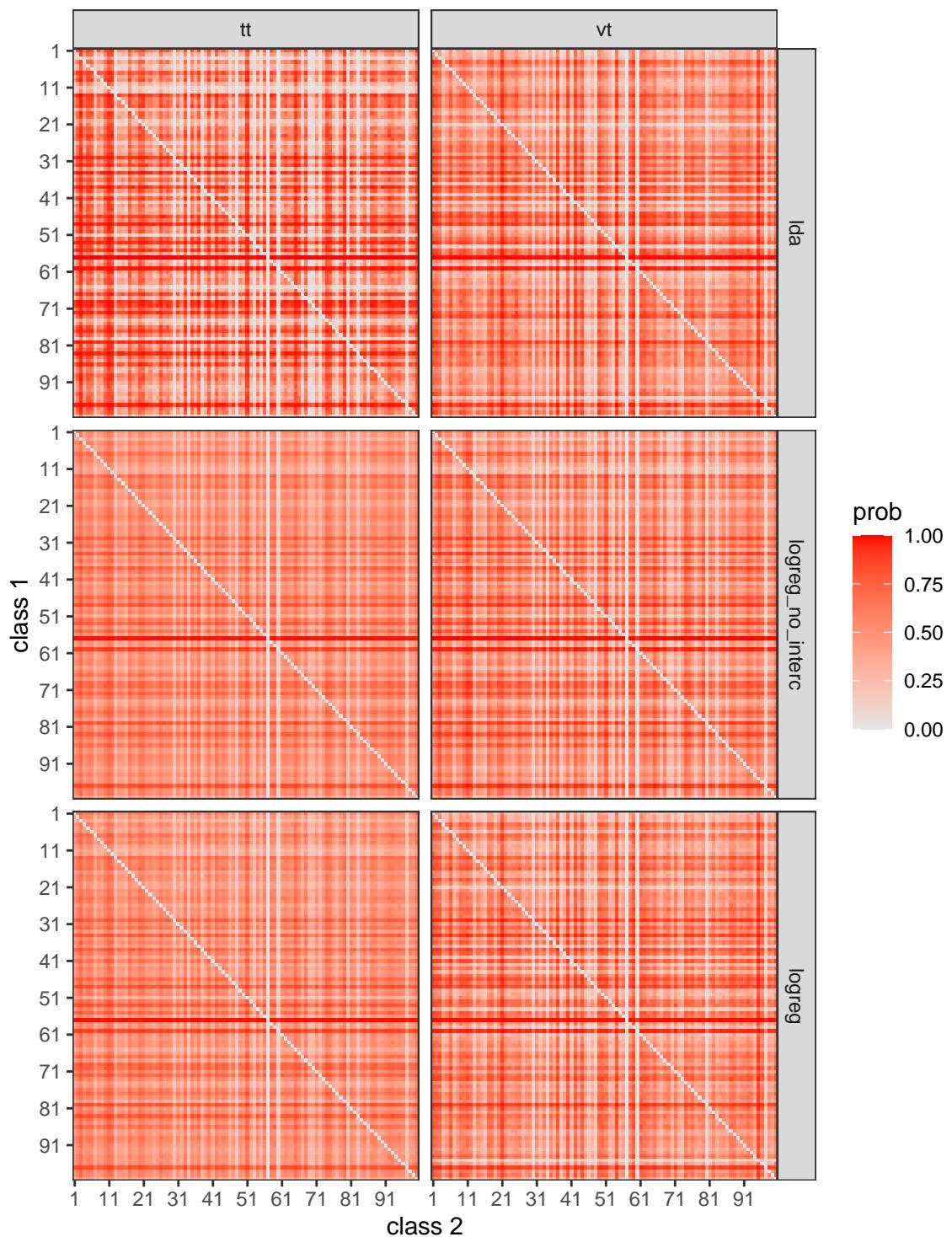
Average pairwise probabilities – class 55



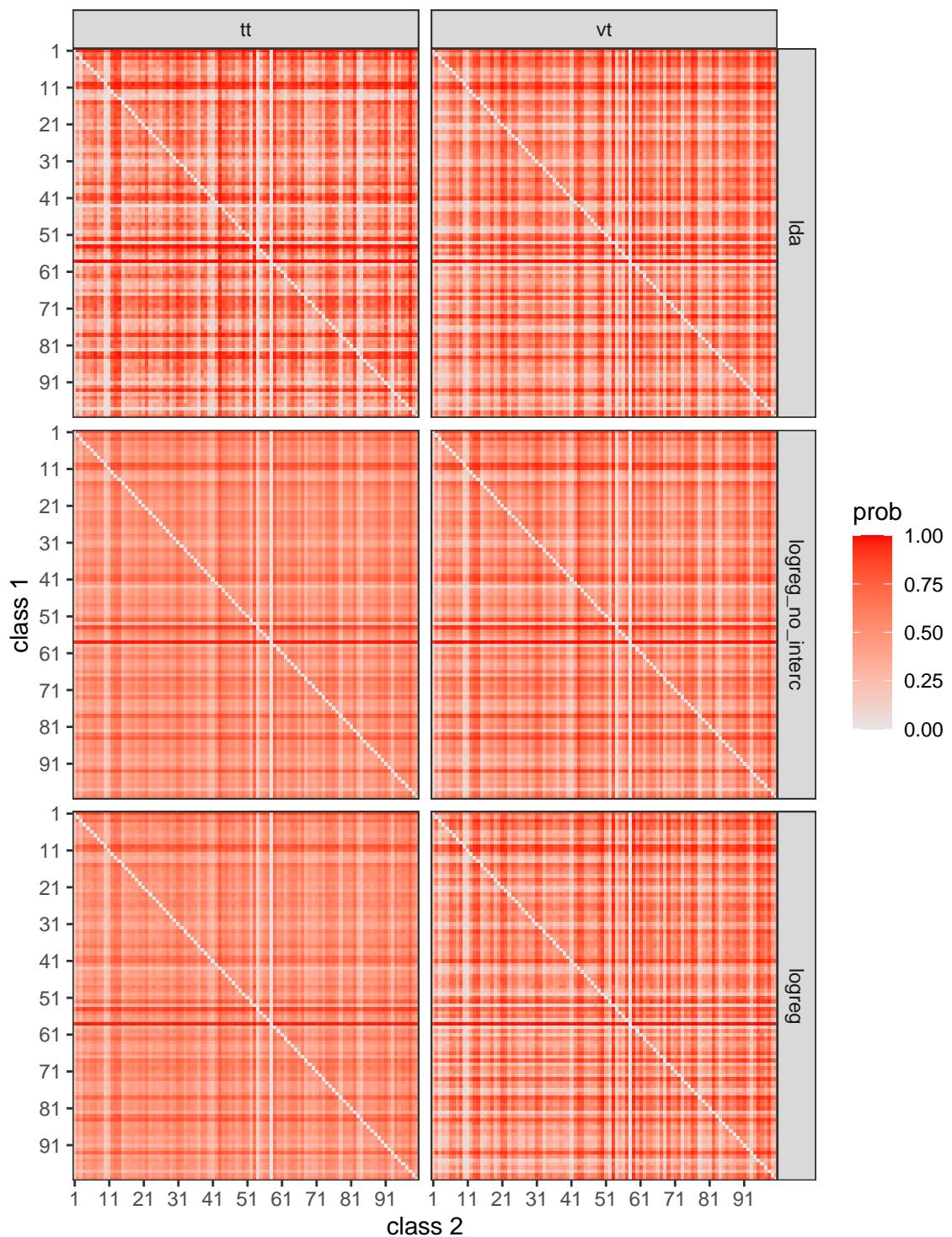
Average pairwise probabilities – class 56



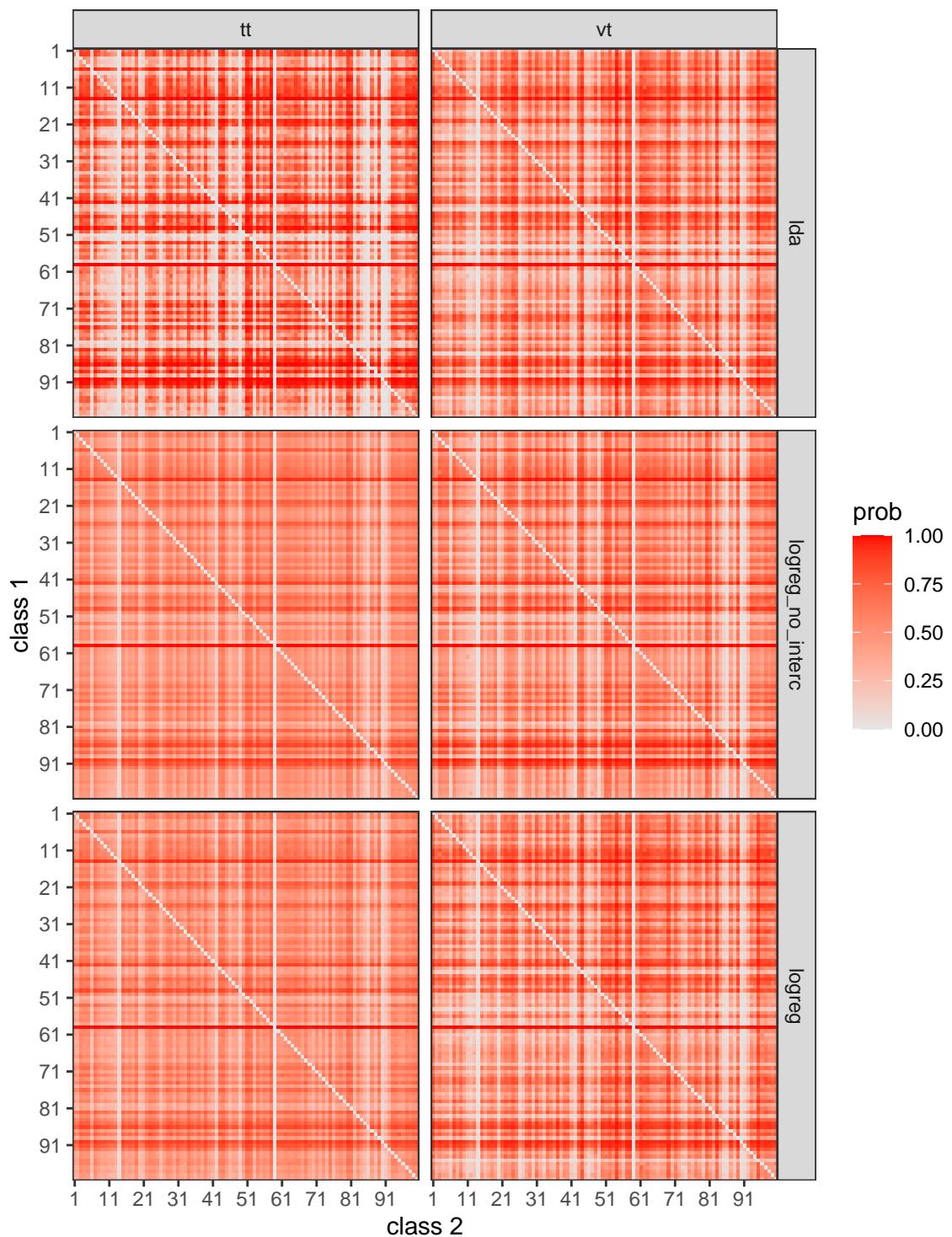
Average pairwise probabilities – class 57



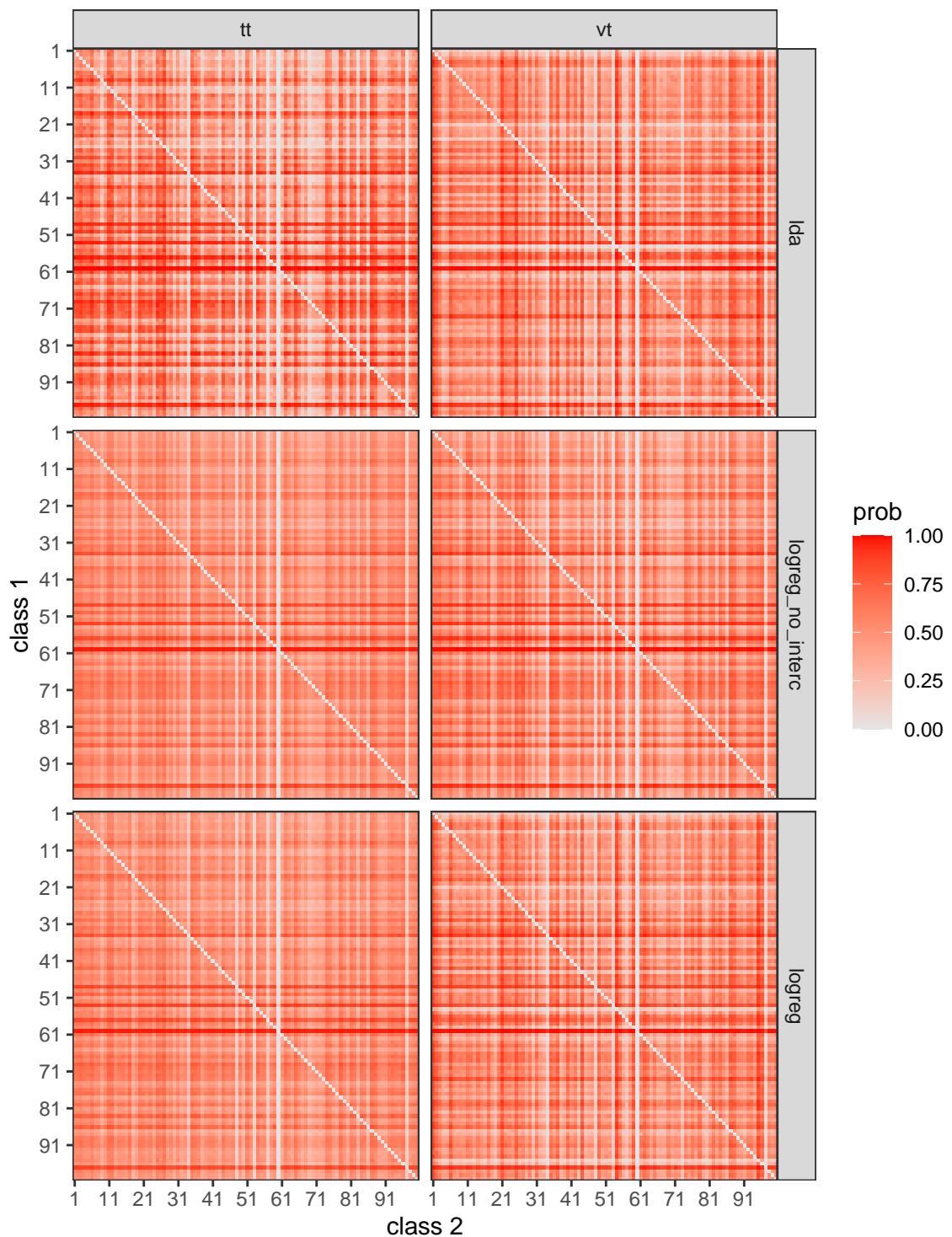
Average pairwise probabilities – class 58



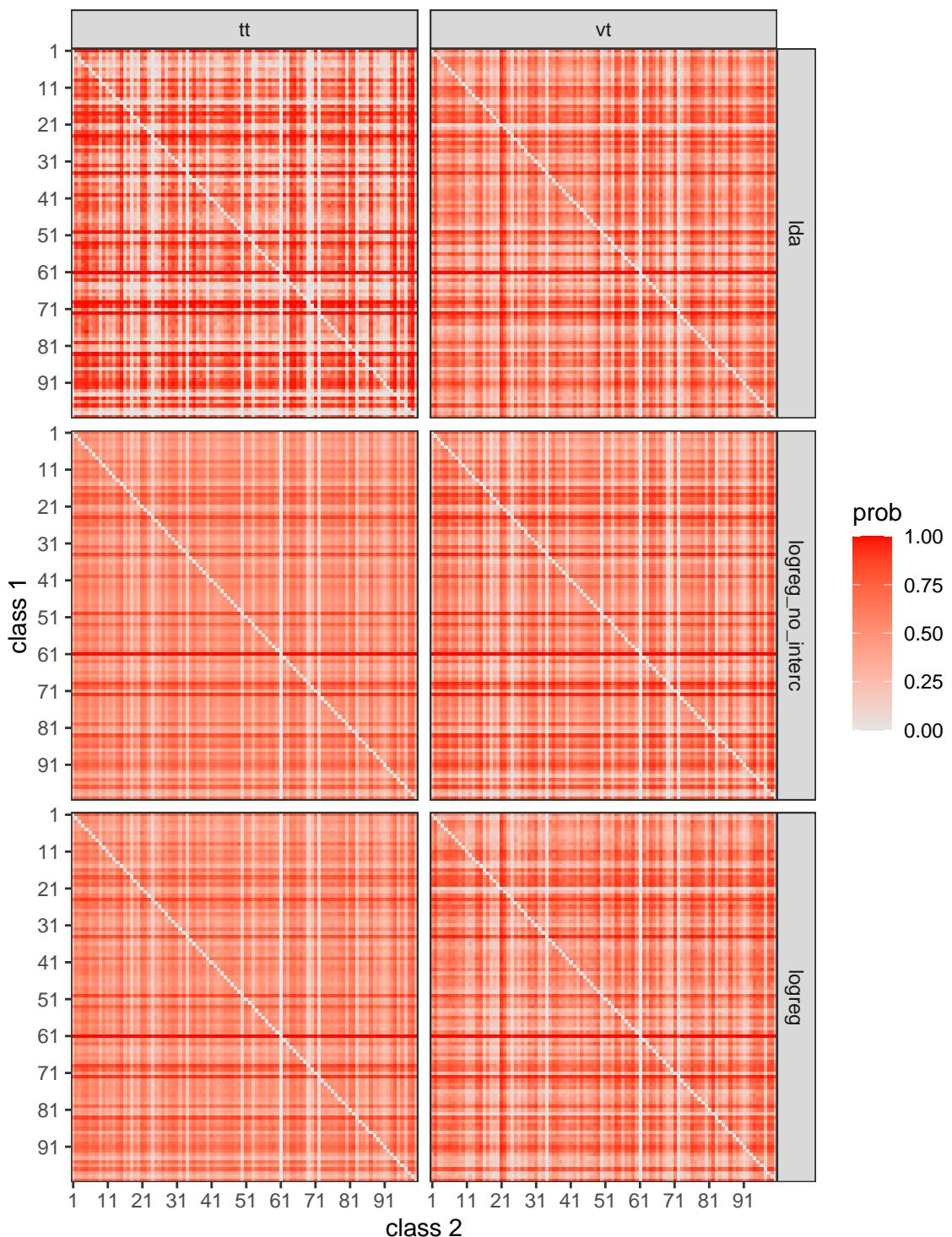
Average pairwise probabilities – class 59



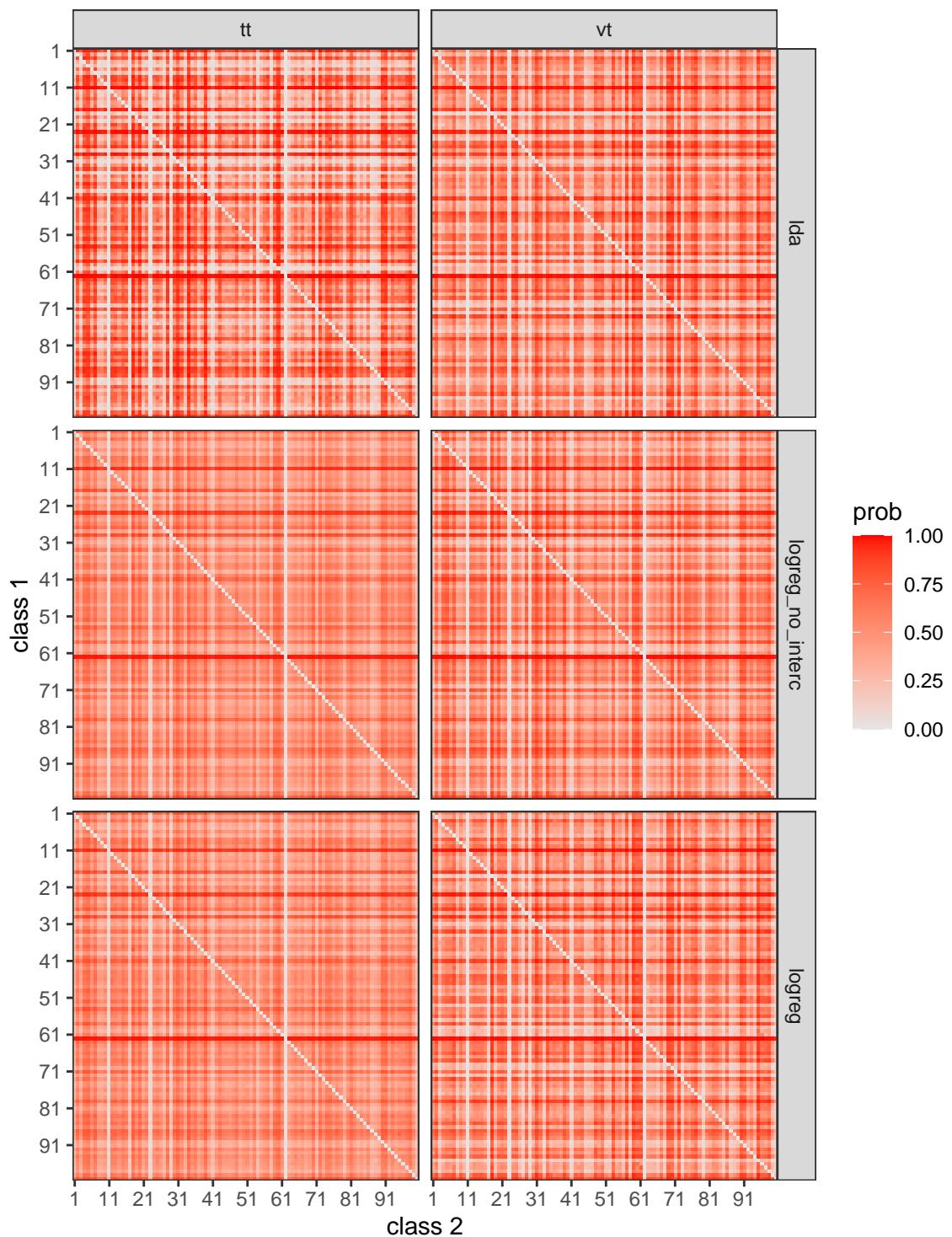
Average pairwise probabilities – class 60



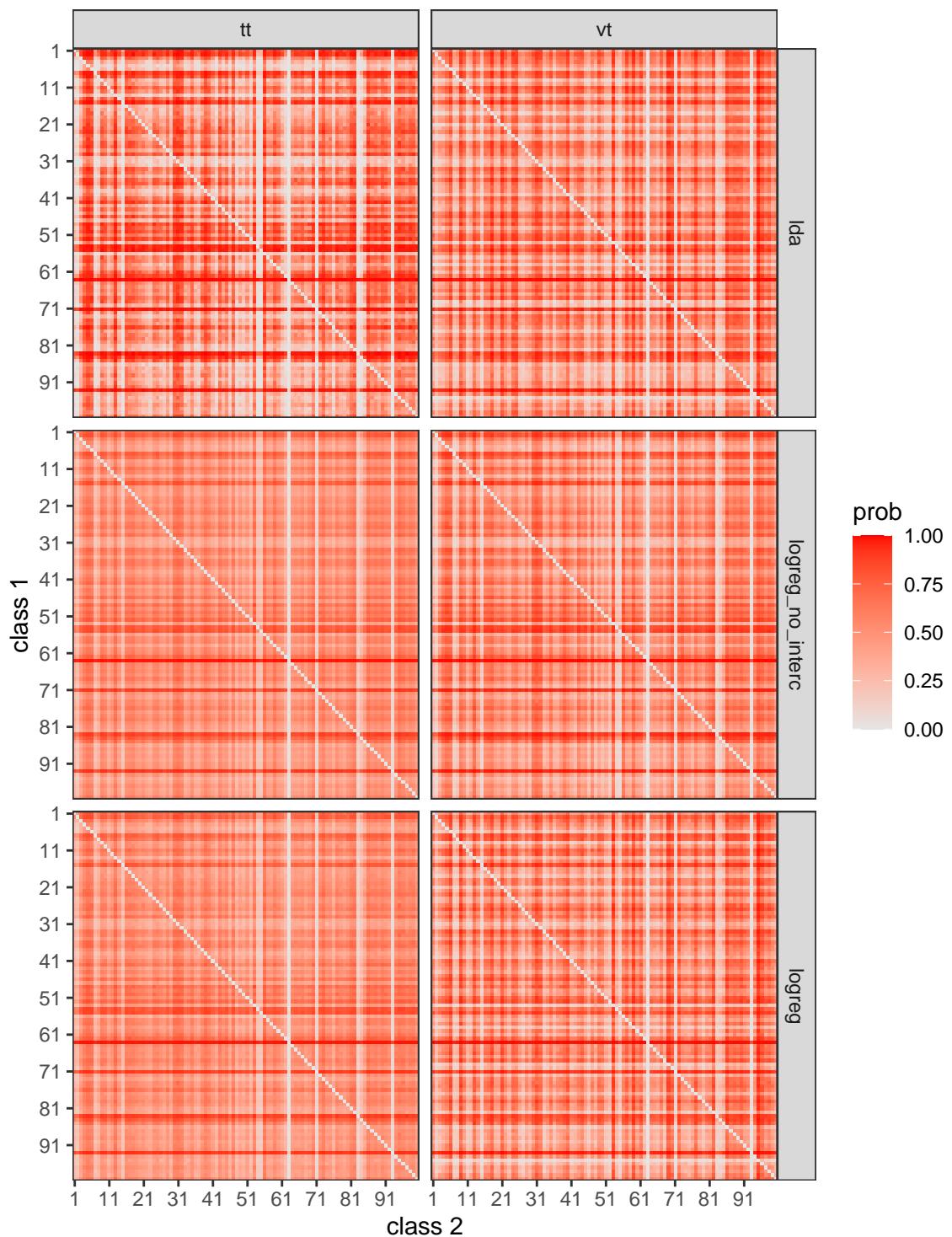
Average pairwise probabilities – class 61



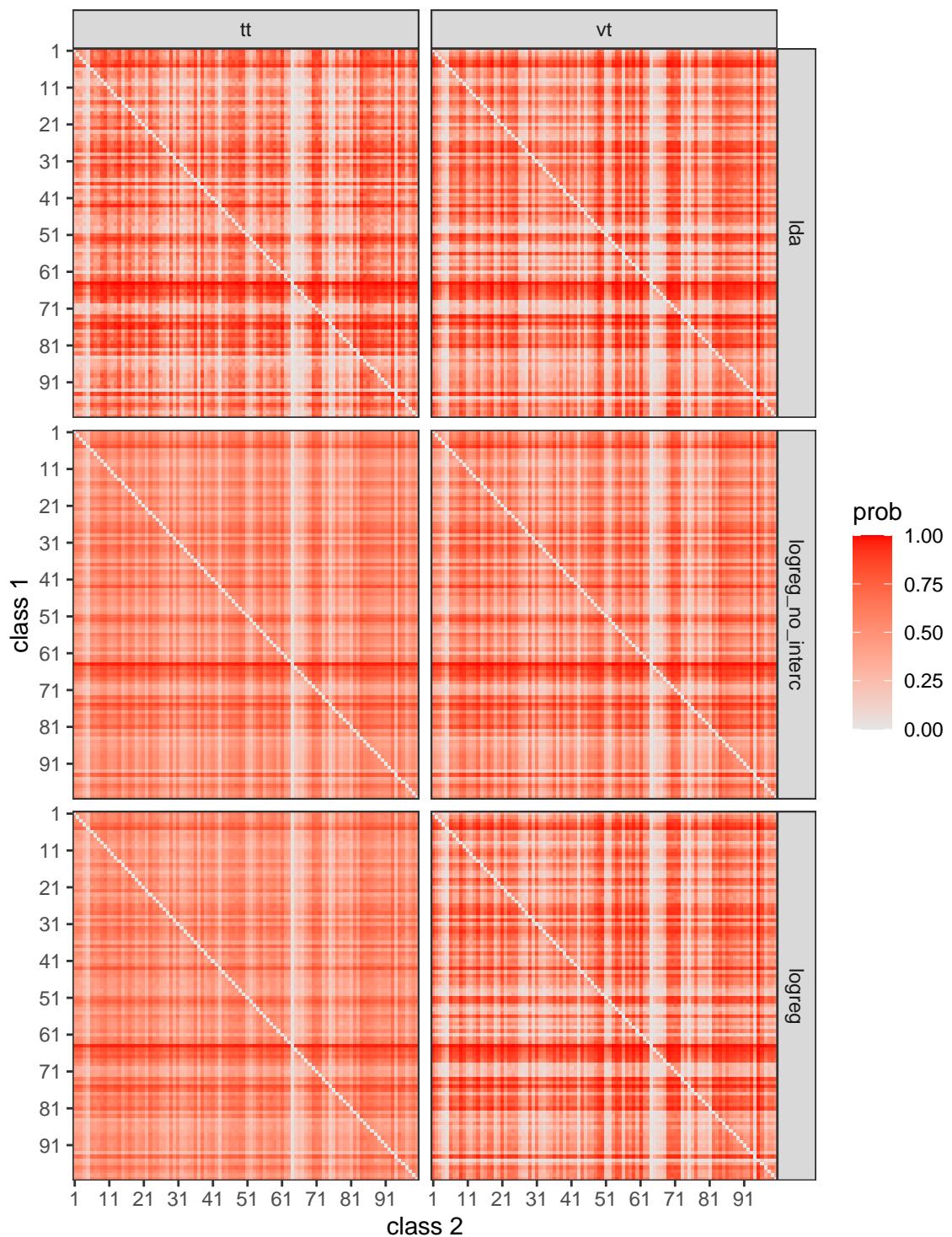
Average pairwise probabilities – class 62



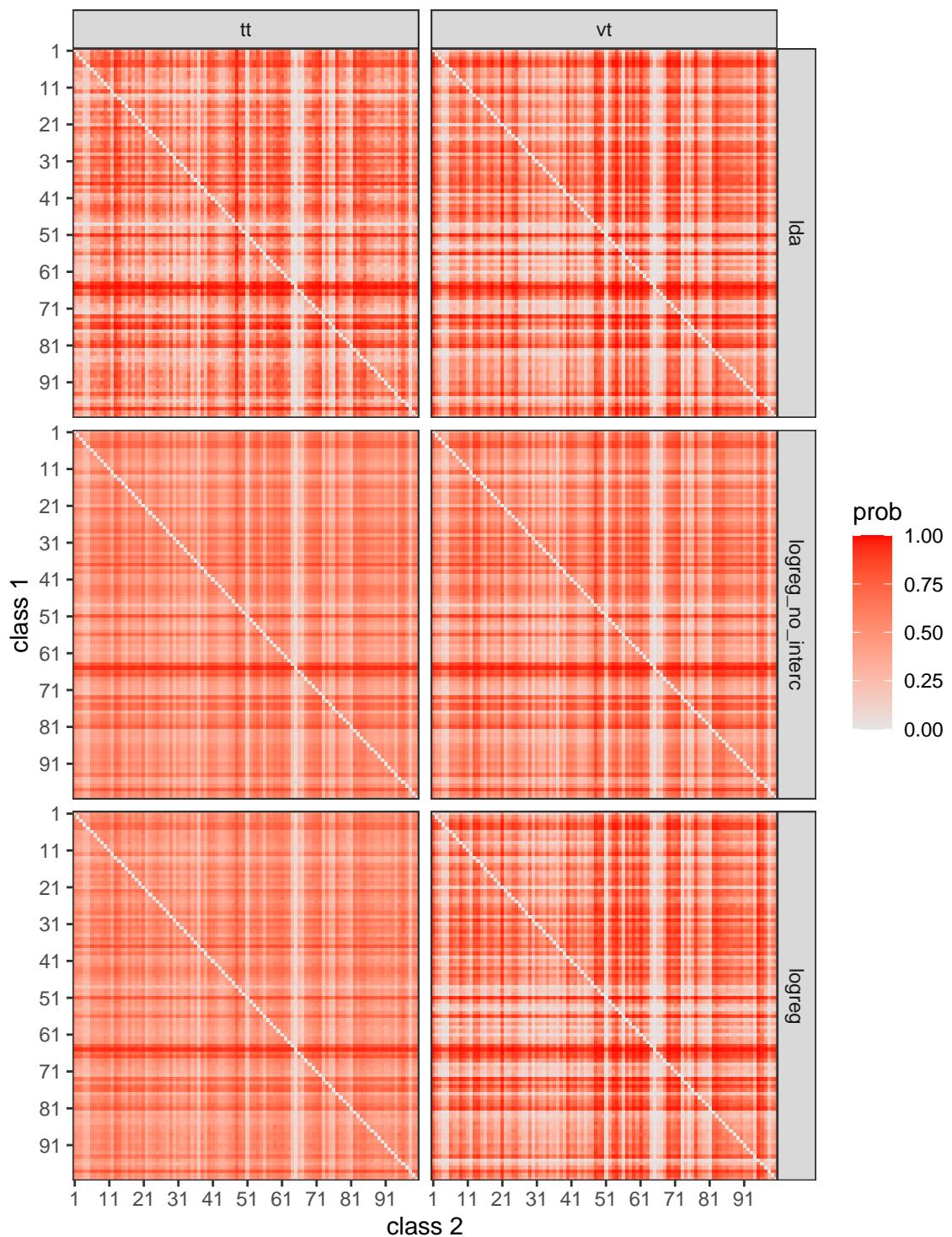
Average pairwise probabilities – class 63



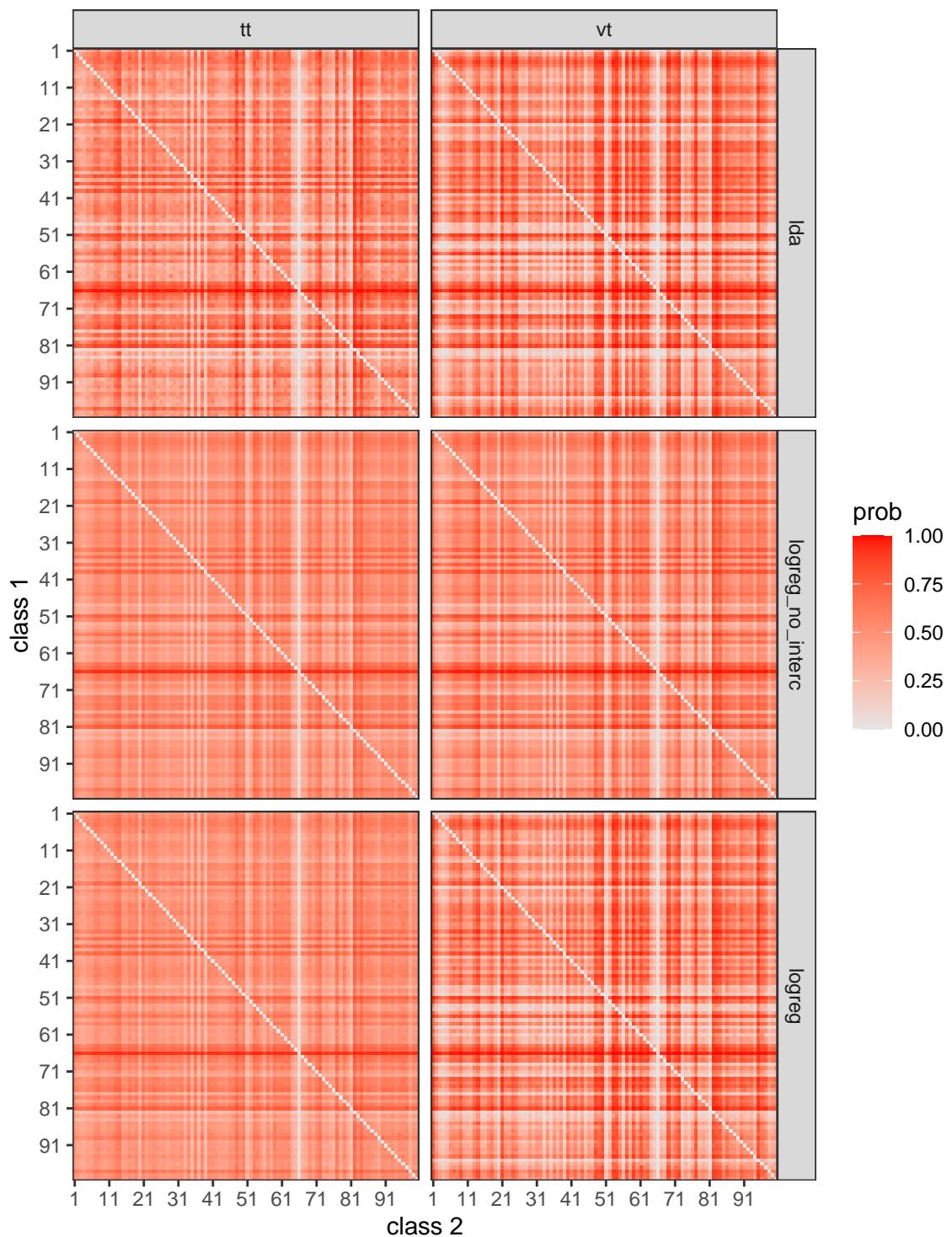
Average pairwise probabilities – class 64



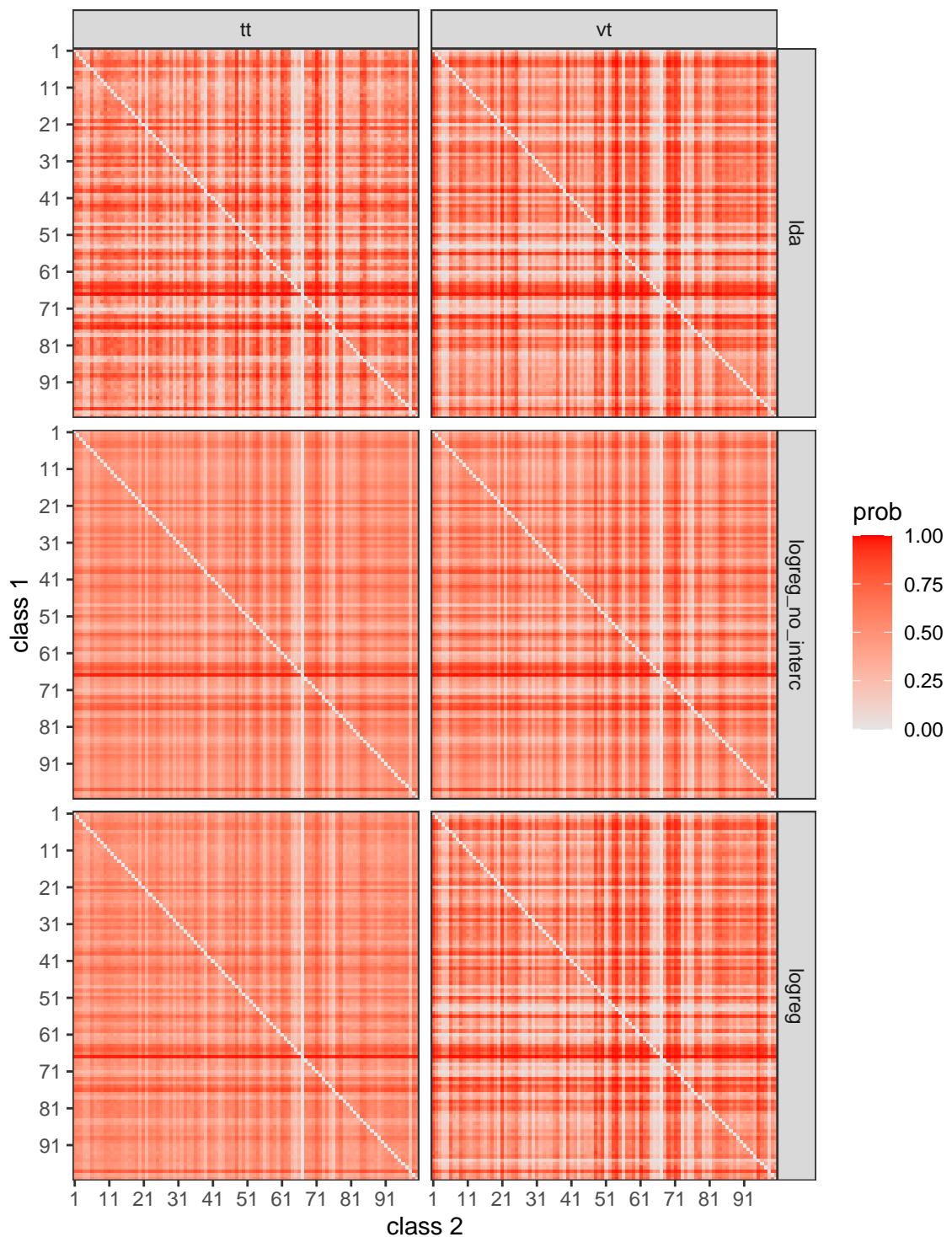
Average pairwise probabilities – class 65



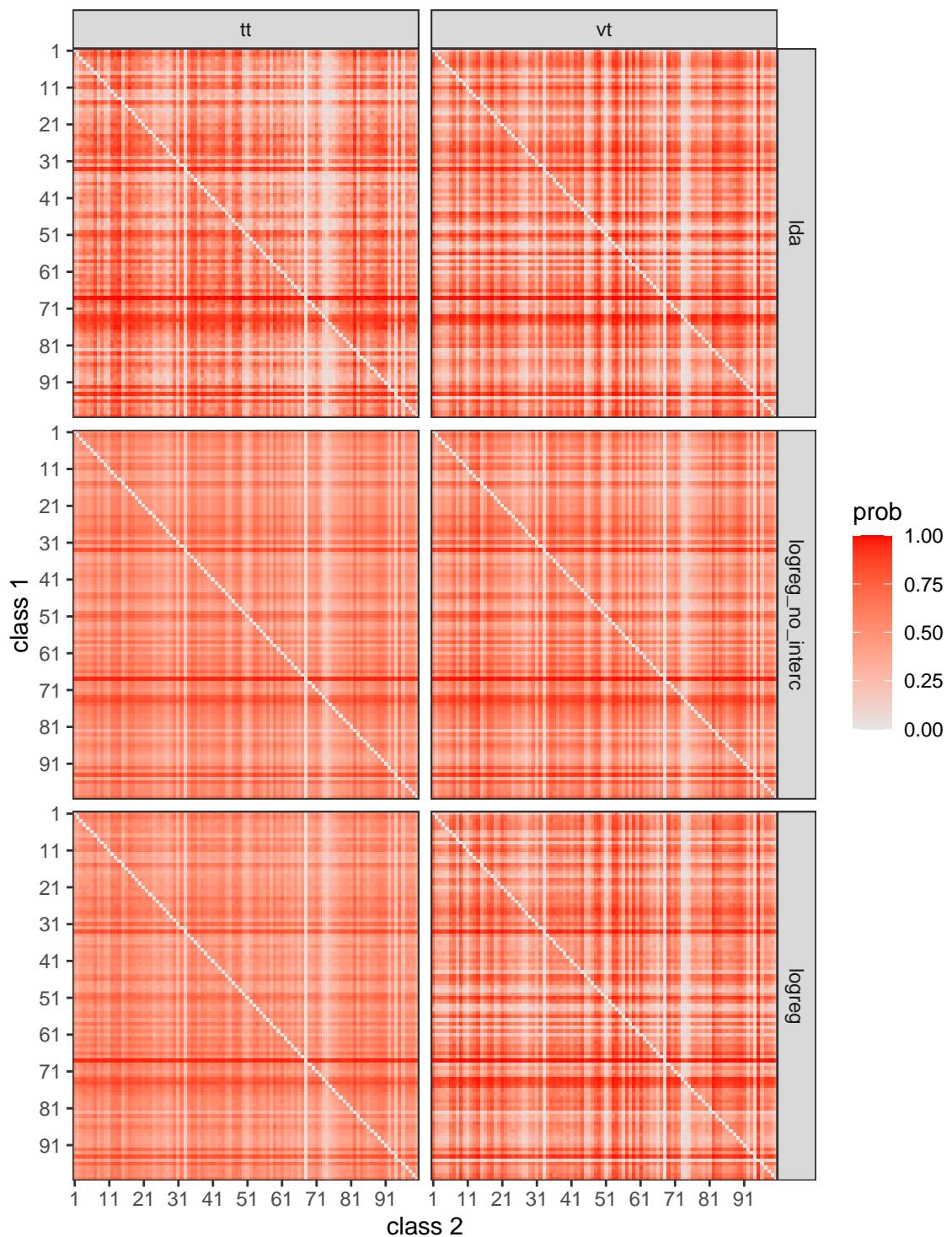
Average pairwise probabilities – class 66



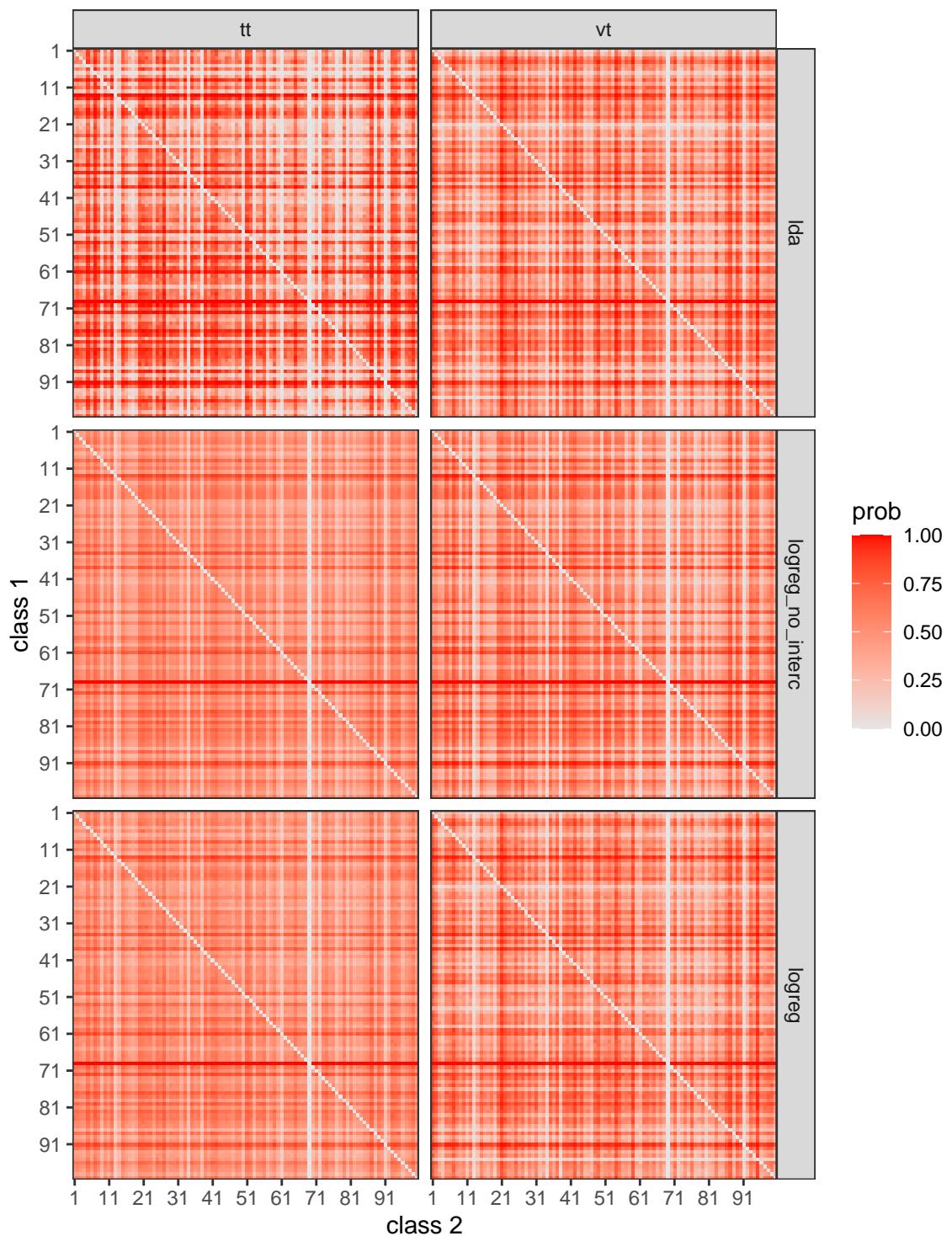
Average pairwise probabilities – class 67



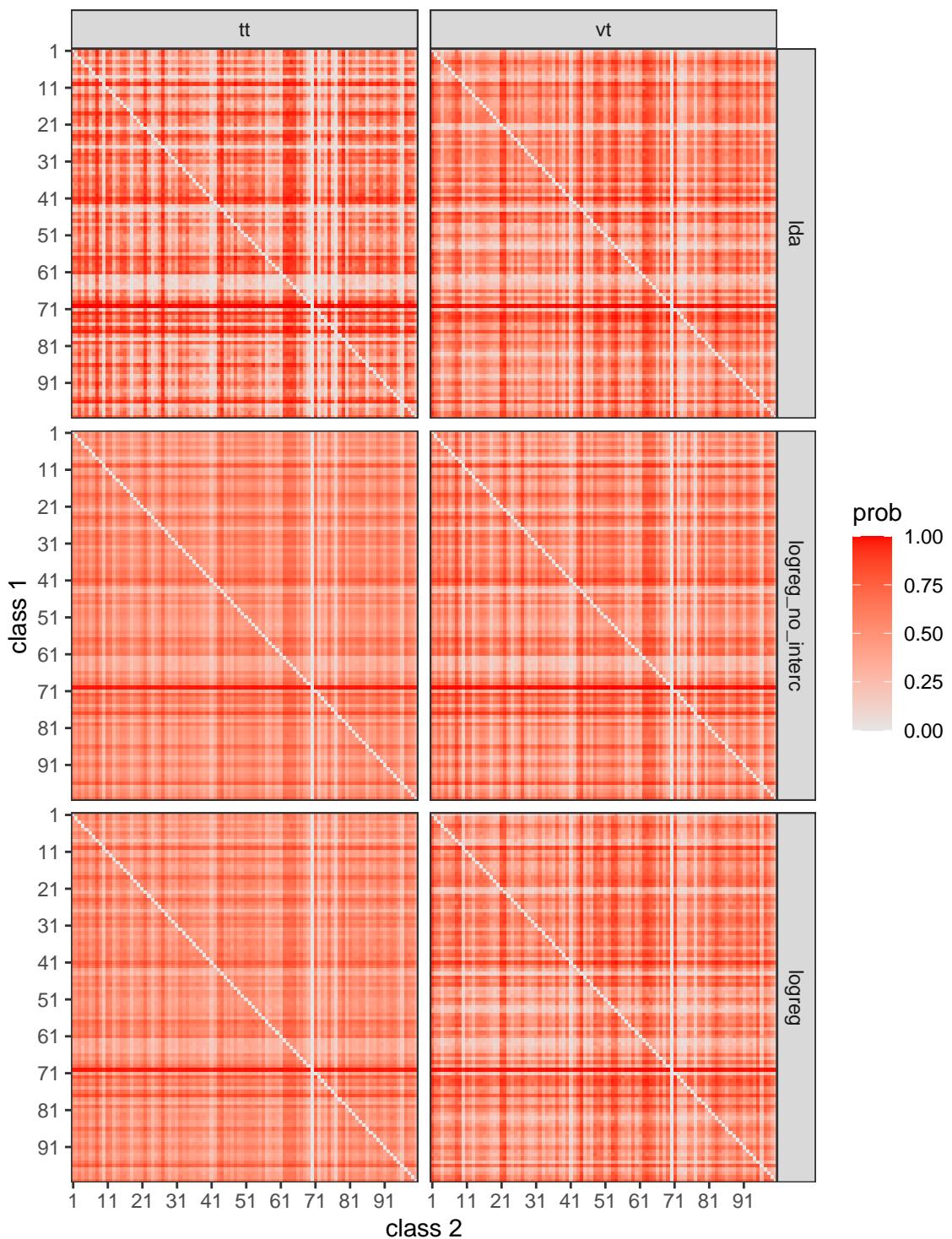
Average pairwise probabilities – class 68



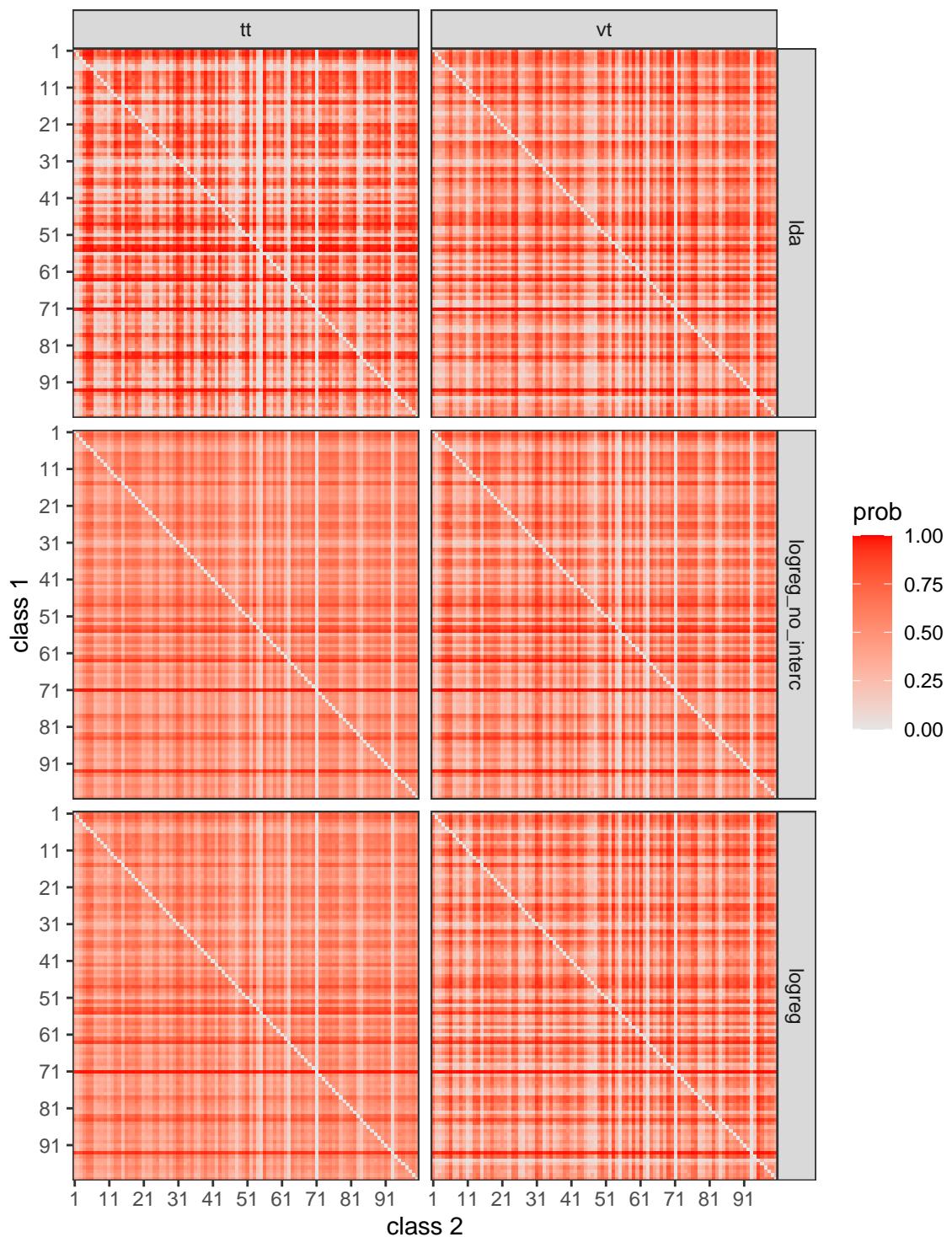
Average pairwise probabilities – class 69



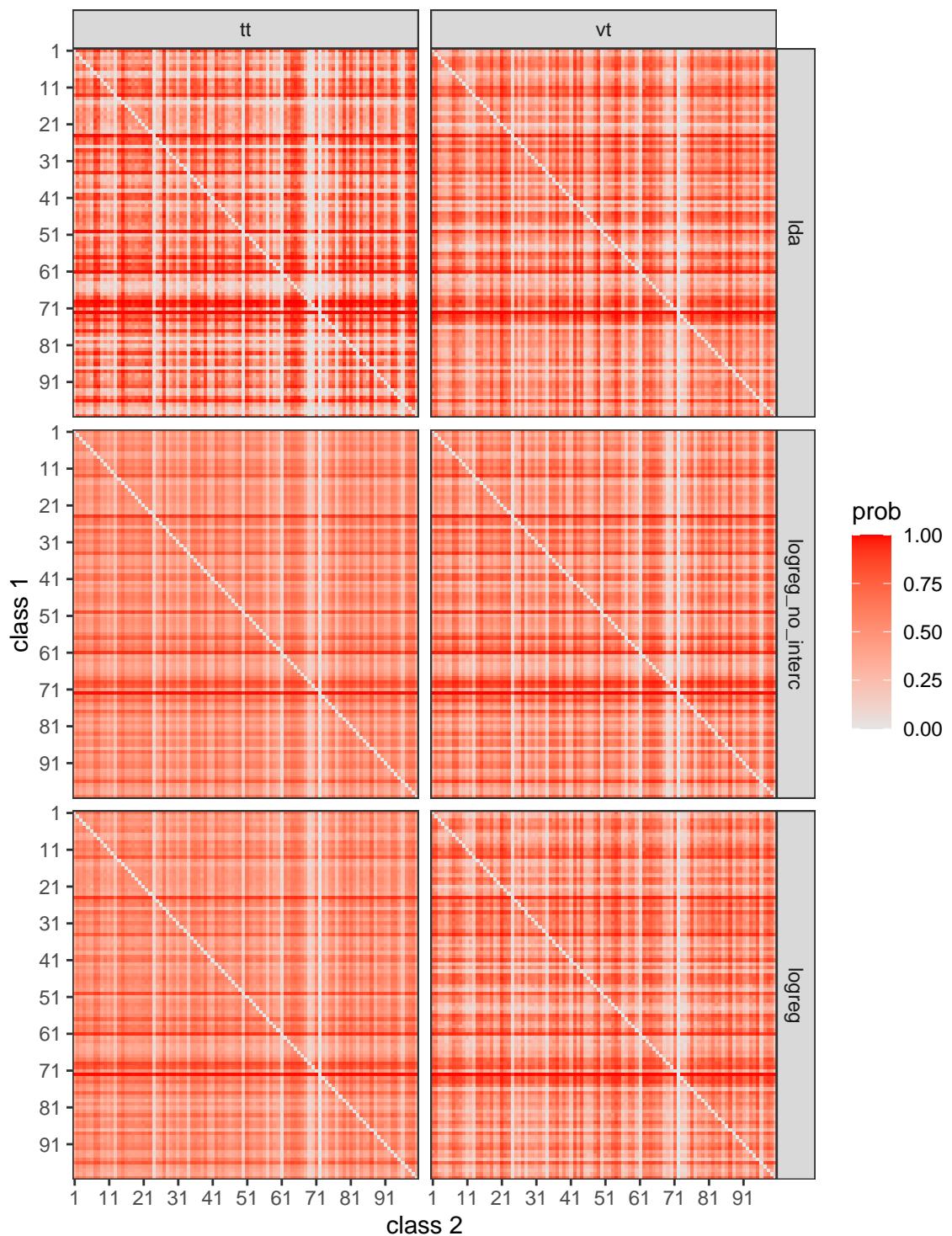
Average pairwise probabilities – class 70



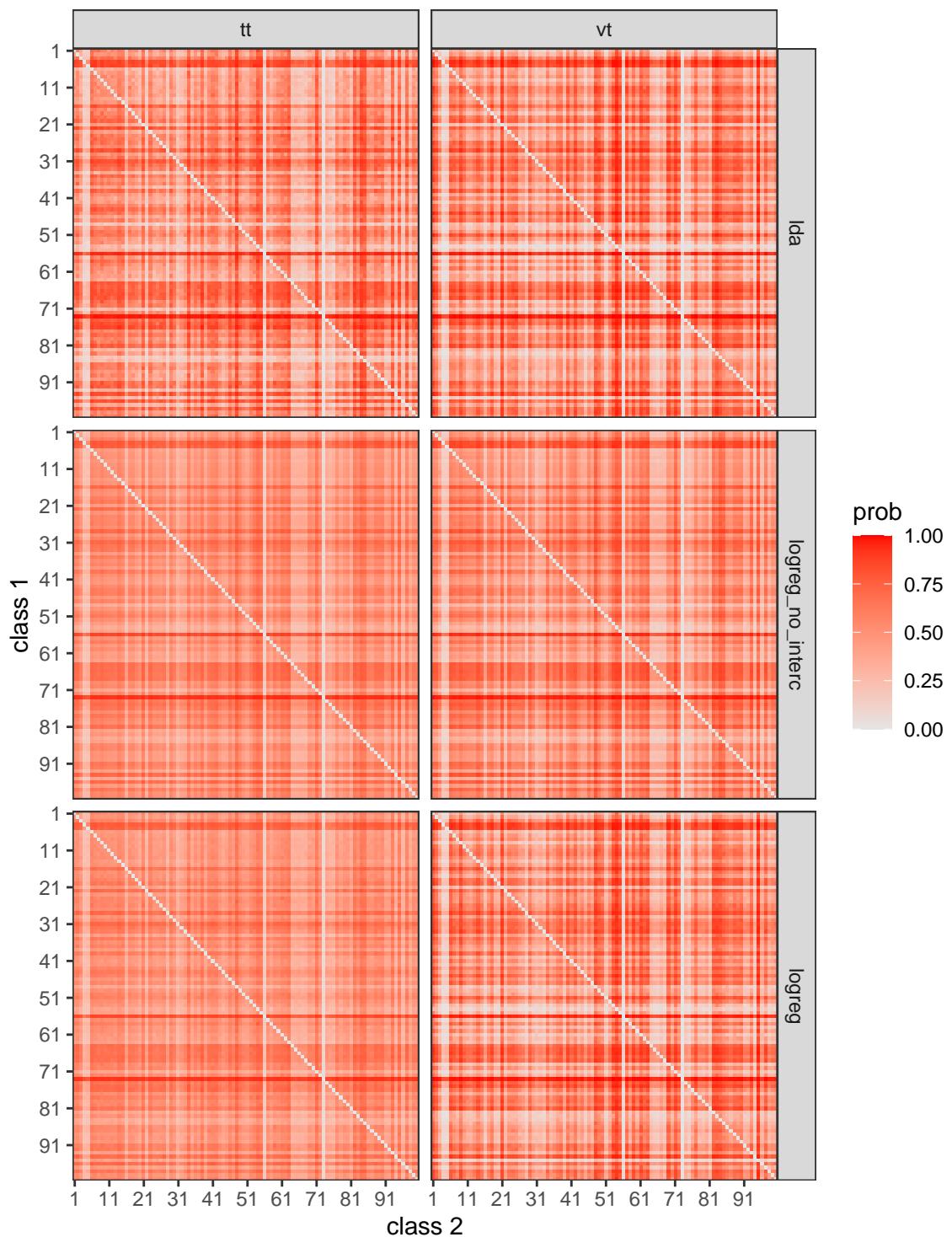
Average pairwise probabilities – class 71



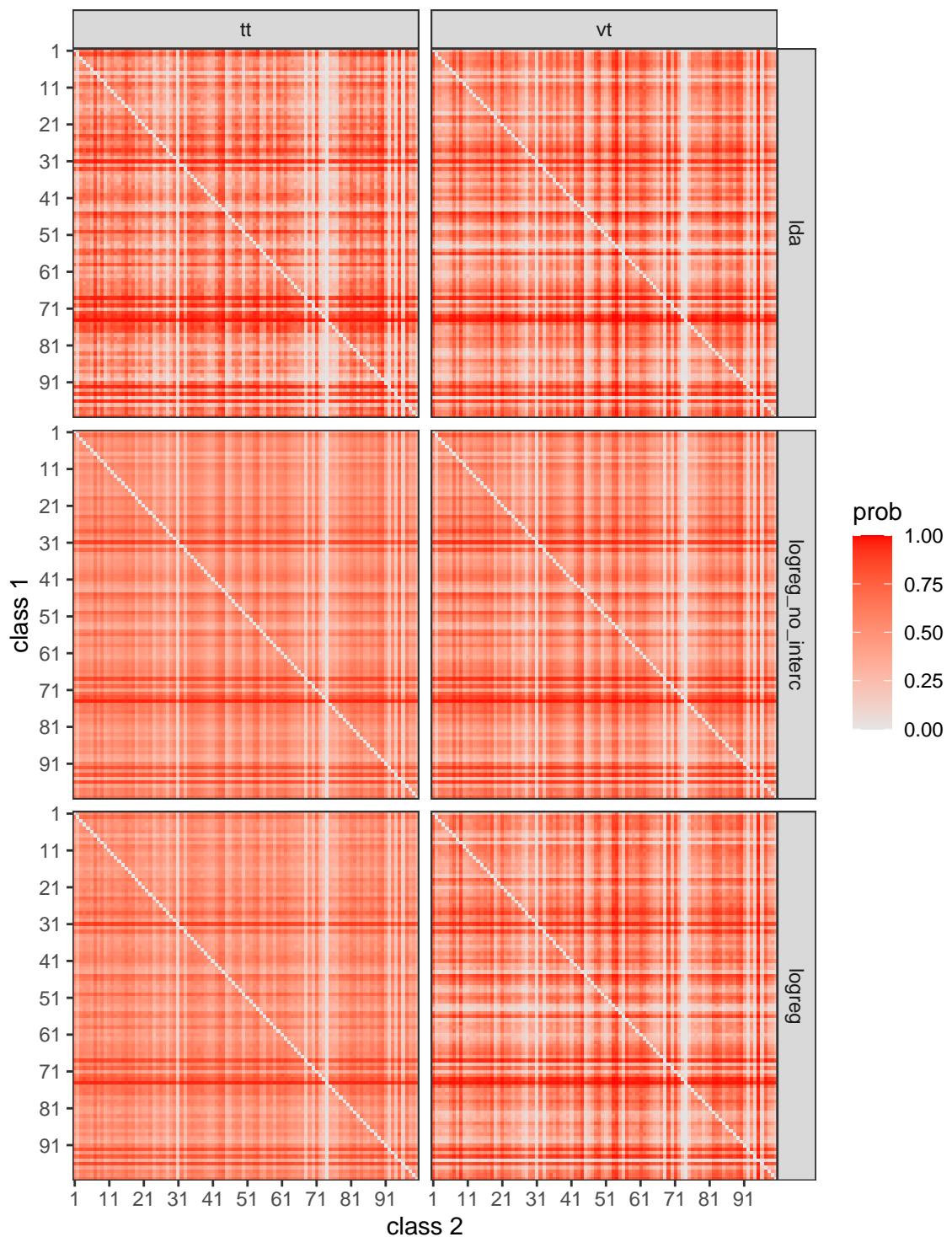
Average pairwise probabilities – class 72



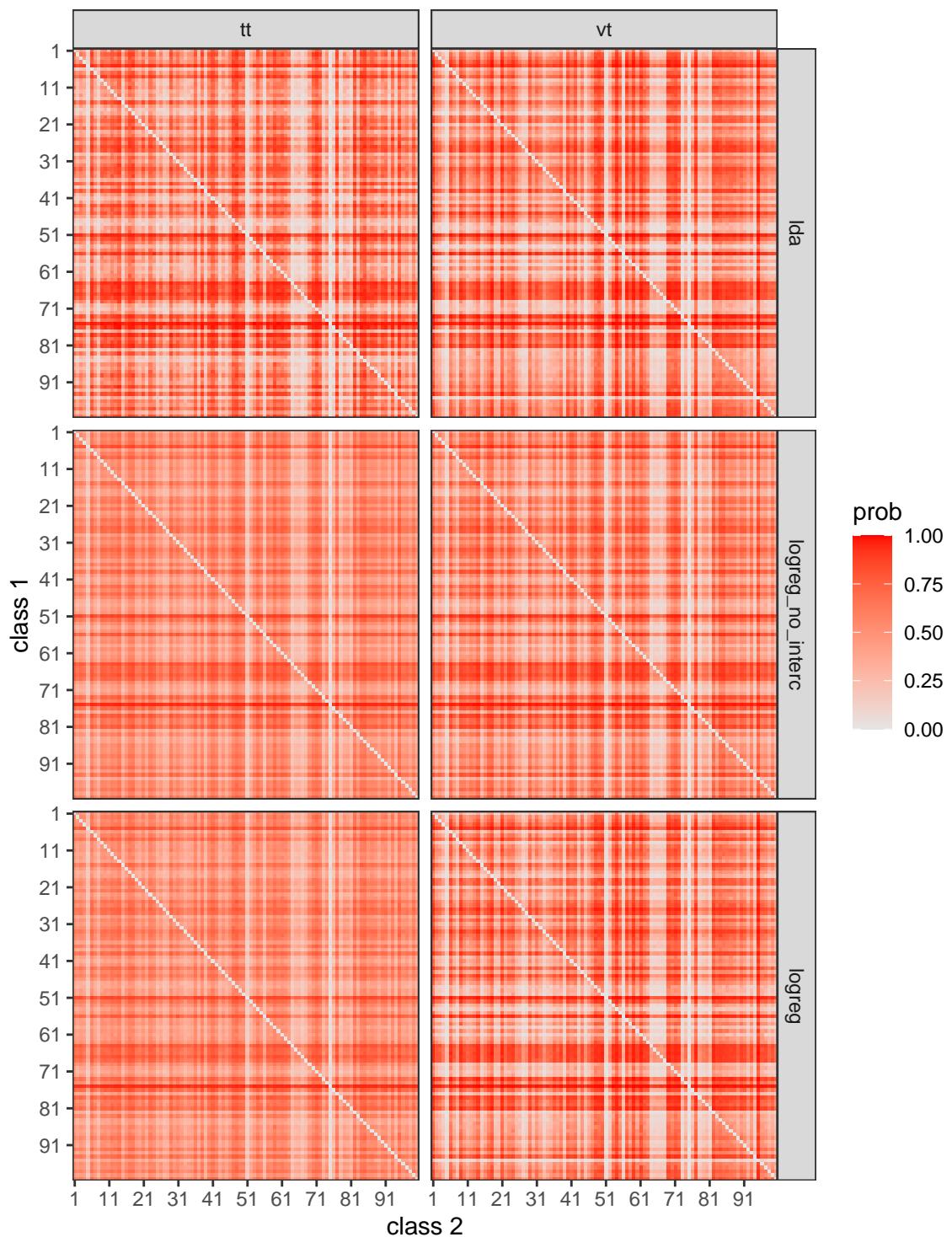
Average pairwise probabilities – class 73



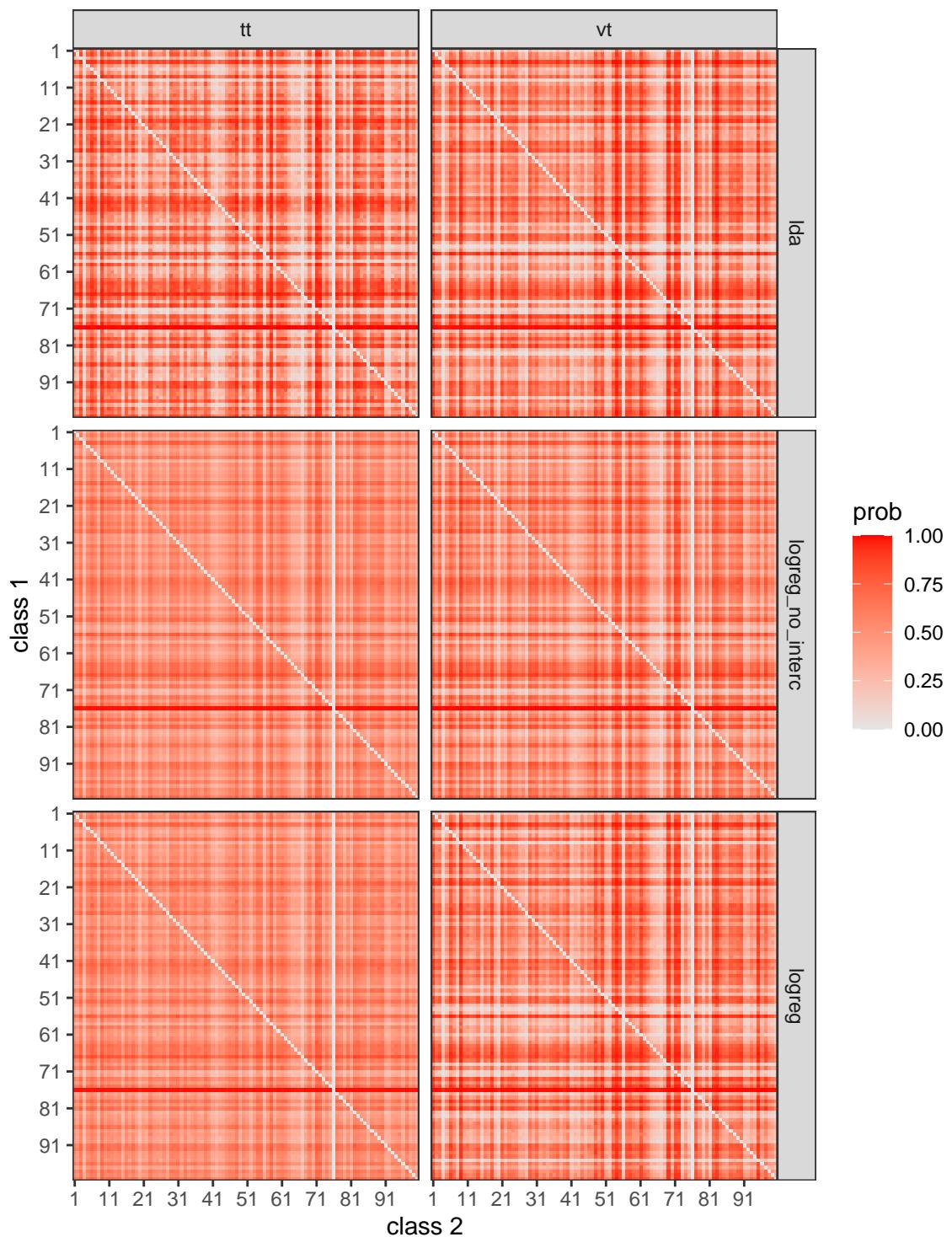
Average pairwise probabilities – class 74



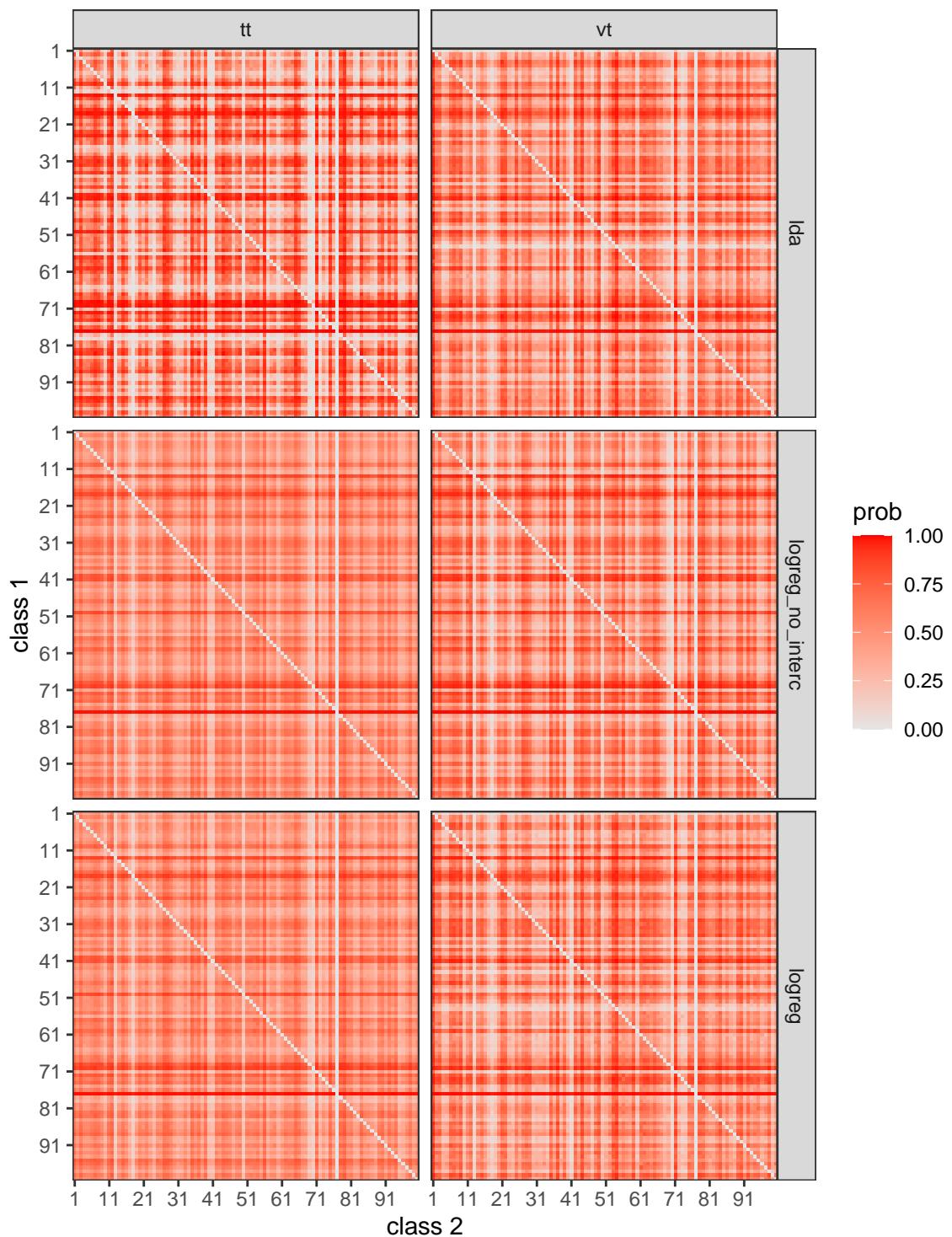
Average pairwise probabilities – class 75



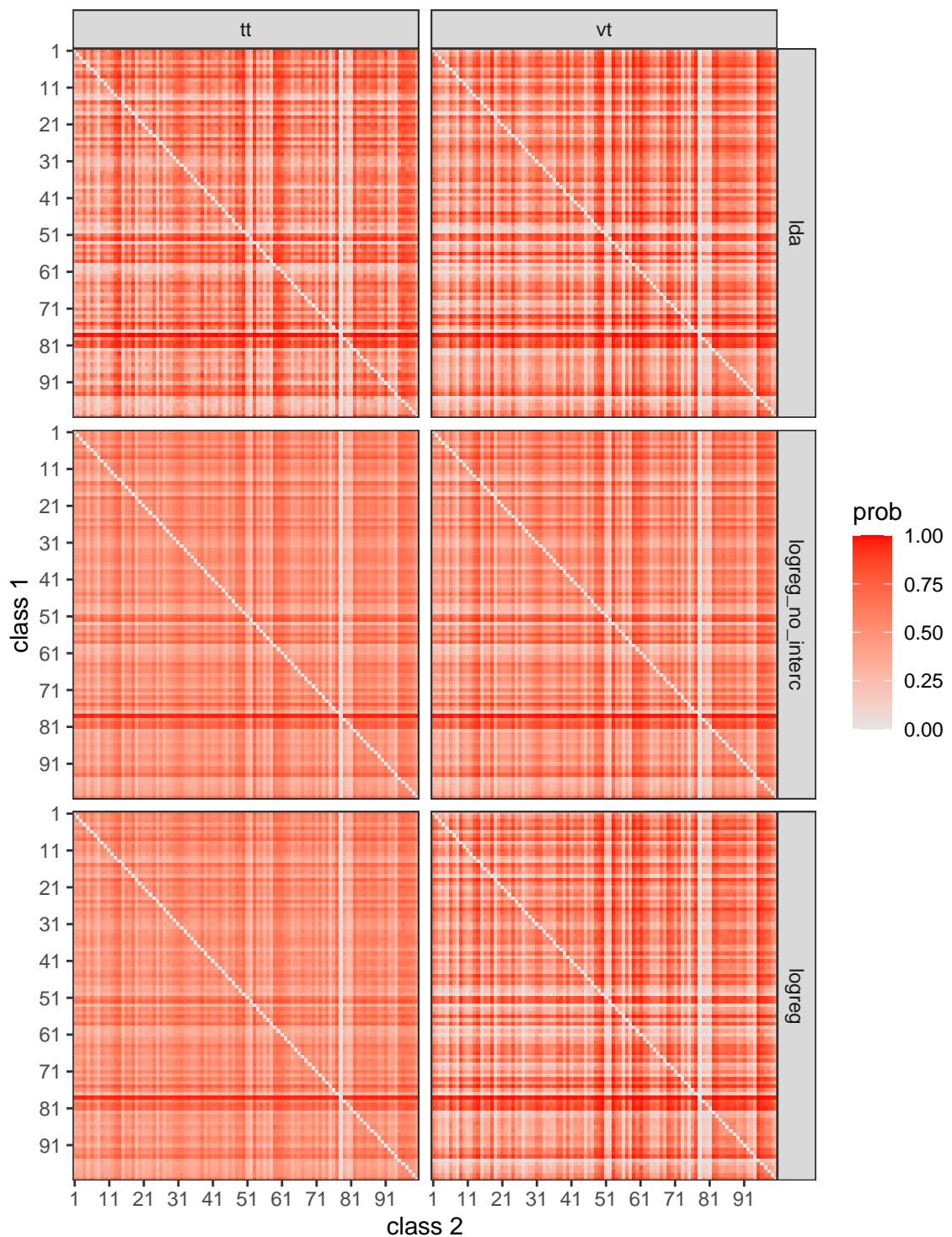
Average pairwise probabilities – class 76



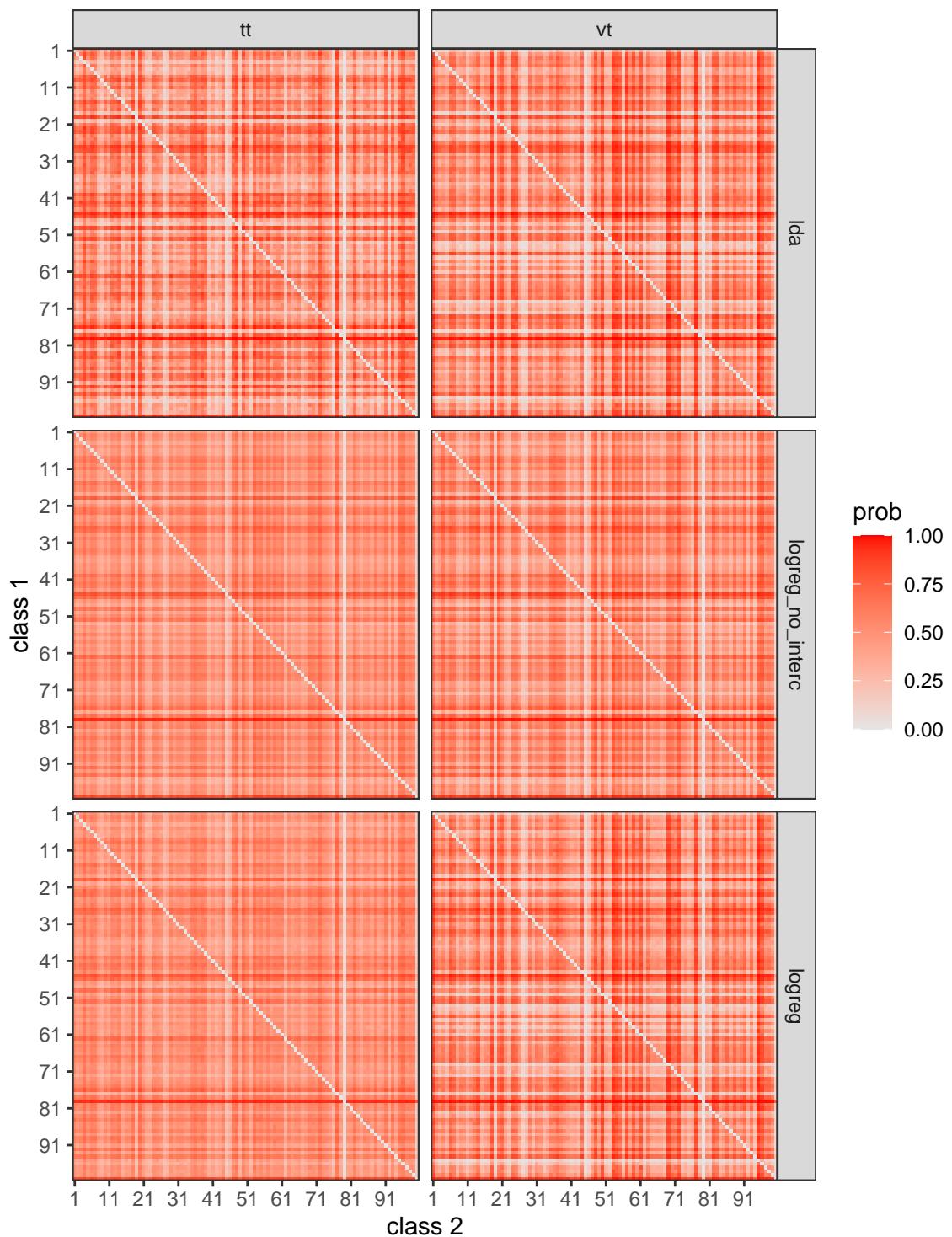
Average pairwise probabilities – class 77



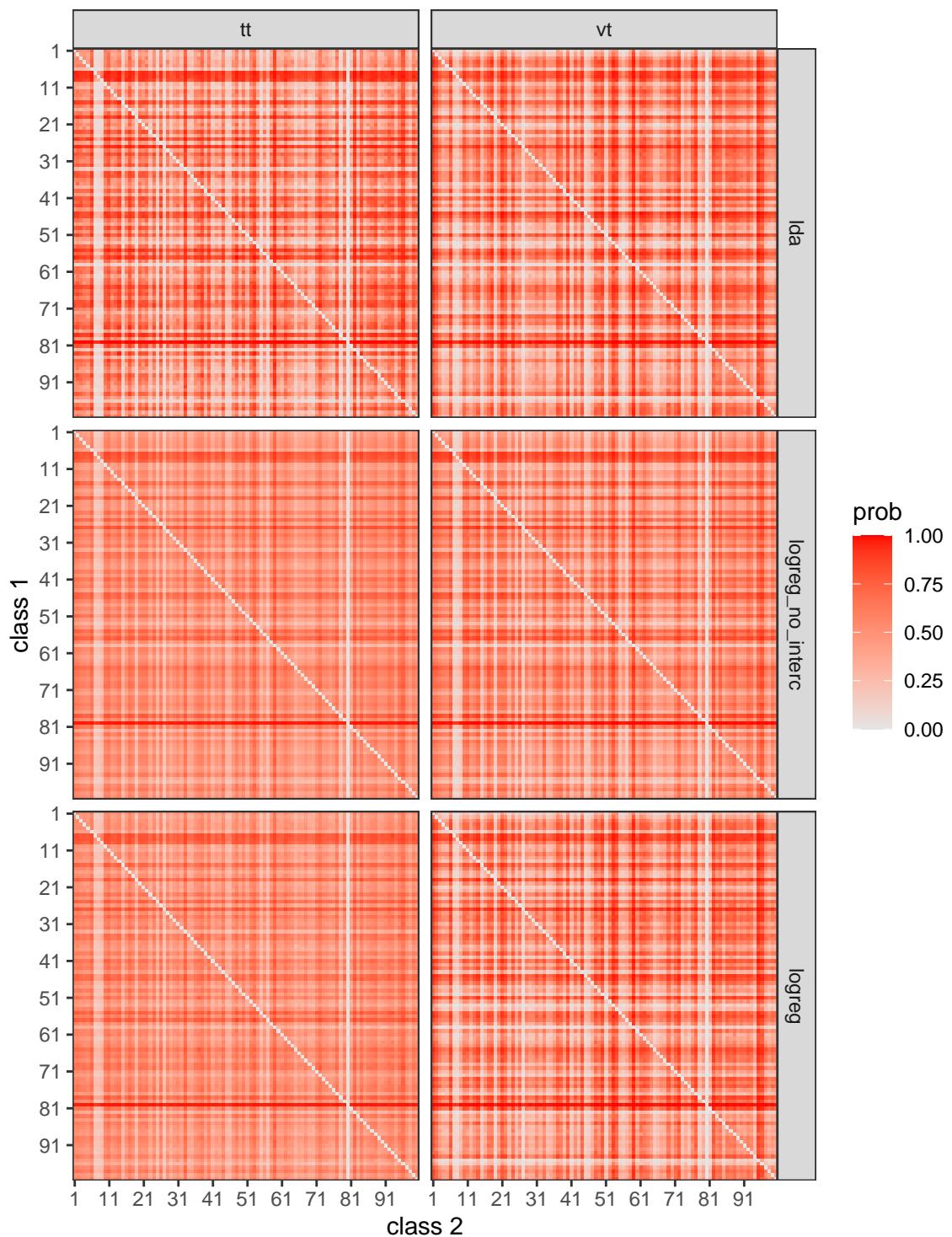
Average pairwise probabilities – class 78



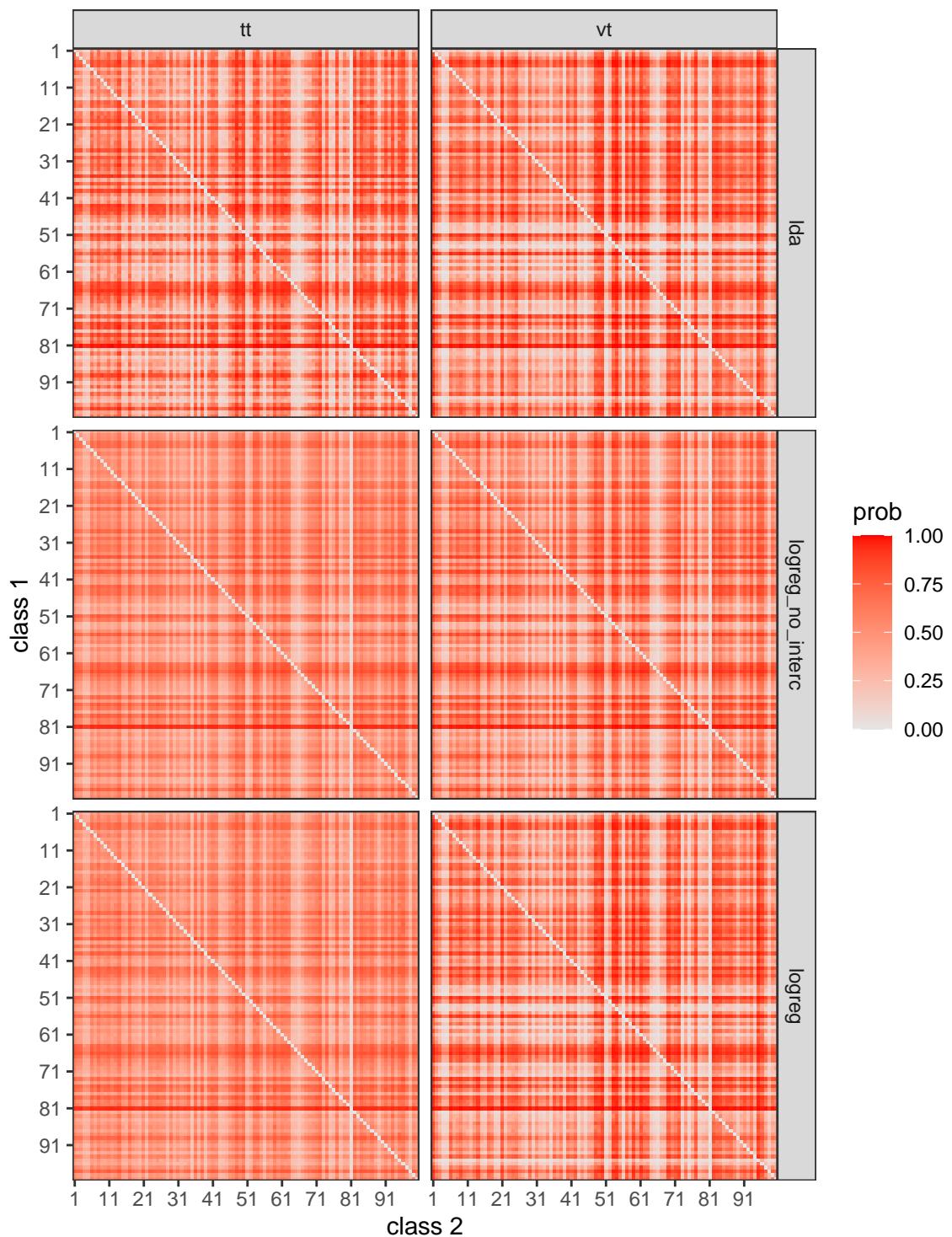
Average pairwise probabilities – class 79



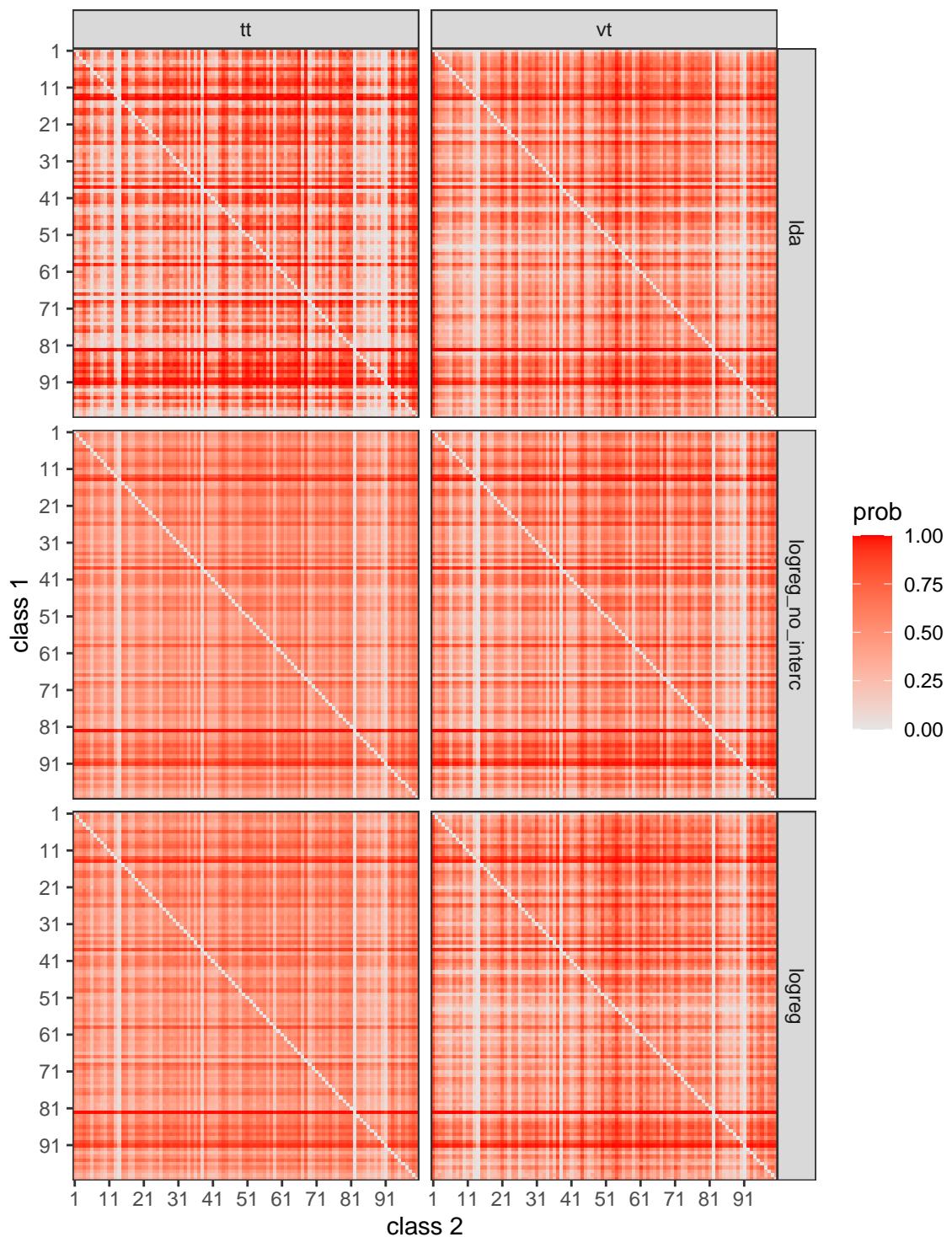
Average pairwise probabilities – class 80



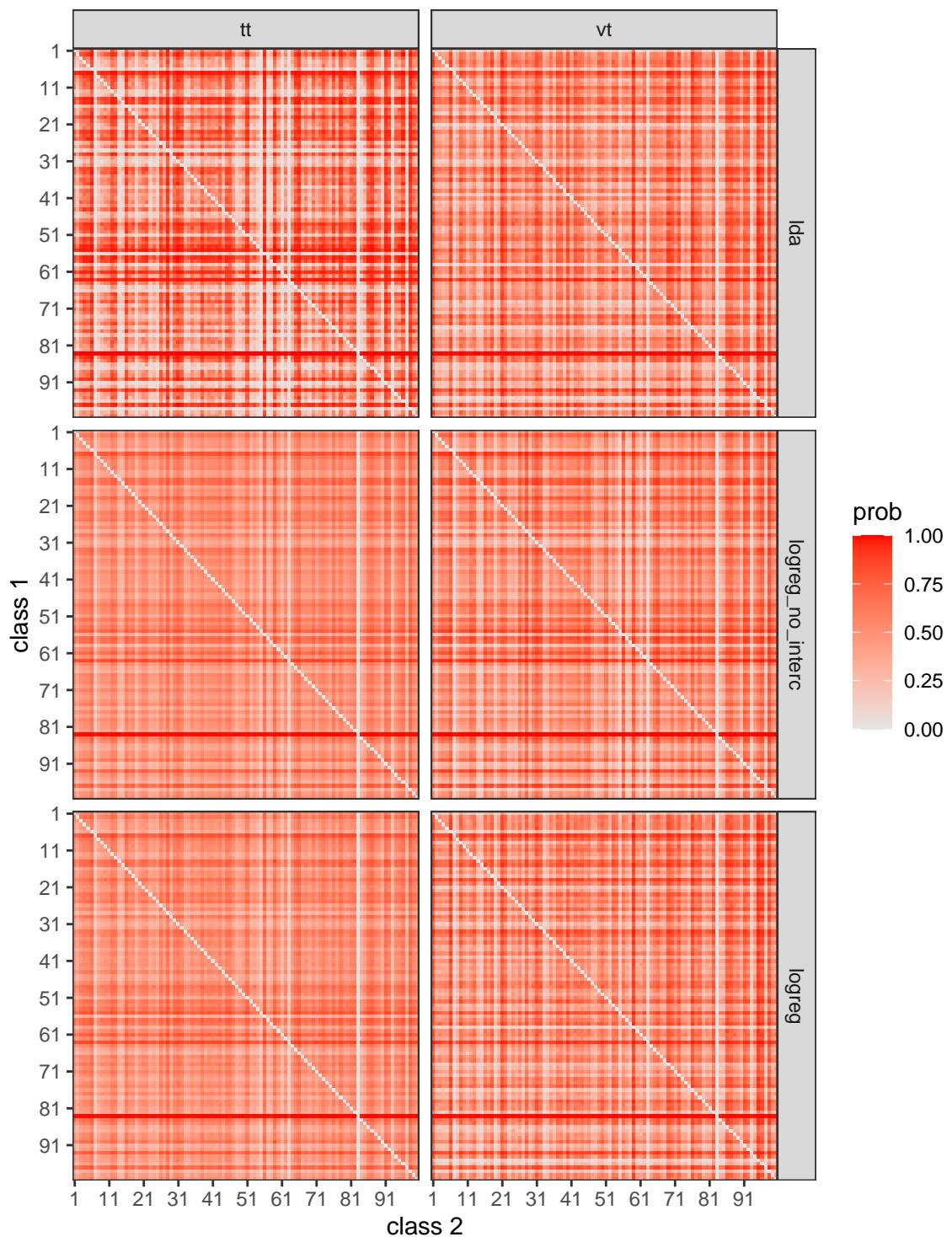
Average pairwise probabilities – class 81



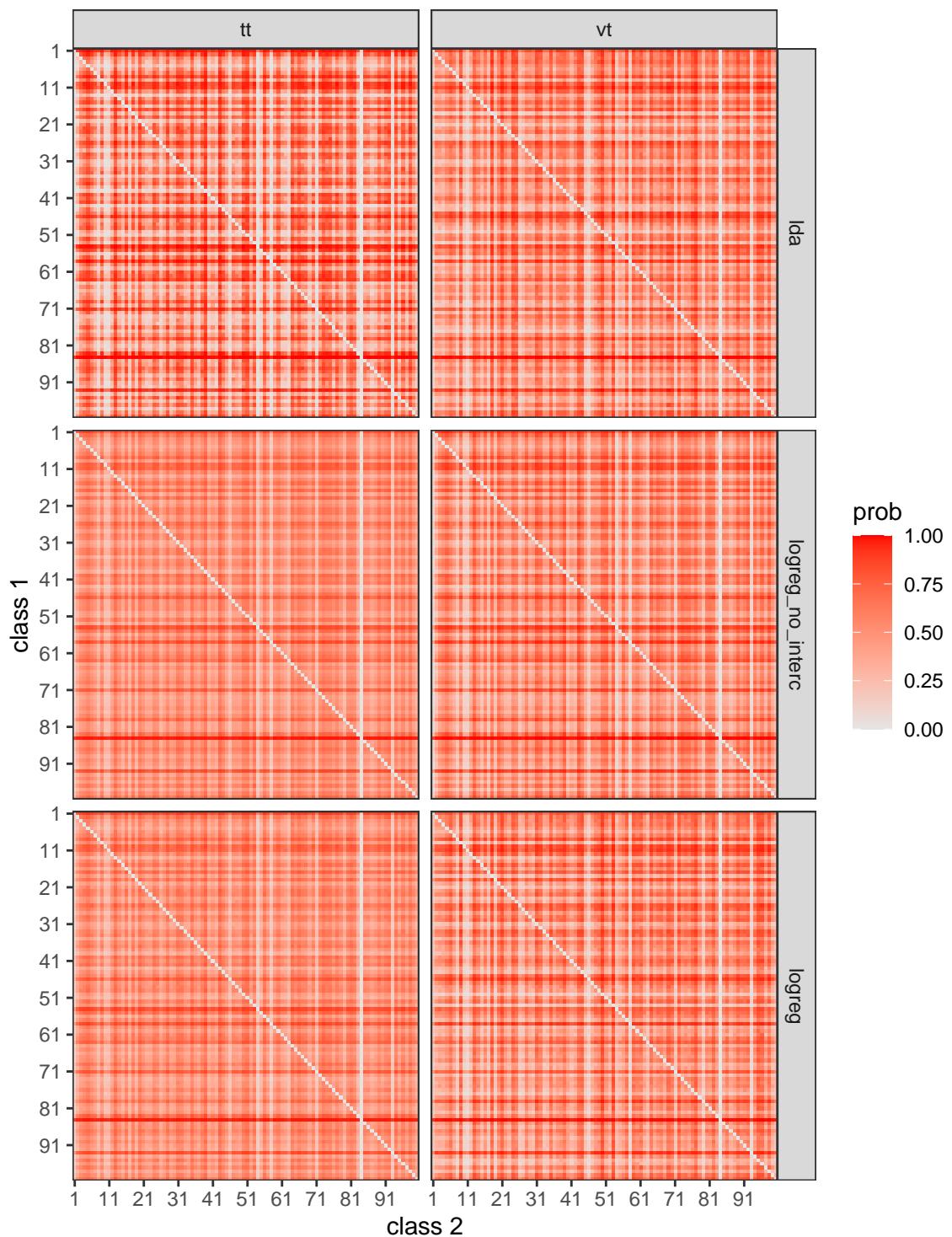
Average pairwise probabilities – class 82



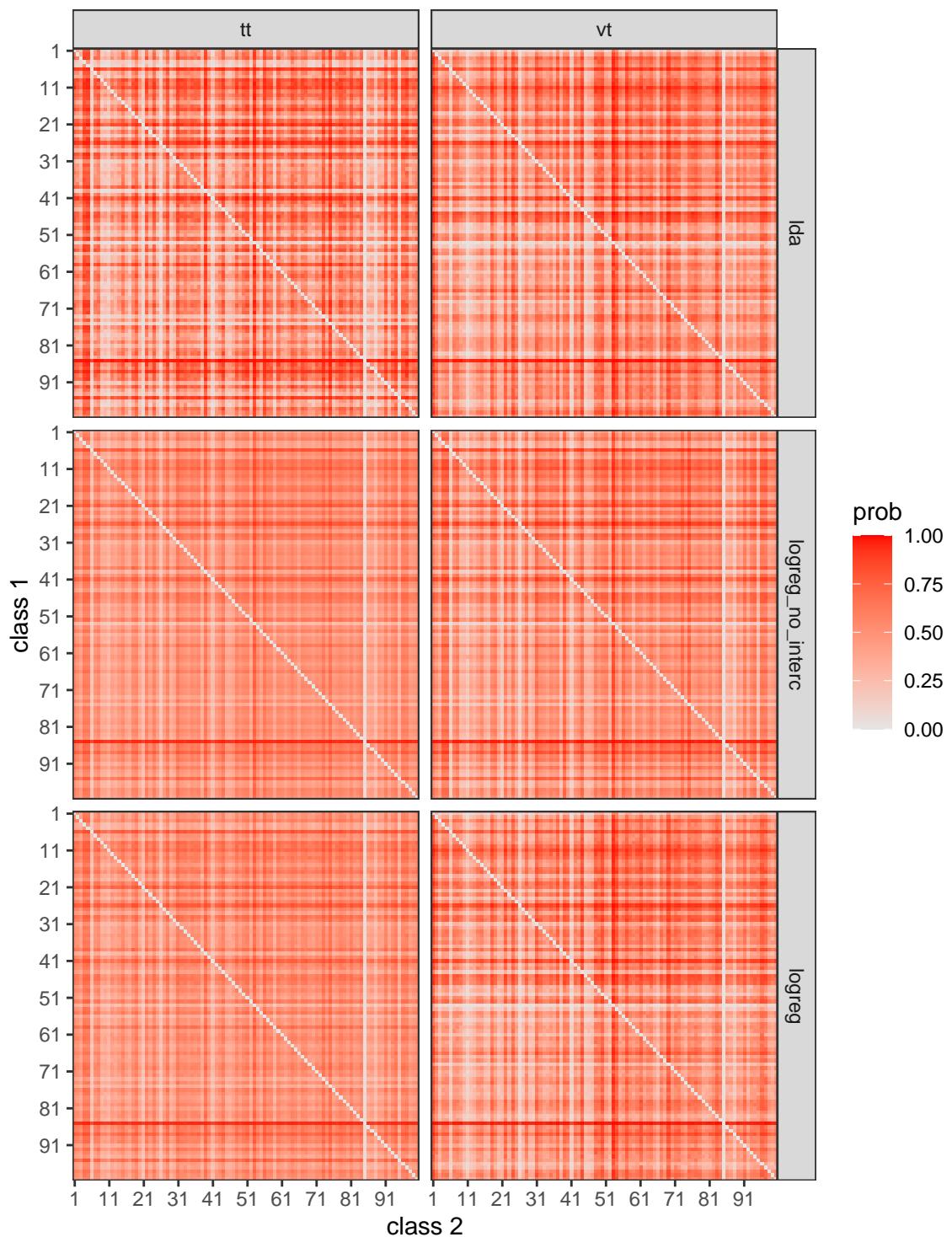
Average pairwise probabilities – class 83



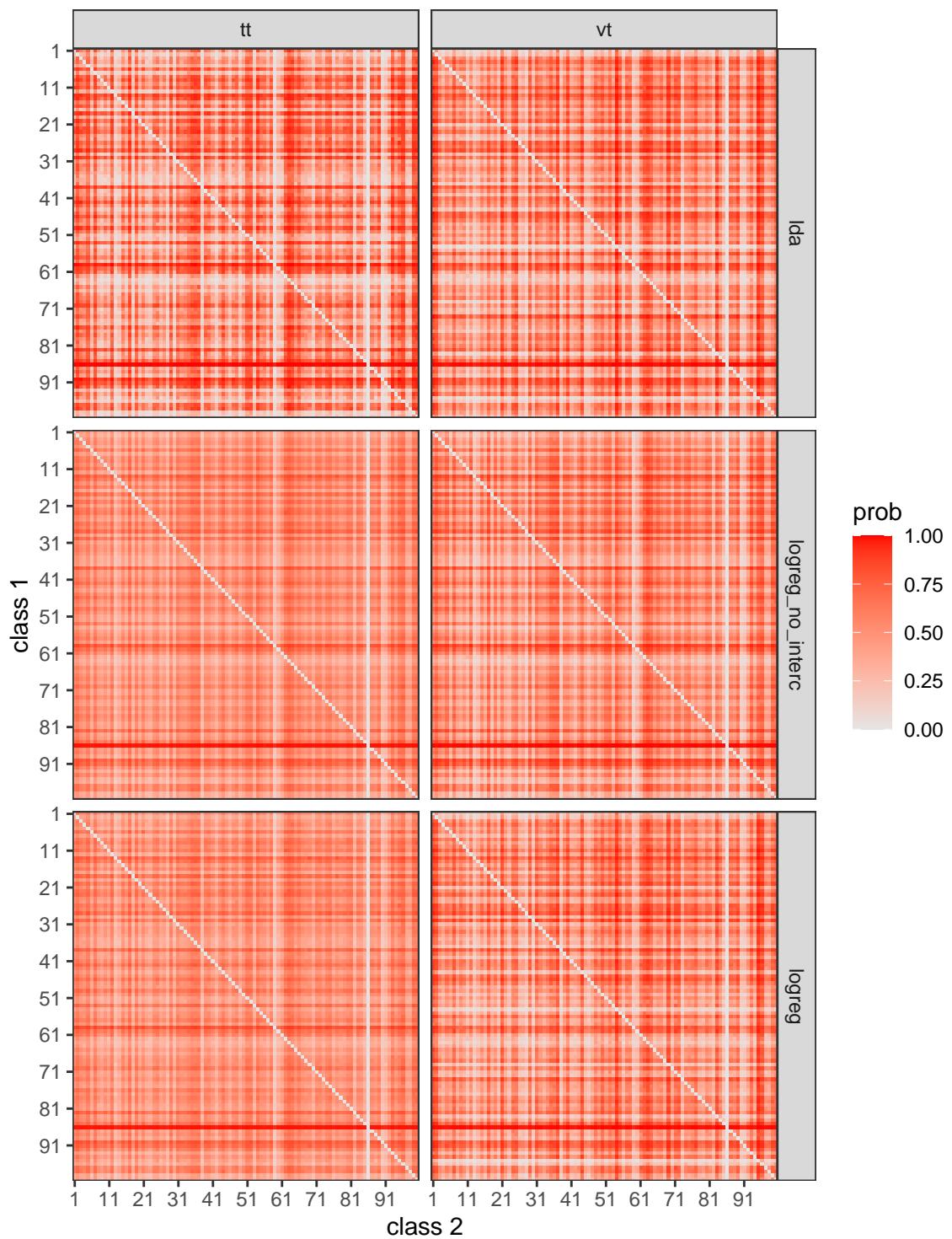
Average pairwise probabilities – class 84



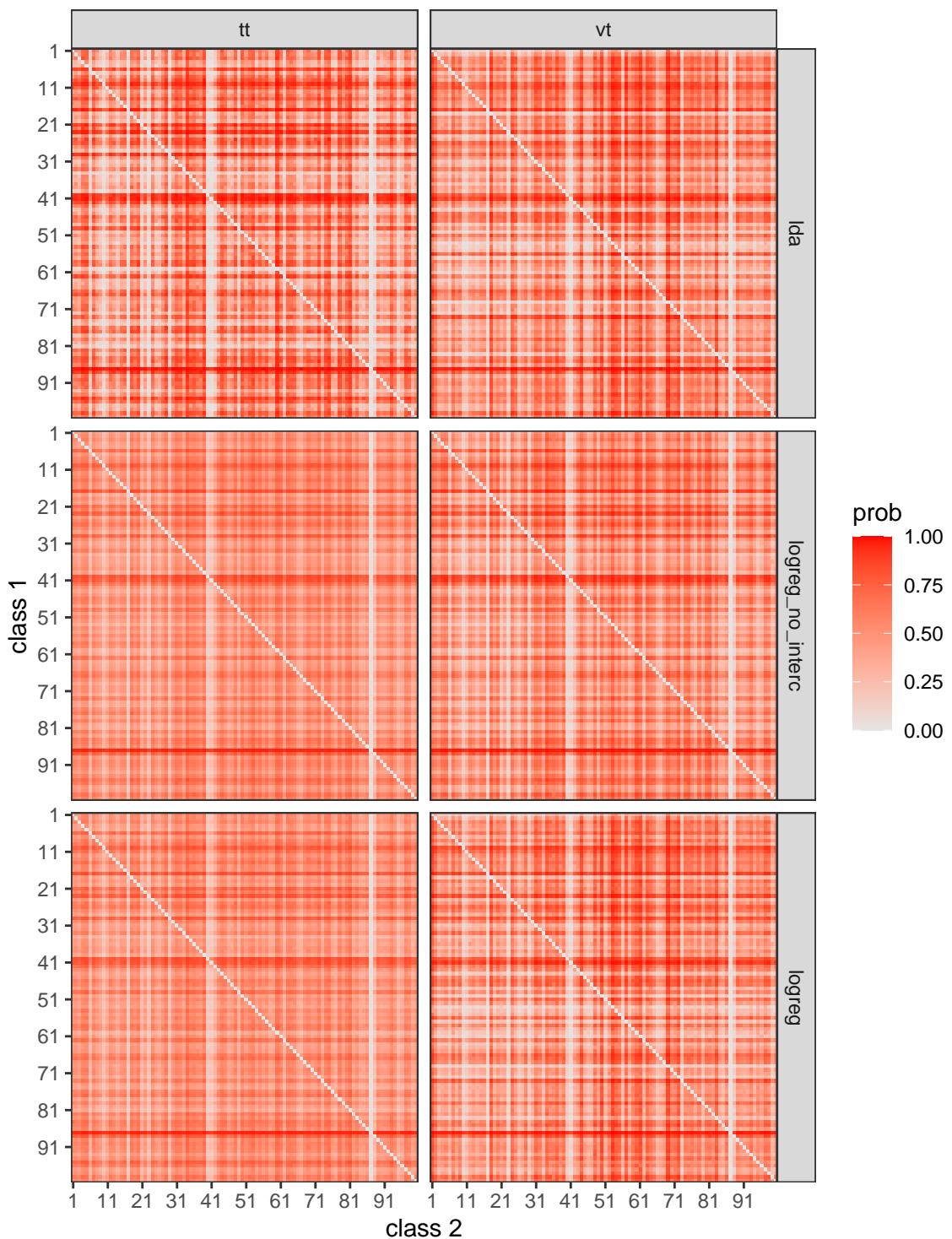
Average pairwise probabilities – class 85



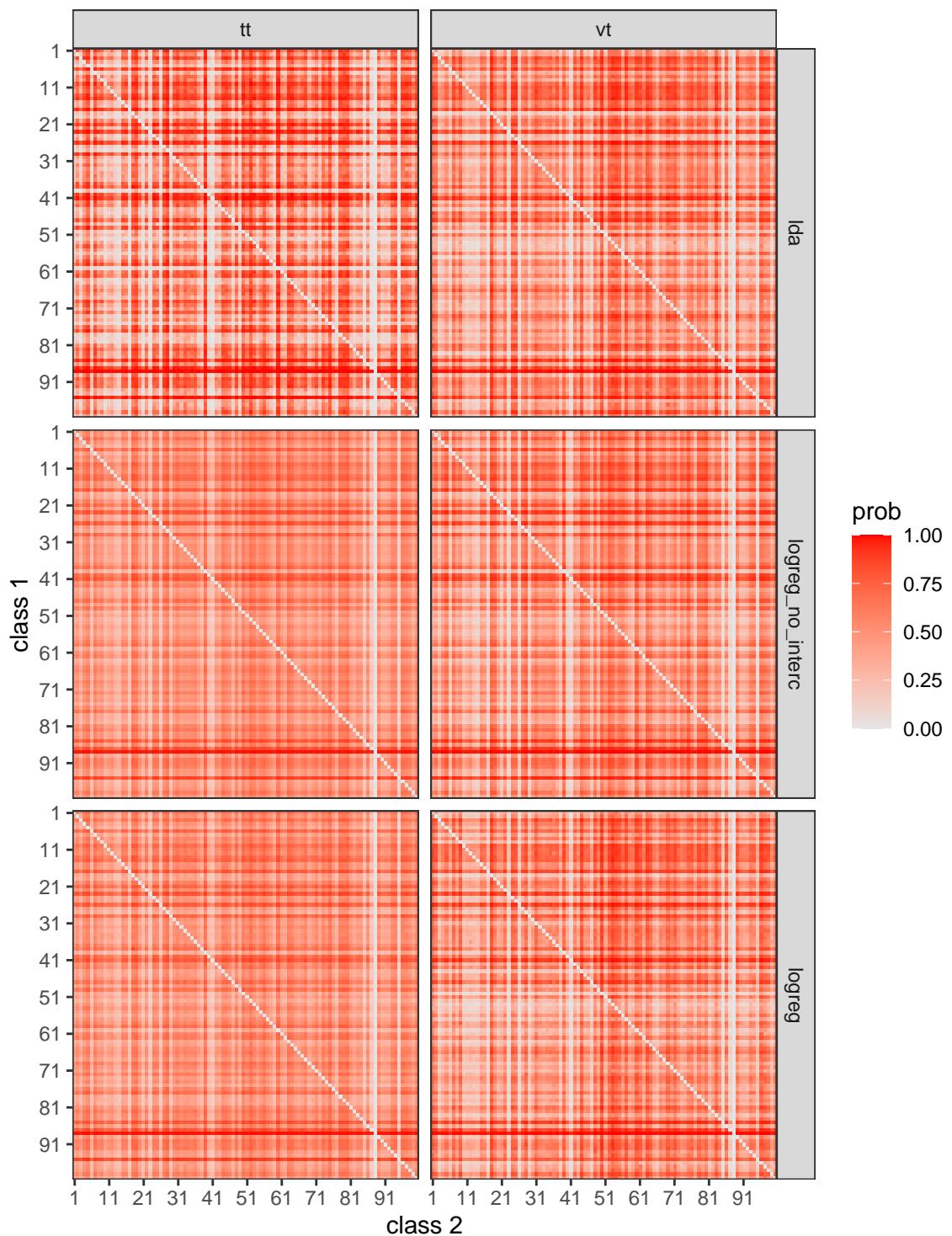
Average pairwise probabilities – class 86



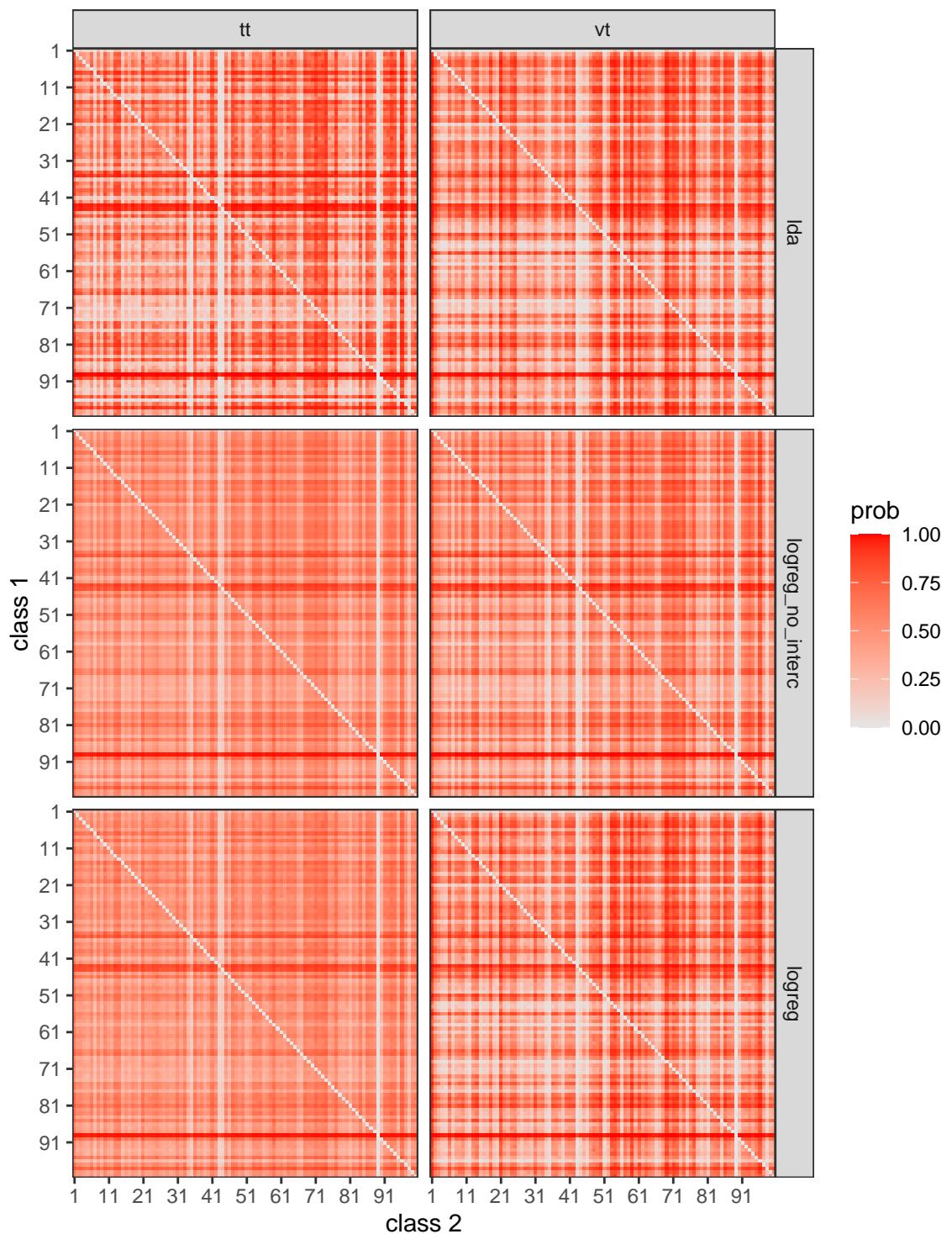
Average pairwise probabilities – class 87



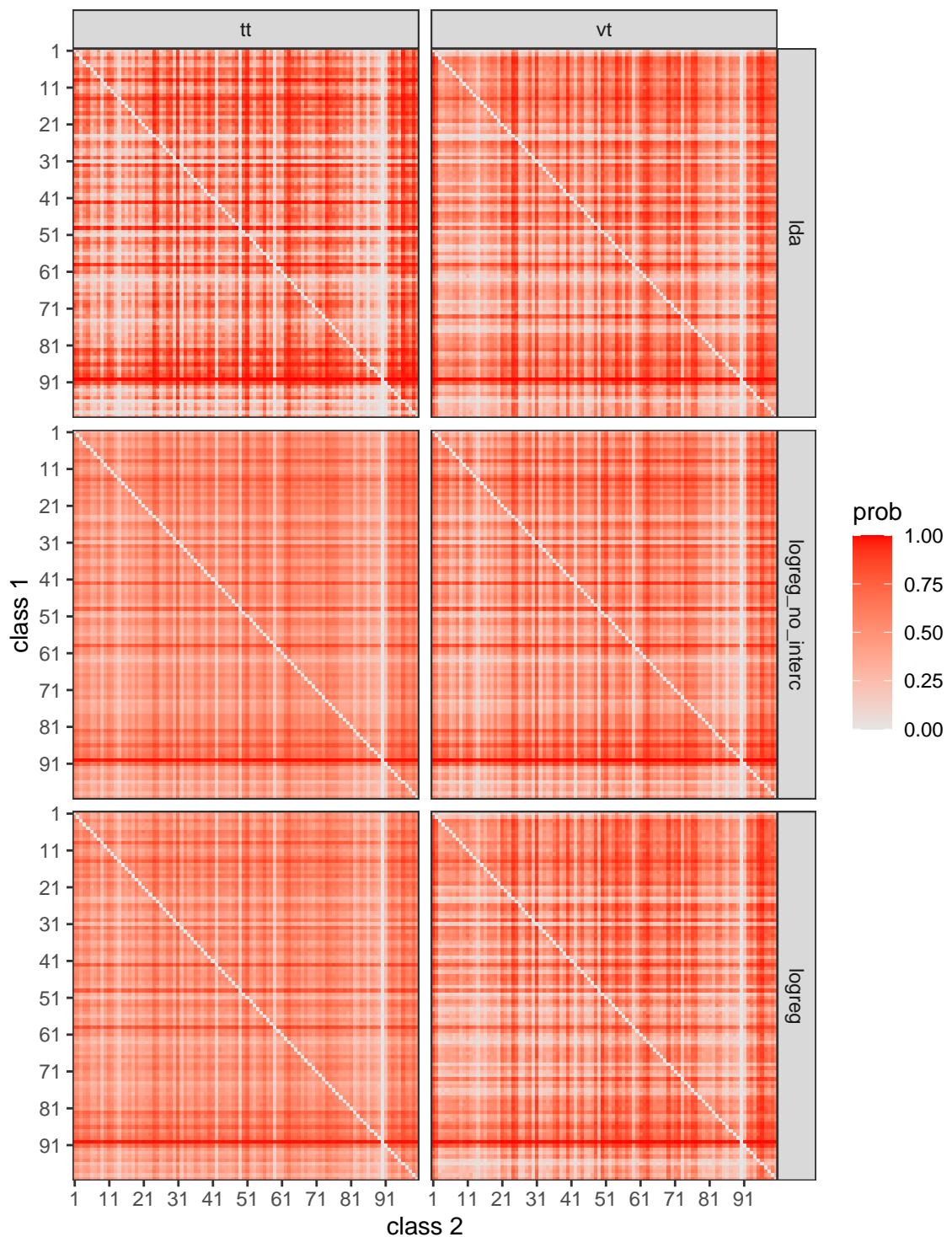
Average pairwise probabilities – class 88



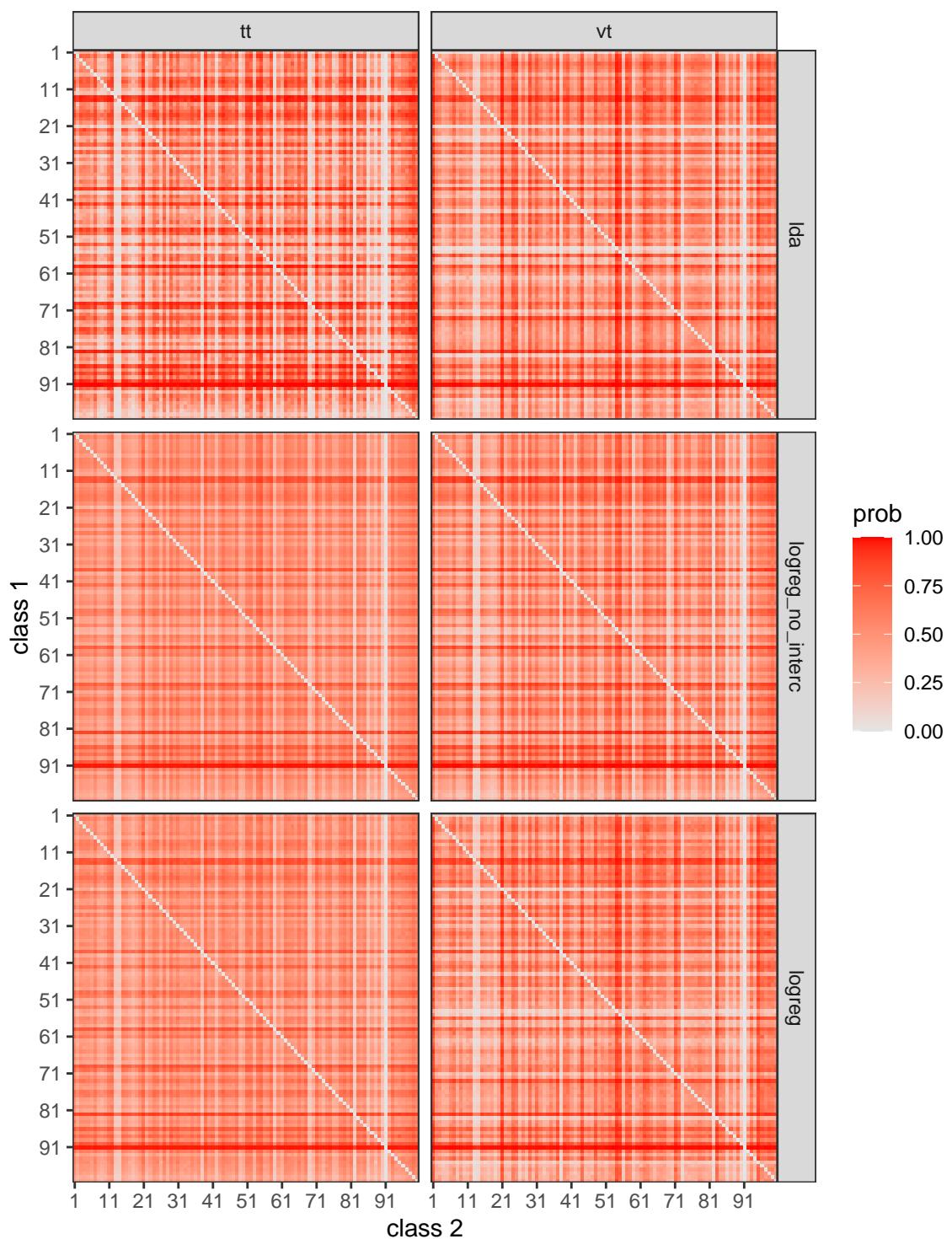
Average pairwise probabilities – class 89



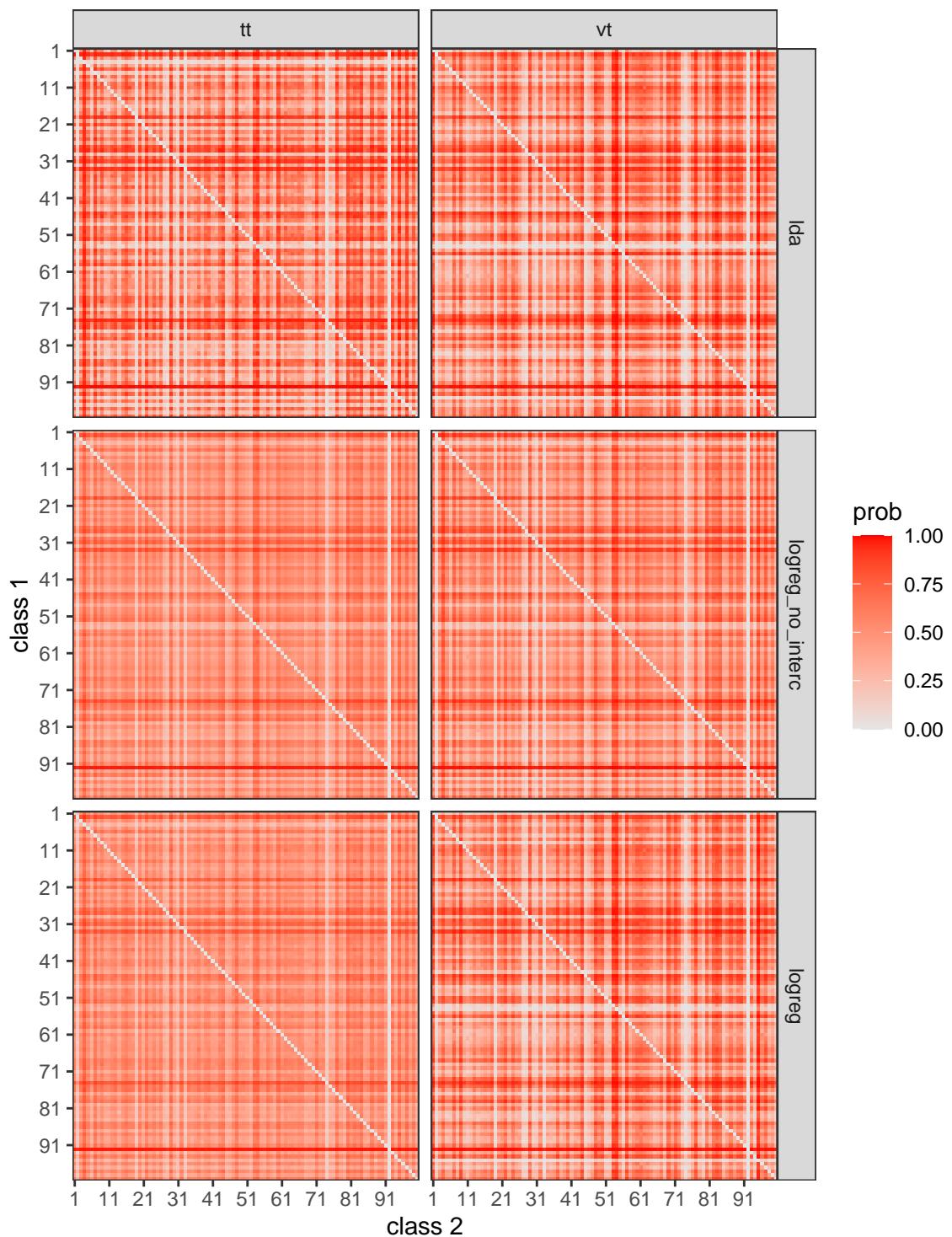
Average pairwise probabilities – class 90



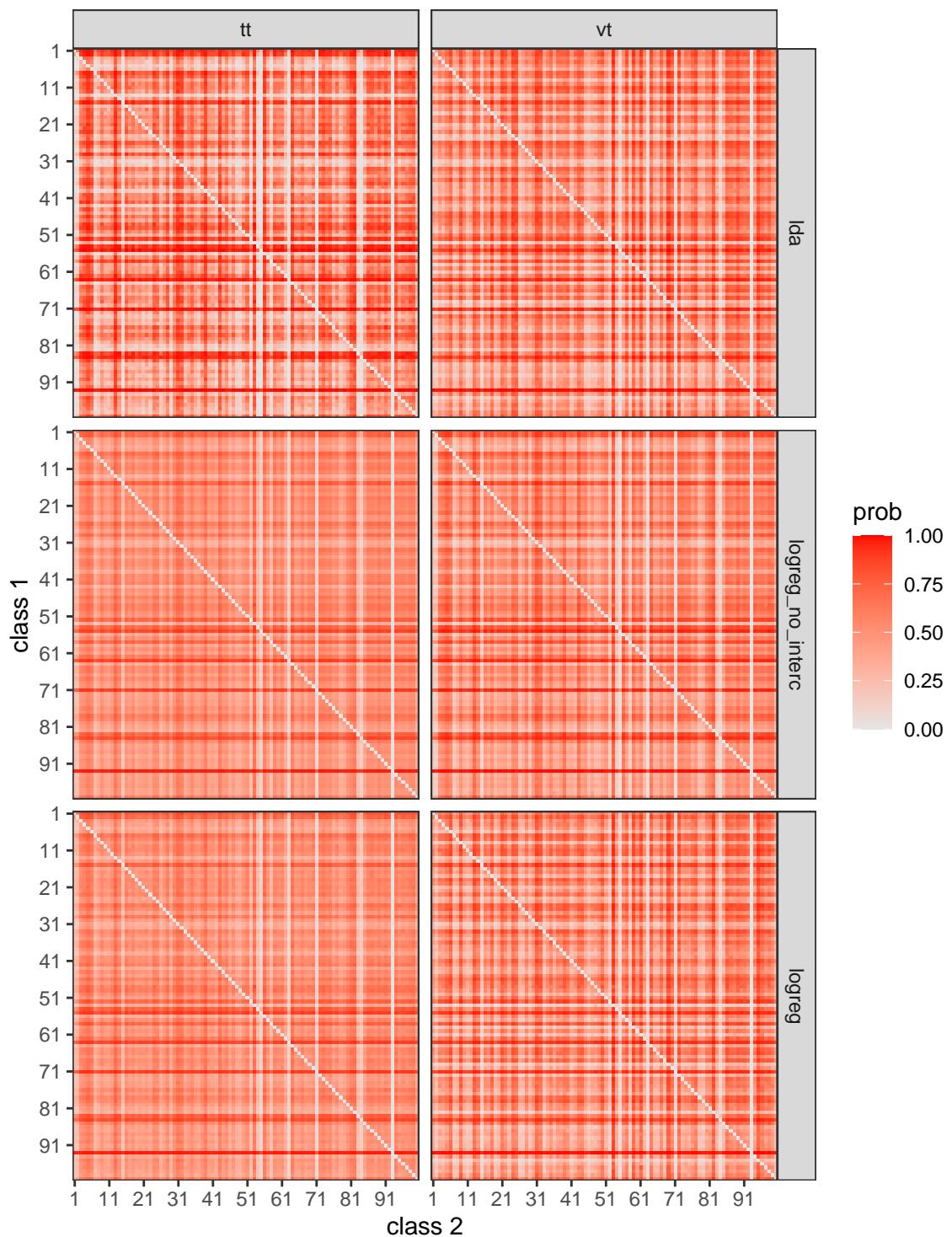
Average pairwise probabilities – class 91



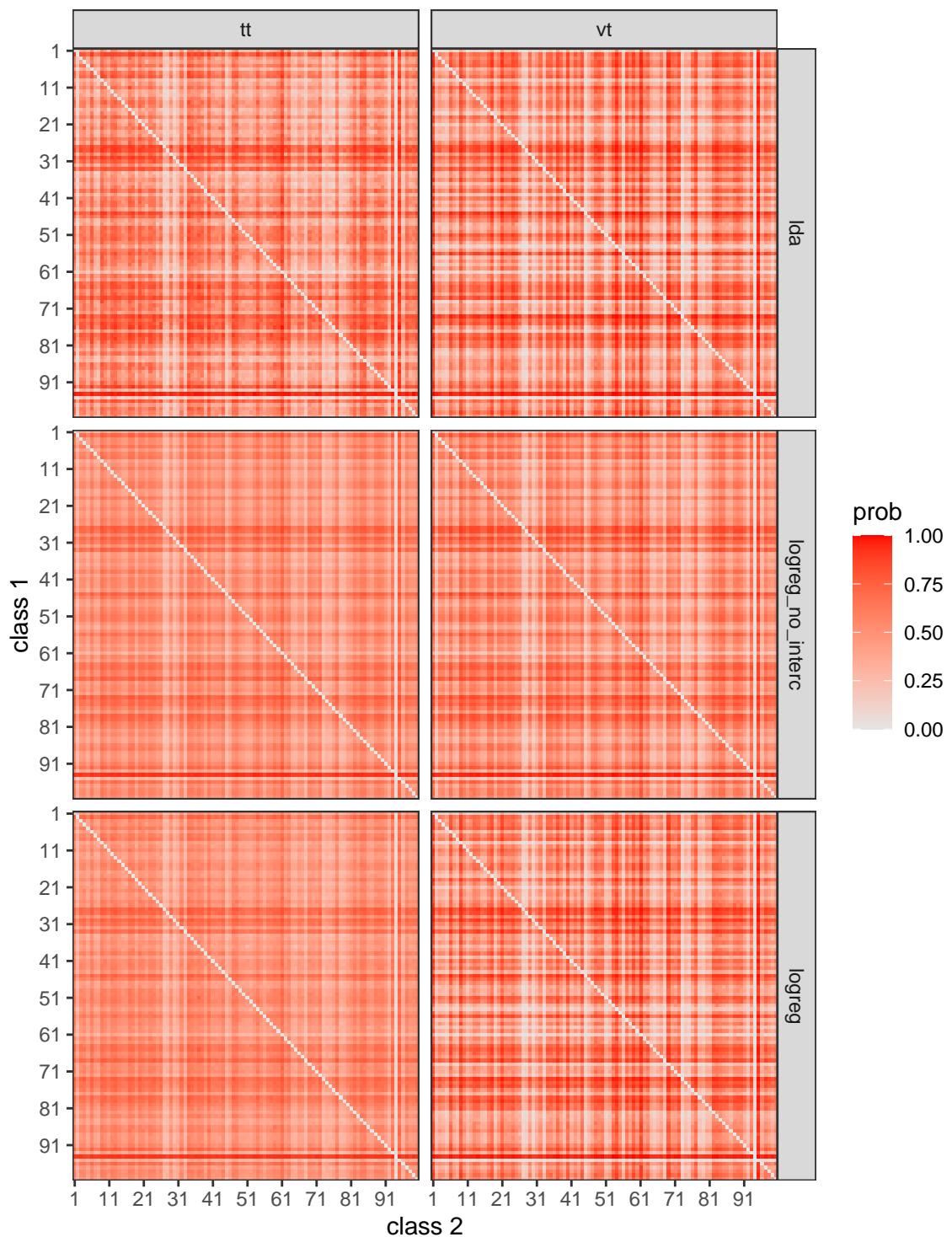
Average pairwise probabilities – class 92



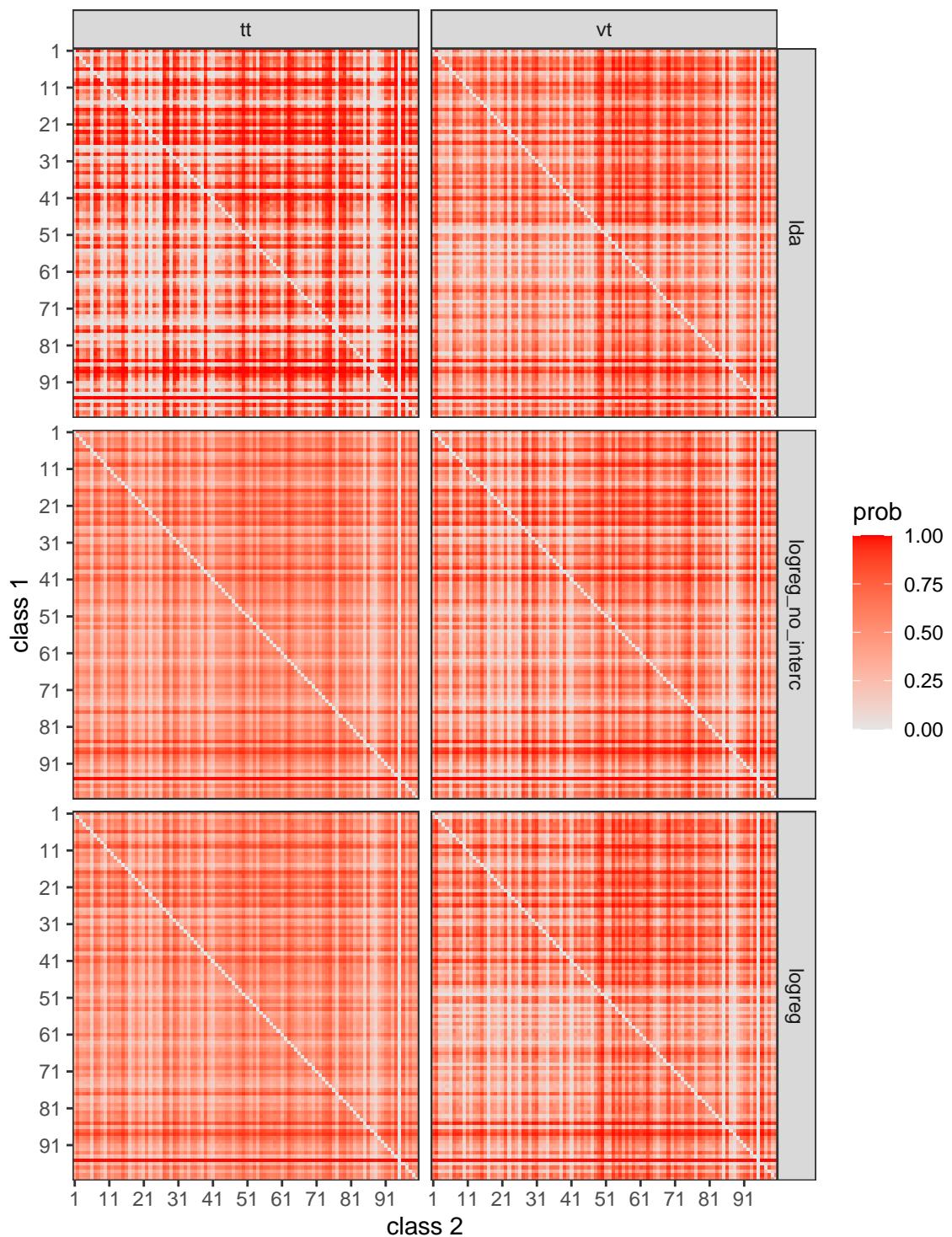
Average pairwise probabilities – class 93



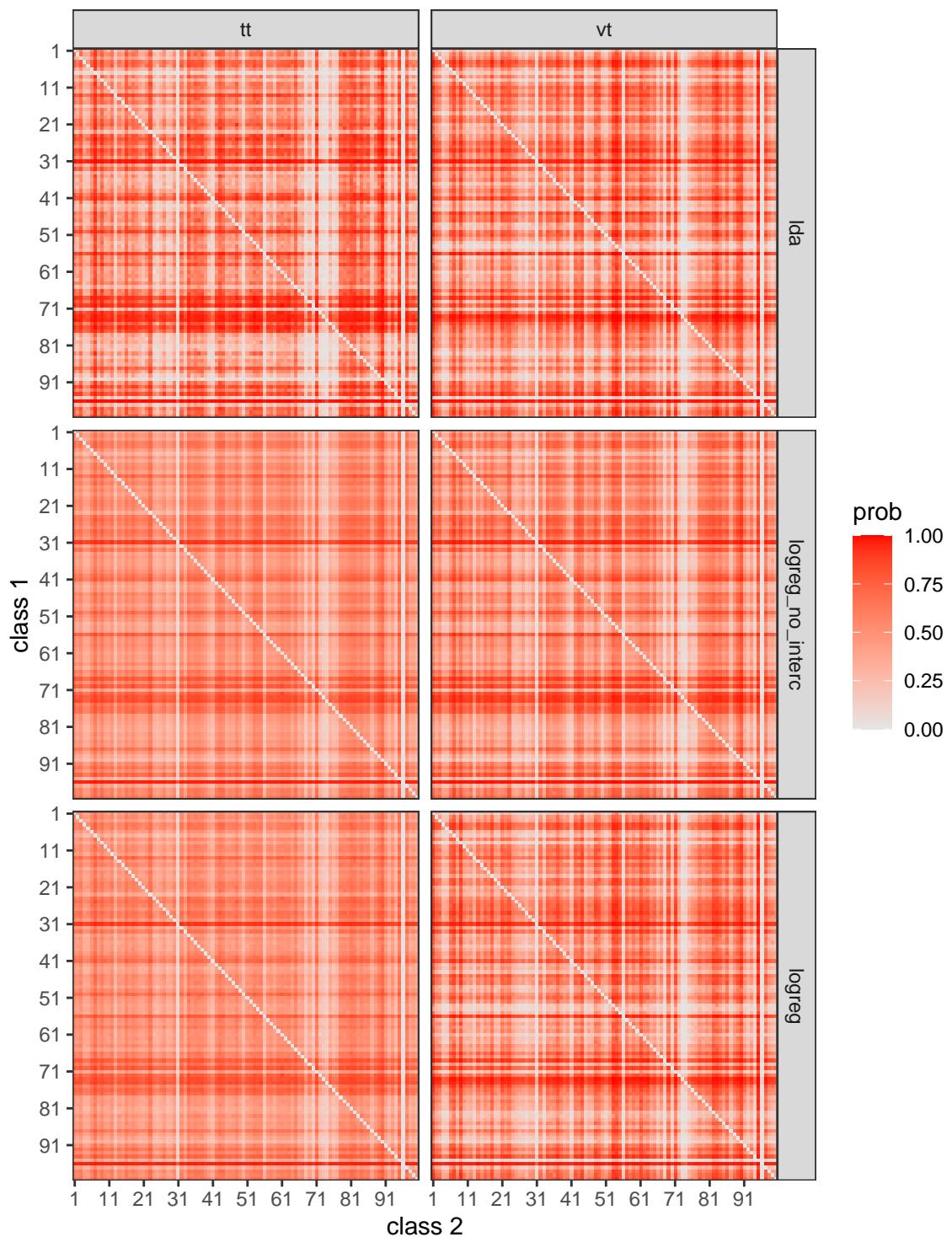
Average pairwise probabilities – class 94



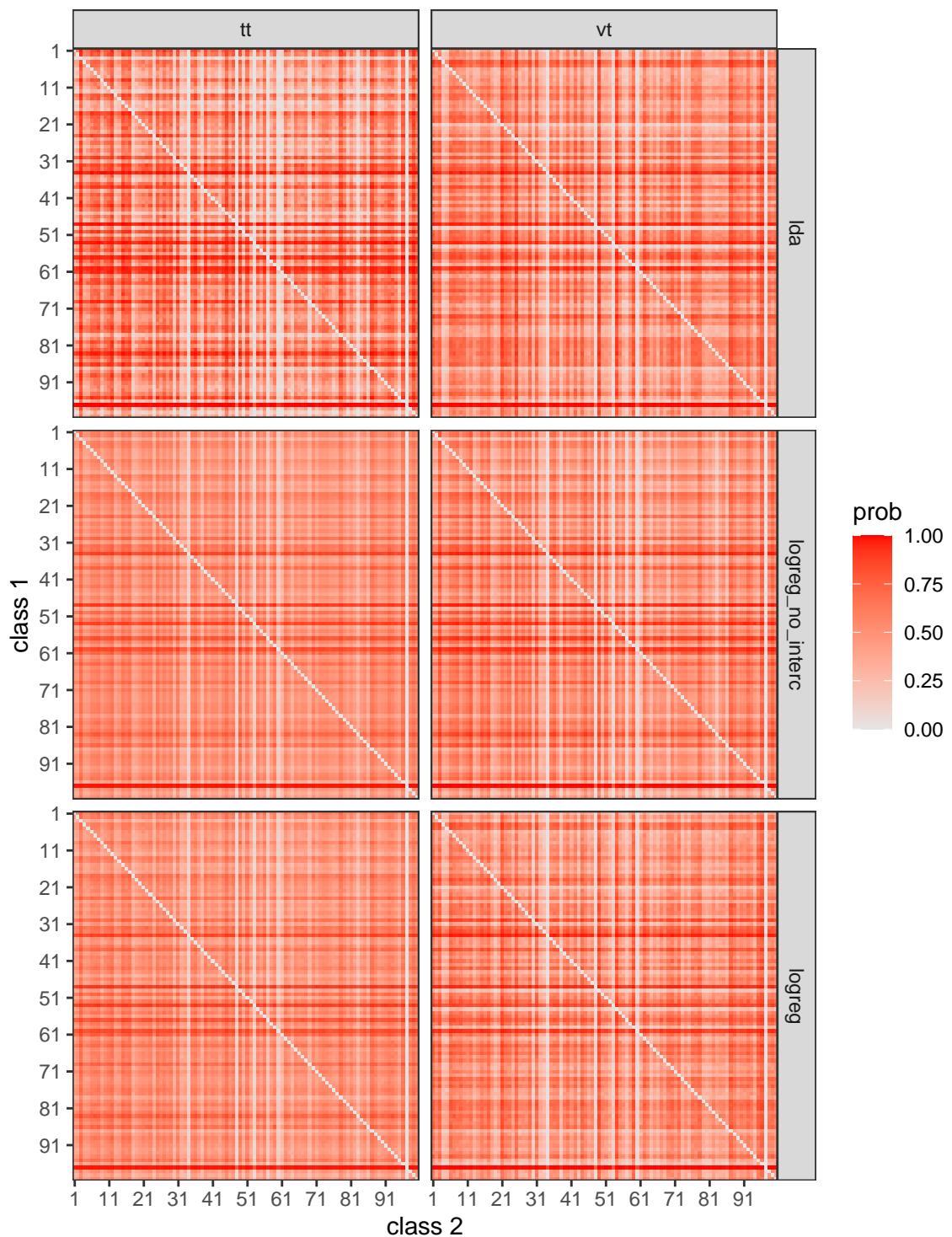
Average pairwise probabilities – class 95



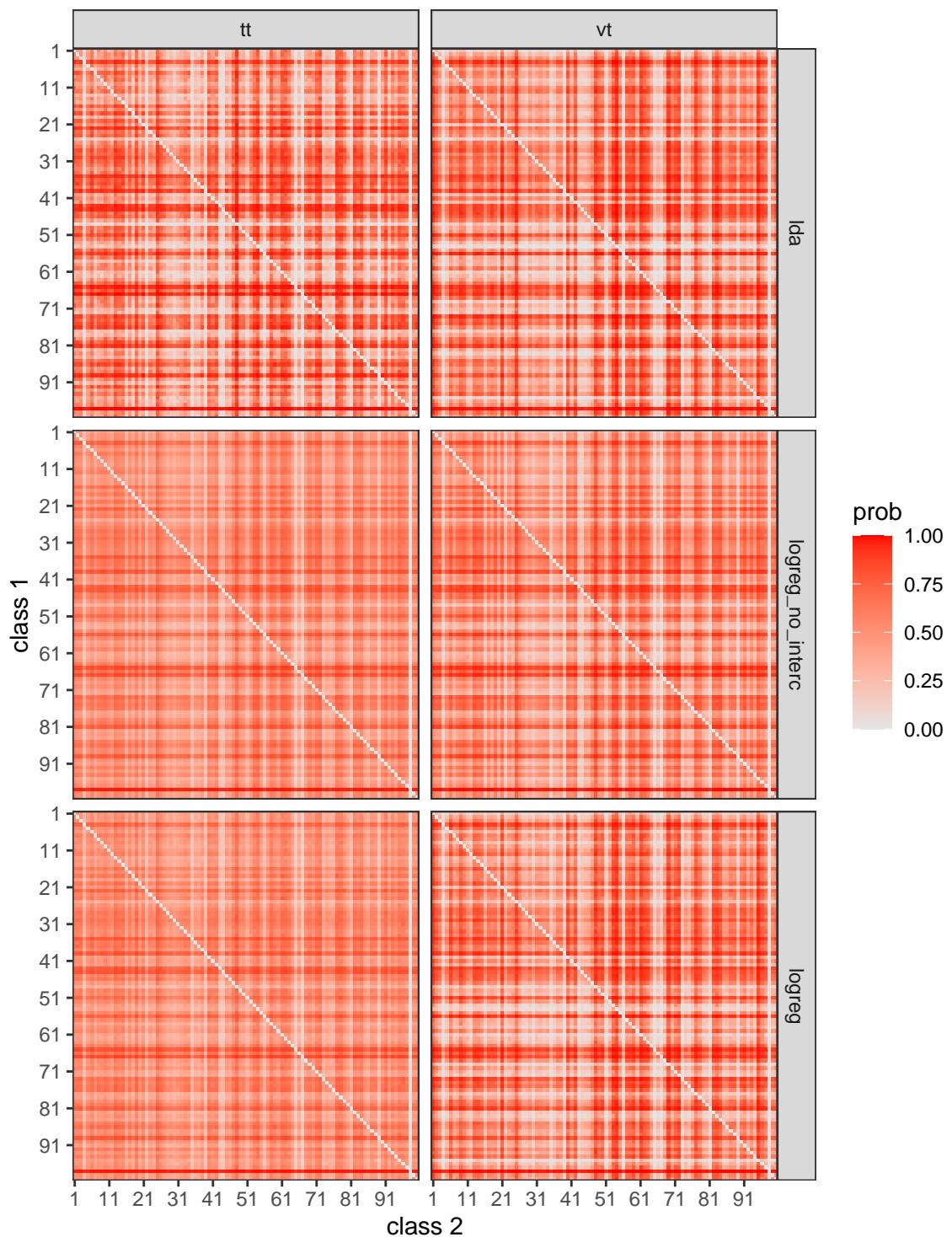
Average pairwise probabilities – class 96



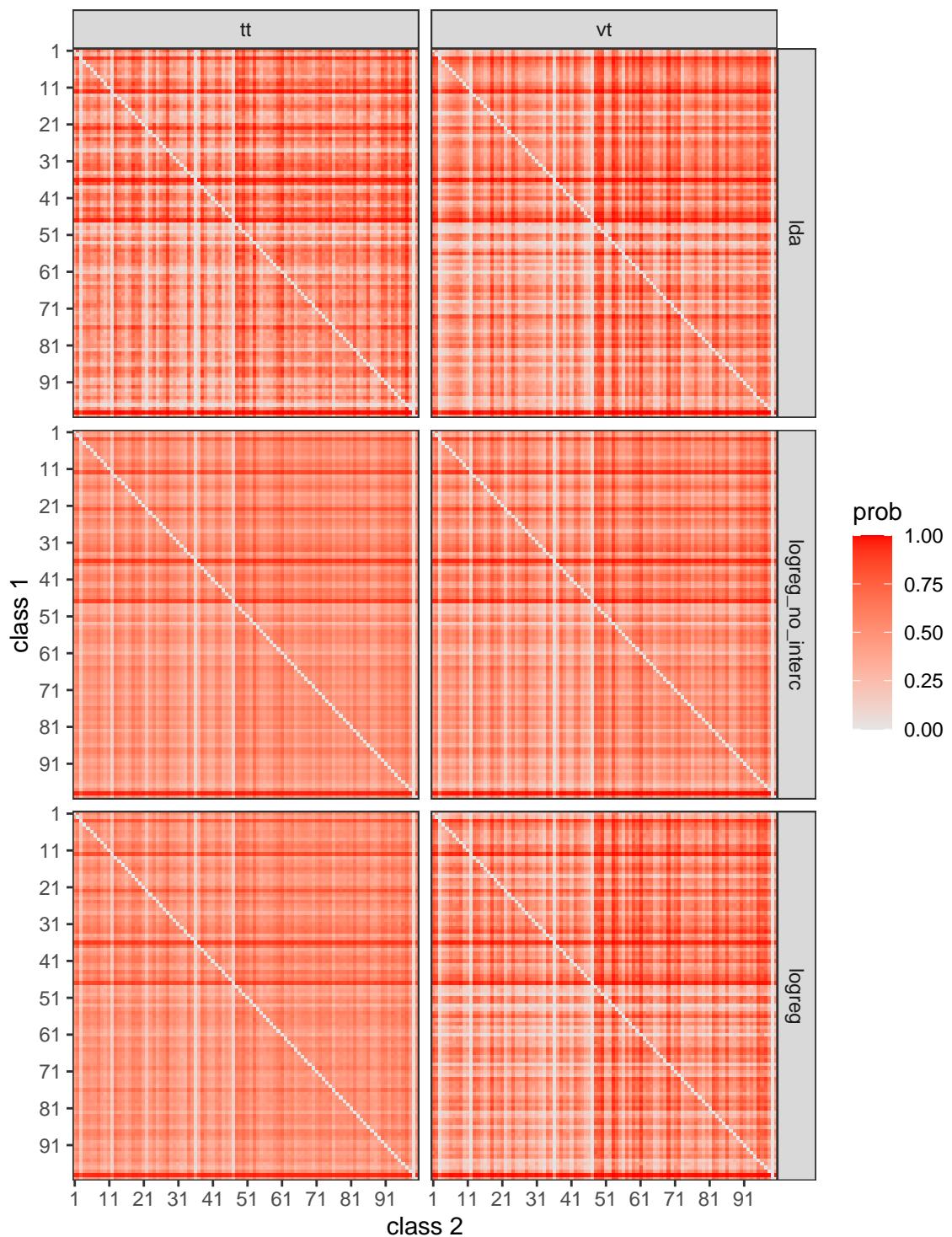
Average pairwise probabilities – class 97



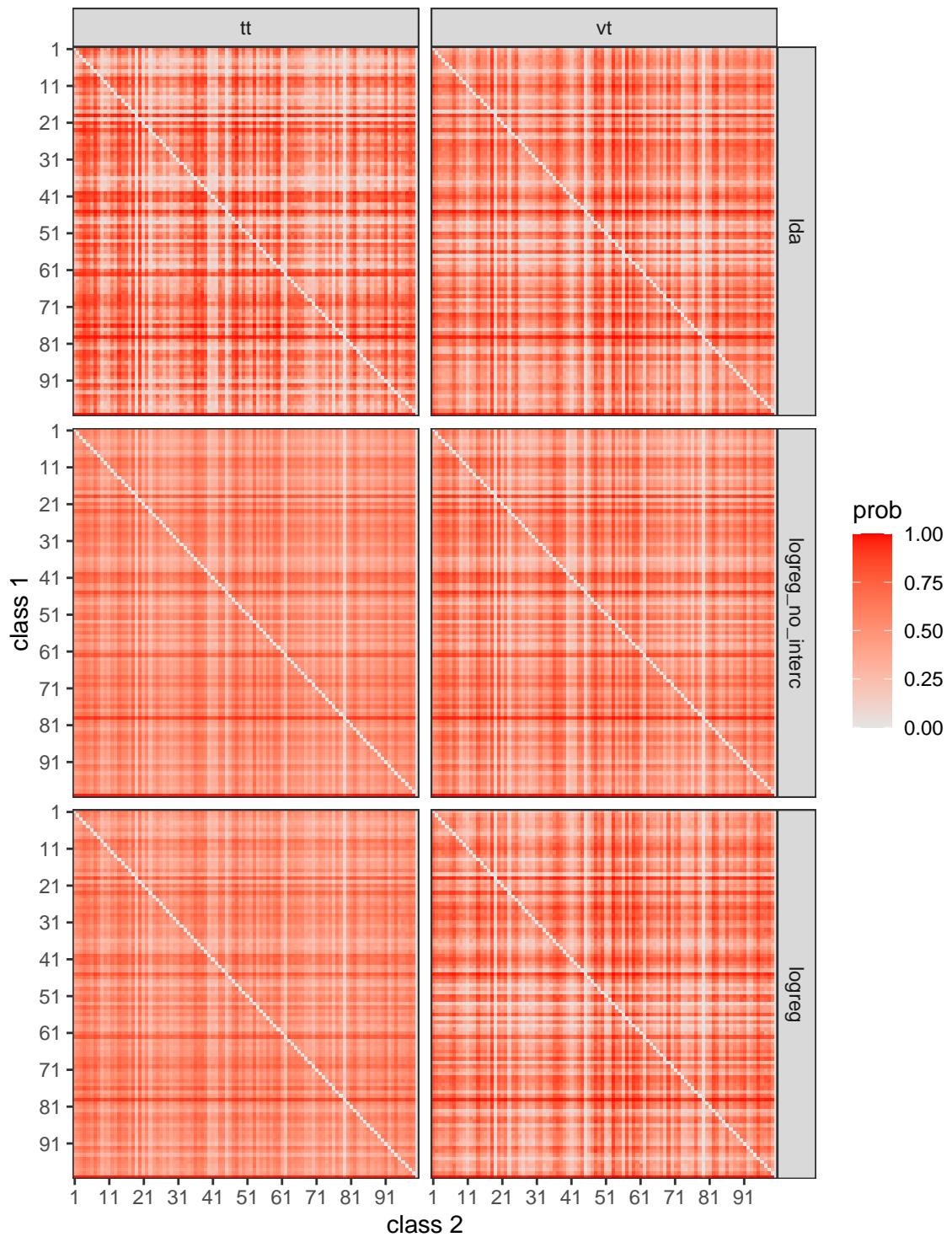
Average pairwise probabilities – class 98



Average pairwise probabilities – class 99



### Average pairwise probabilities – class 100



trices for logreg are less sharp - this is visible mainly for tt.

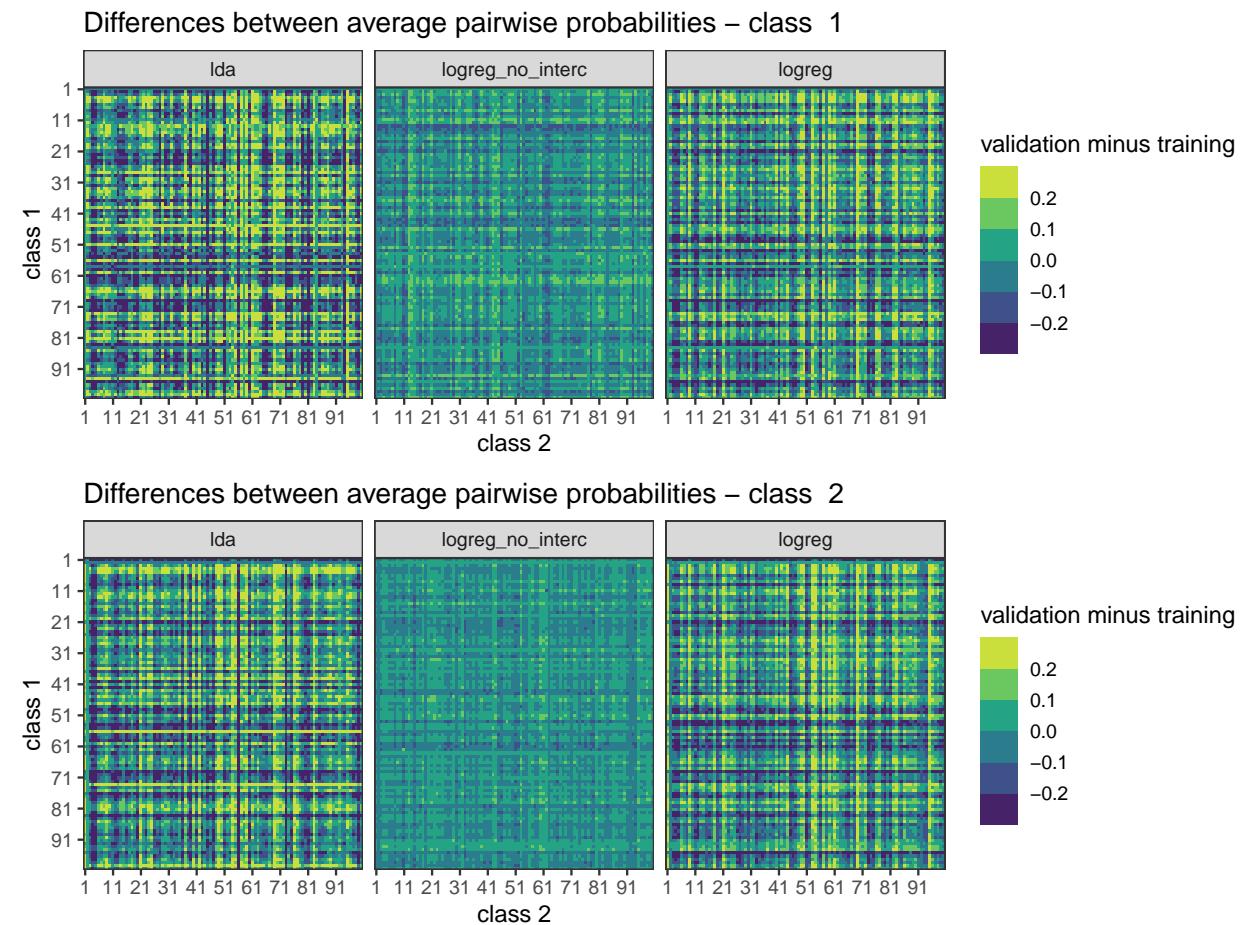
```
for (cls in 1:classes)
{
  cur_class_Rs <- df_aggr_Rs_diff %>% filter(class == cls)
  plot_cls <- ggplot(cur_class_Rs, aes(x = class2, y = class1)) +
```

```

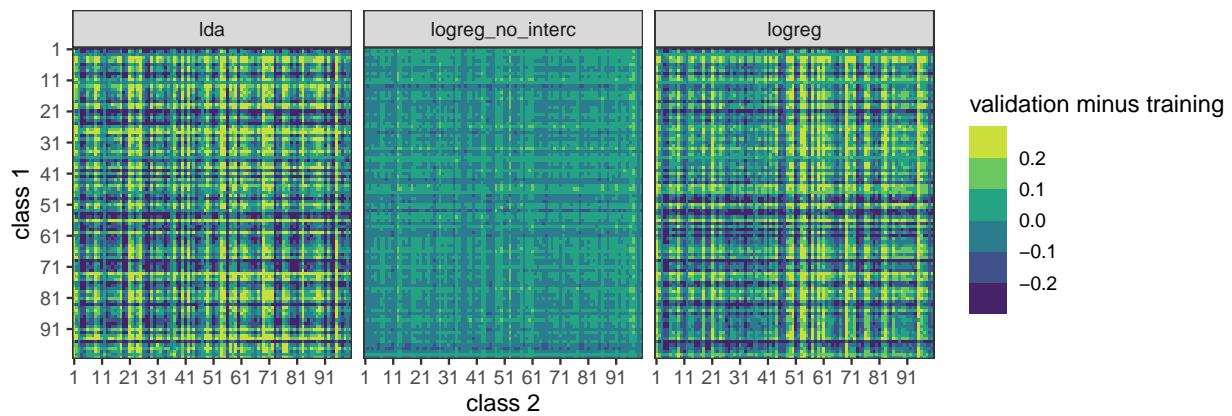
geom_raster(aes(fill=val_min_train)) +
facet_wrap(~combining_method) +
scale_fill_binned(type="viridis", limits=c(-0.3, 0.3), name="validation minus training") +
scale_y_discrete(limits=rev, breaks=seq(1, 100, 10)) +
scale_x_discrete(breaks=seq(1, 100, 10)) +
labs(x="class 2", y="class 1", title=paste("Differences between average pairwise probabilities - class 1"))
theme_bw()

print(plot_cls)
}

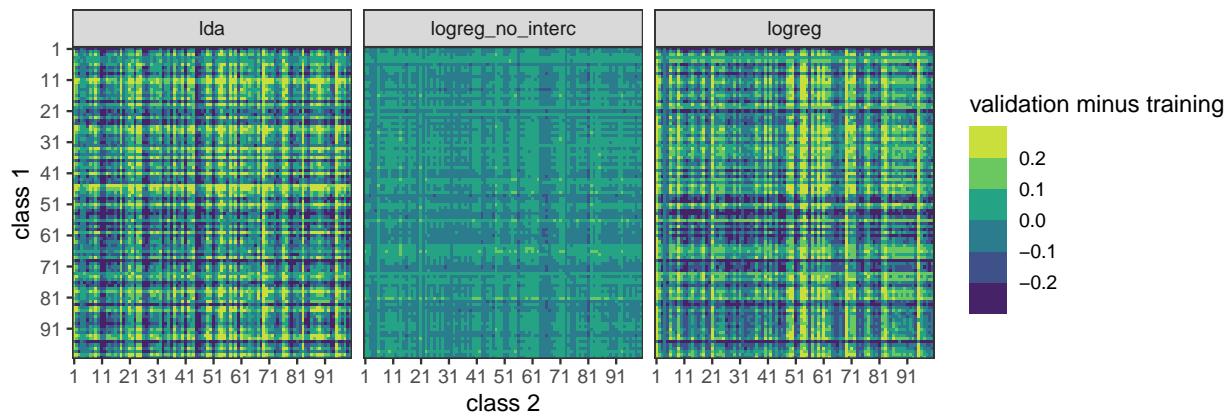
```



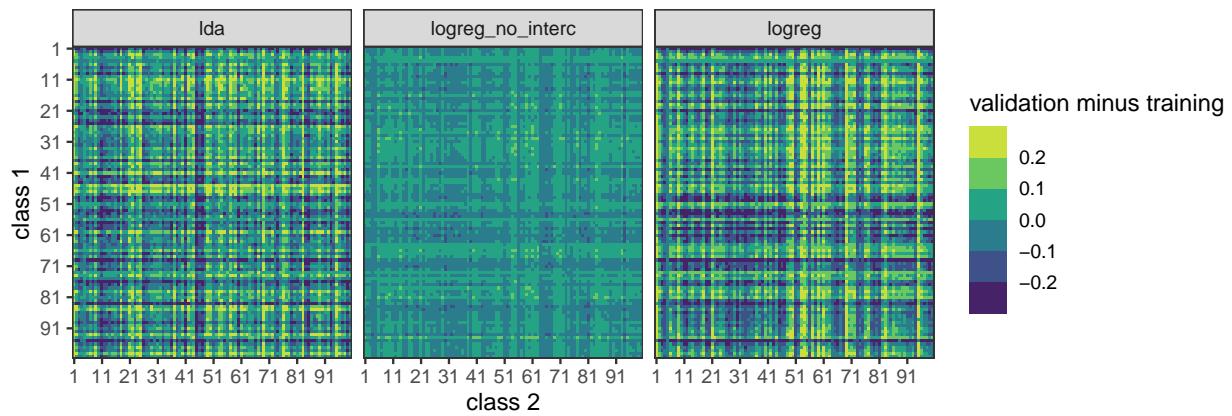
Differences between average pairwise probabilities – class 3



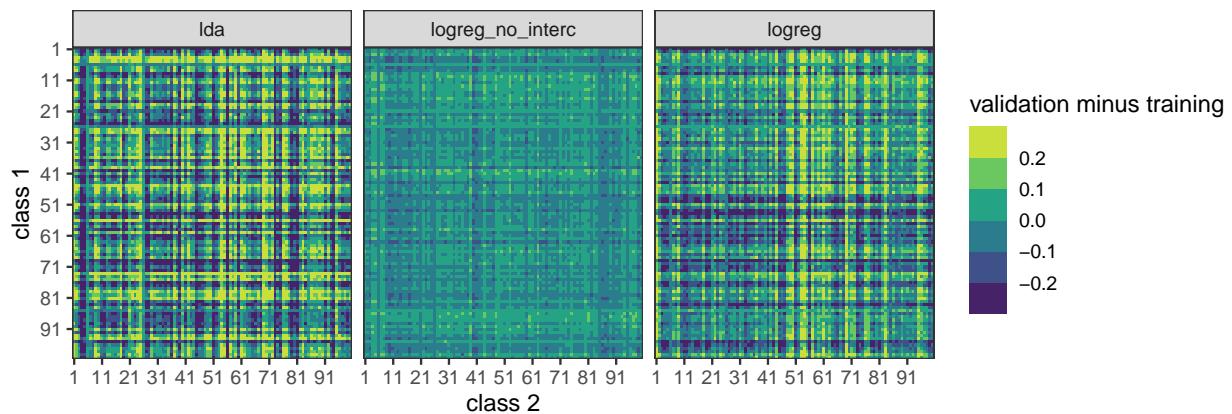
Differences between average pairwise probabilities – class 4



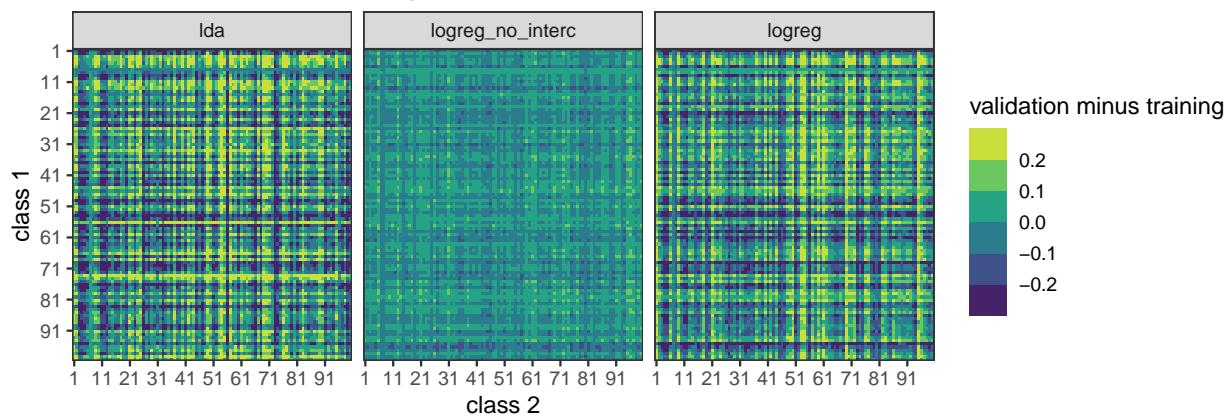
Differences between average pairwise probabilities – class 5



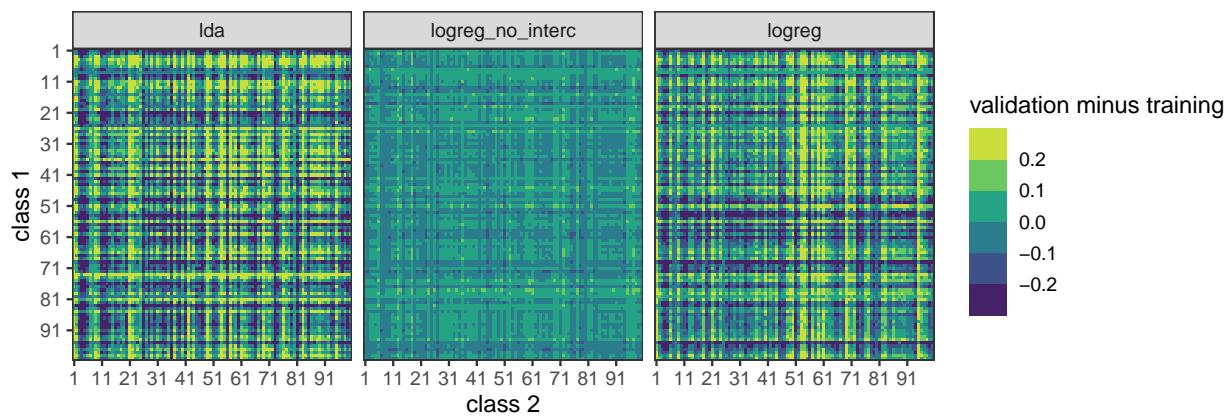
Differences between average pairwise probabilities – class 6



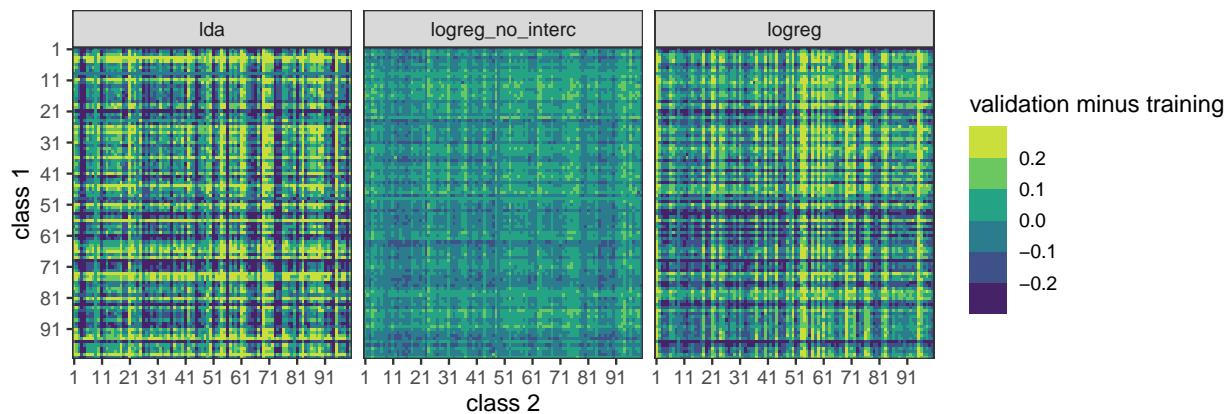
Differences between average pairwise probabilities – class 7



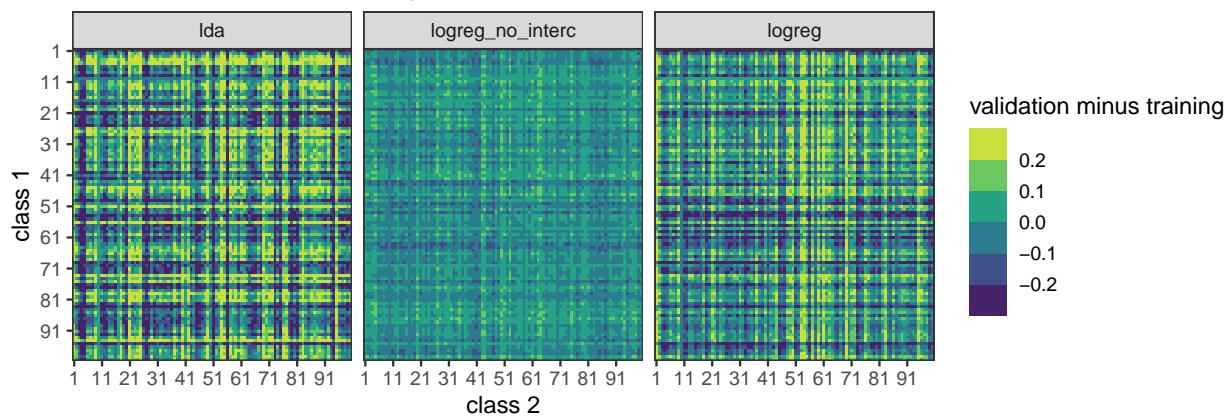
Differences between average pairwise probabilities – class 8



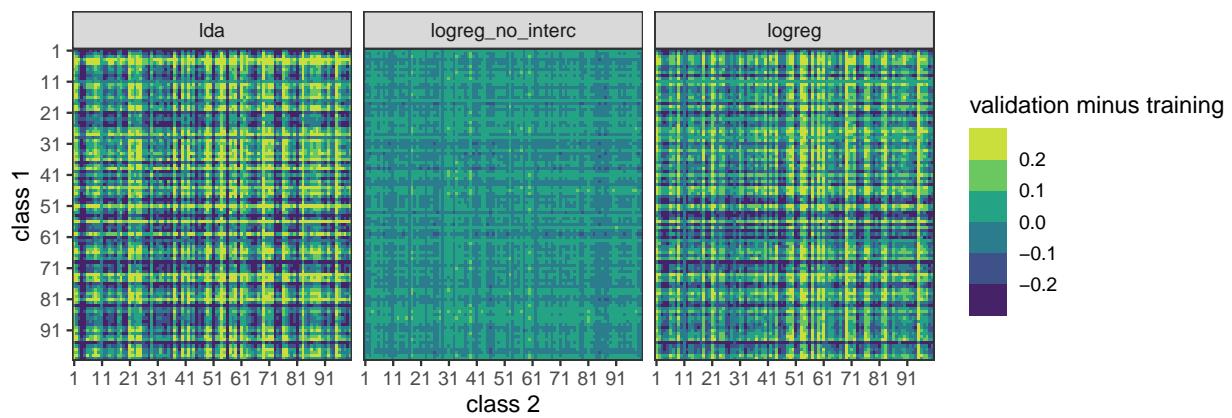
Differences between average pairwise probabilities – class 9



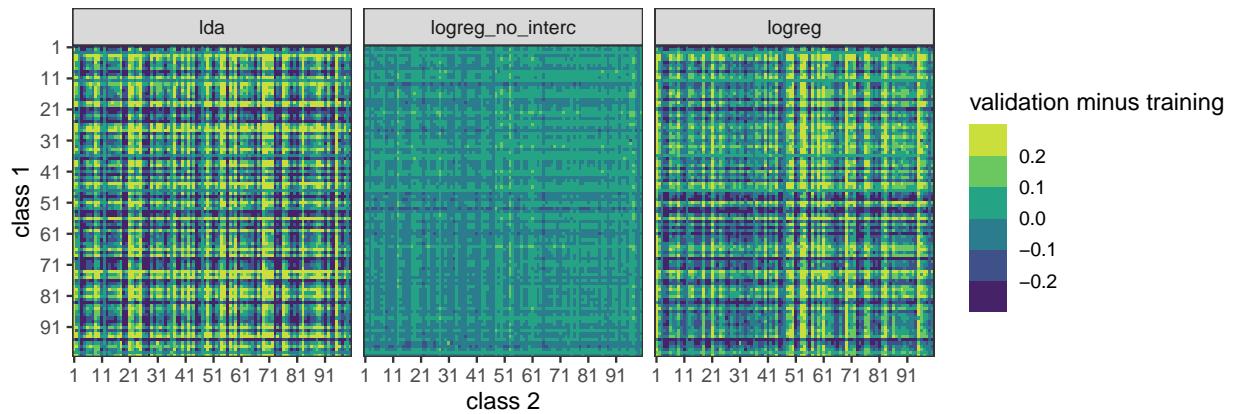
Differences between average pairwise probabilities – class 10



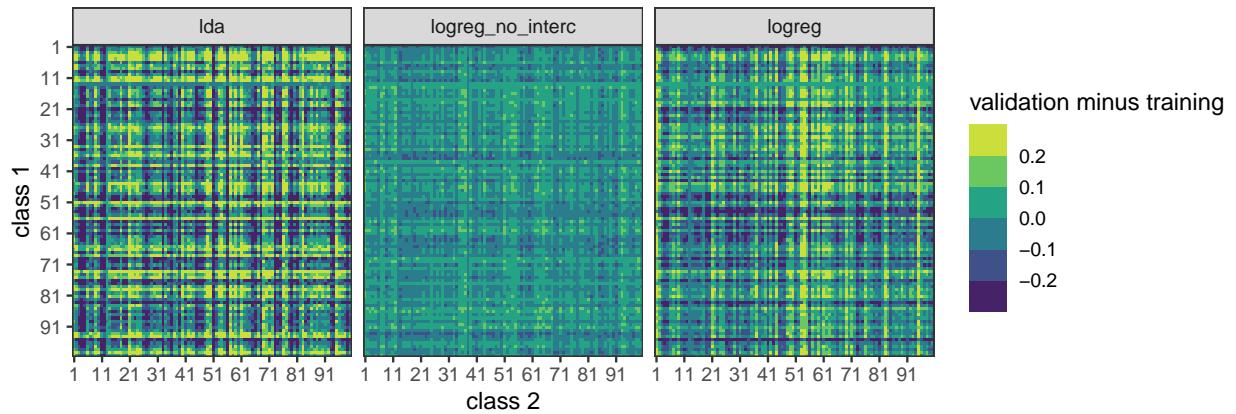
Differences between average pairwise probabilities – class 11



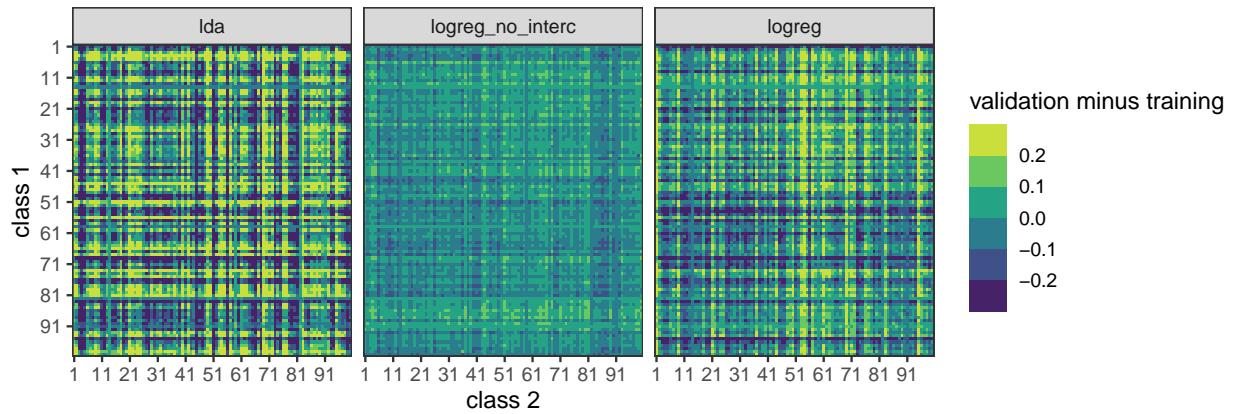
Differences between average pairwise probabilities – class 12



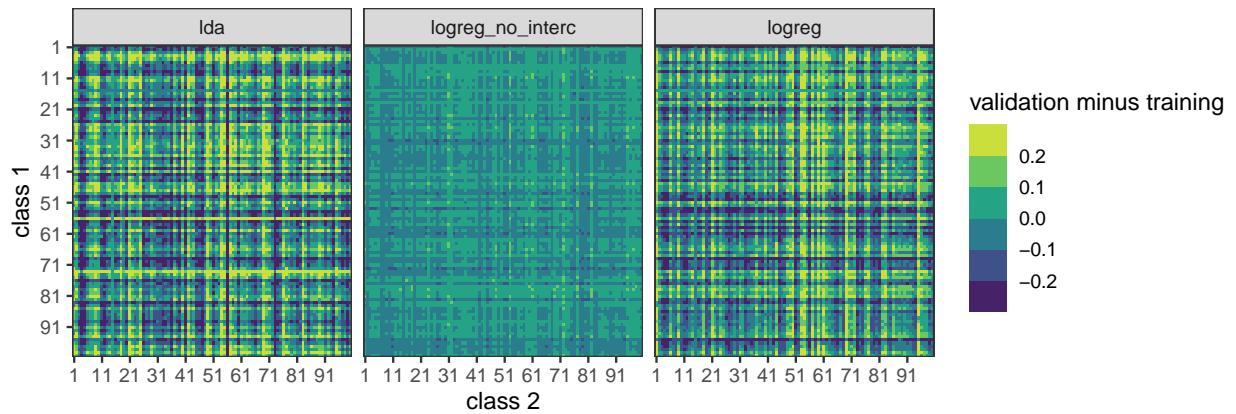
Differences between average pairwise probabilities – class 13



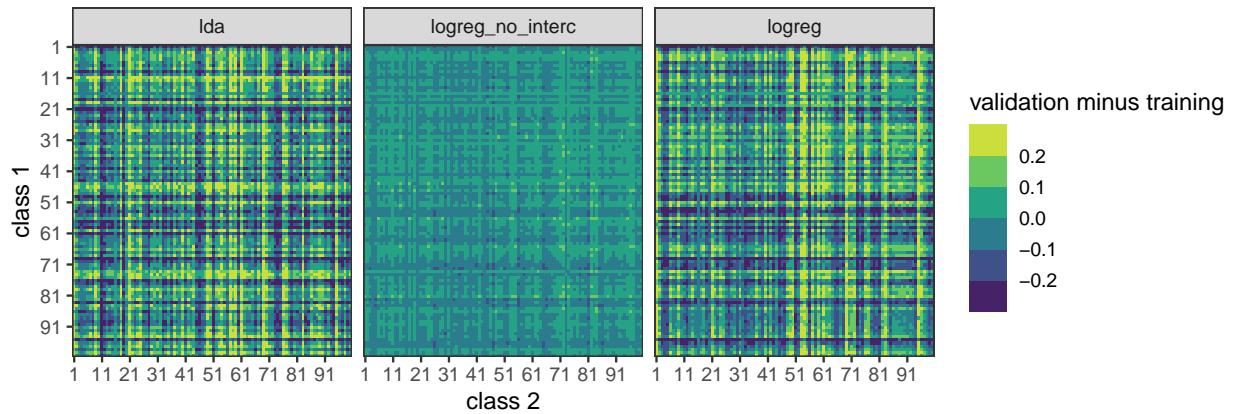
Differences between average pairwise probabilities – class 14



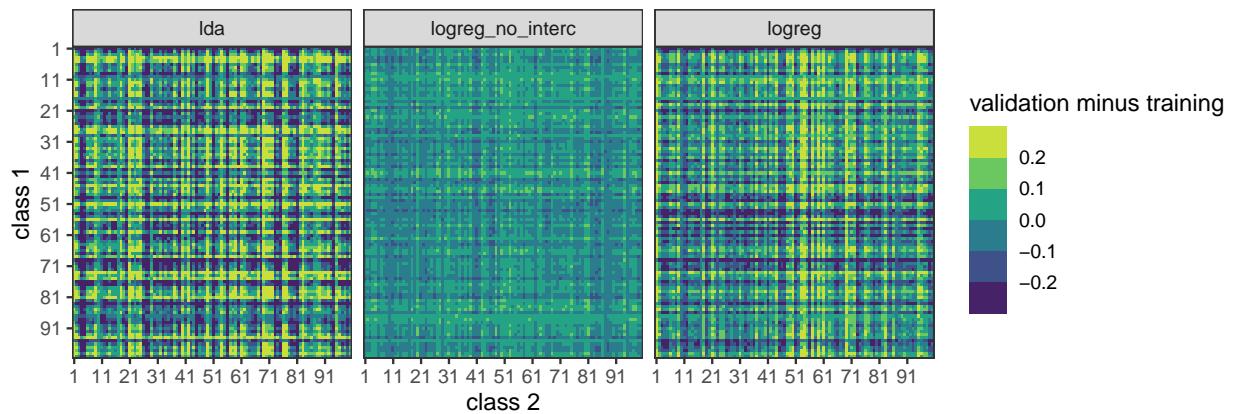
Differences between average pairwise probabilities – class 15



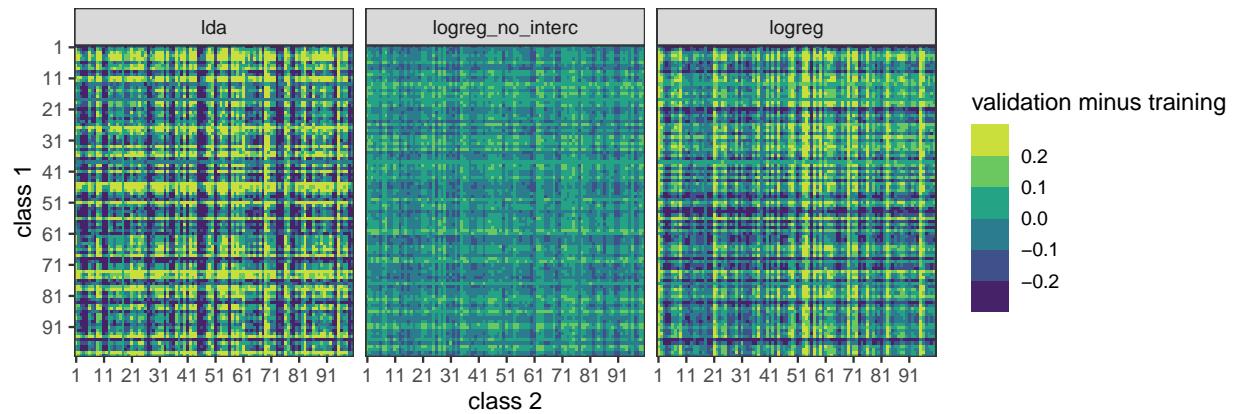
Differences between average pairwise probabilities – class 16



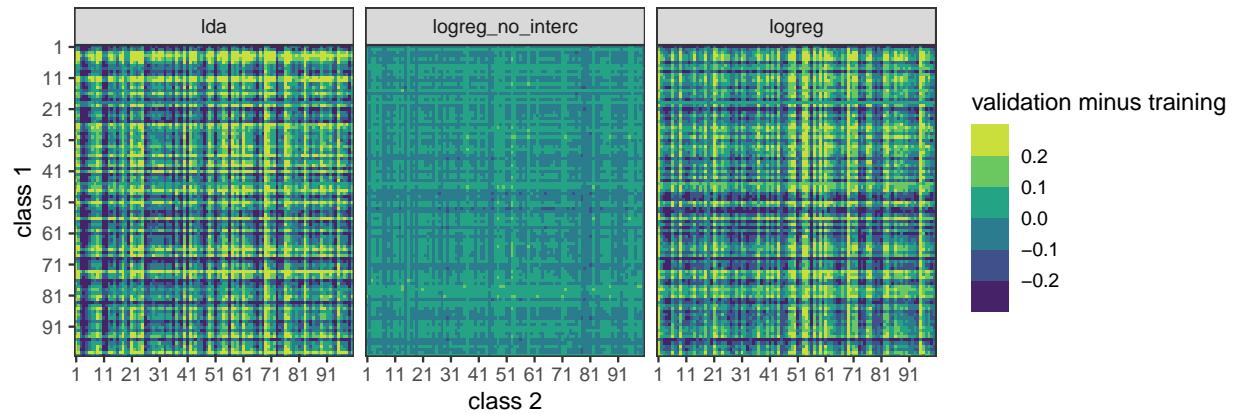
Differences between average pairwise probabilities – class 17



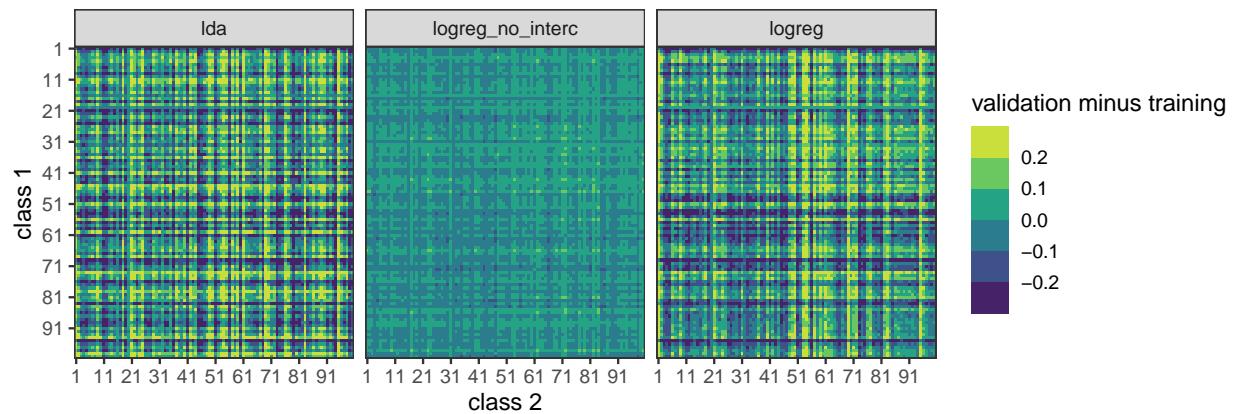
Differences between average pairwise probabilities – class 18



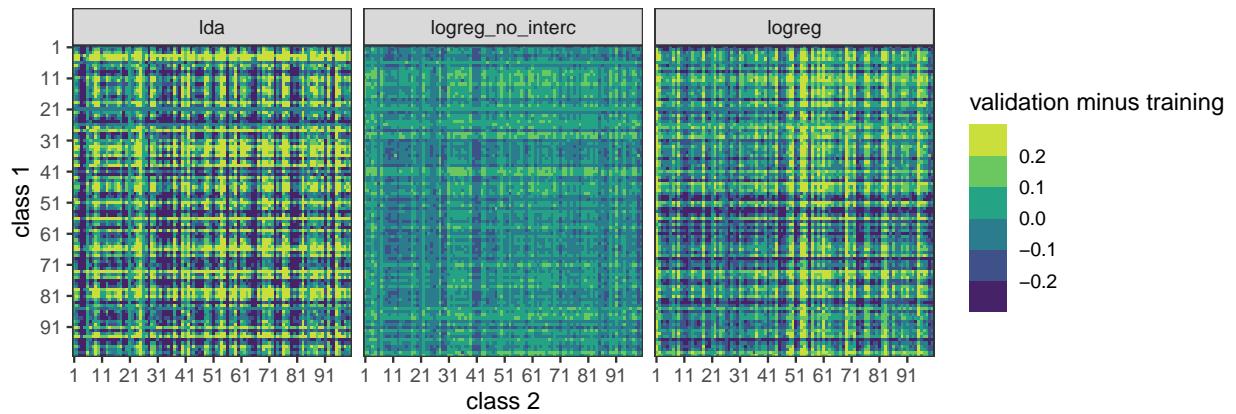
Differences between average pairwise probabilities – class 19



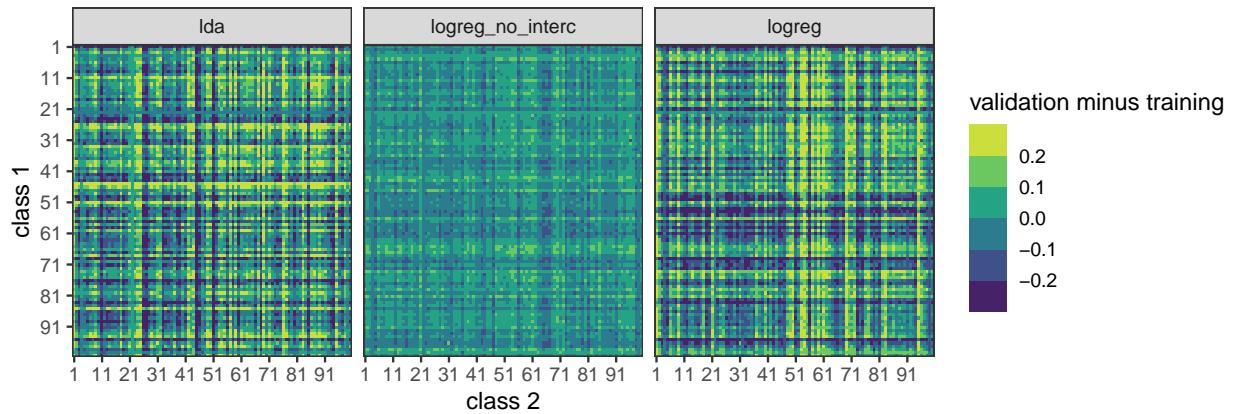
Differences between average pairwise probabilities – class 20



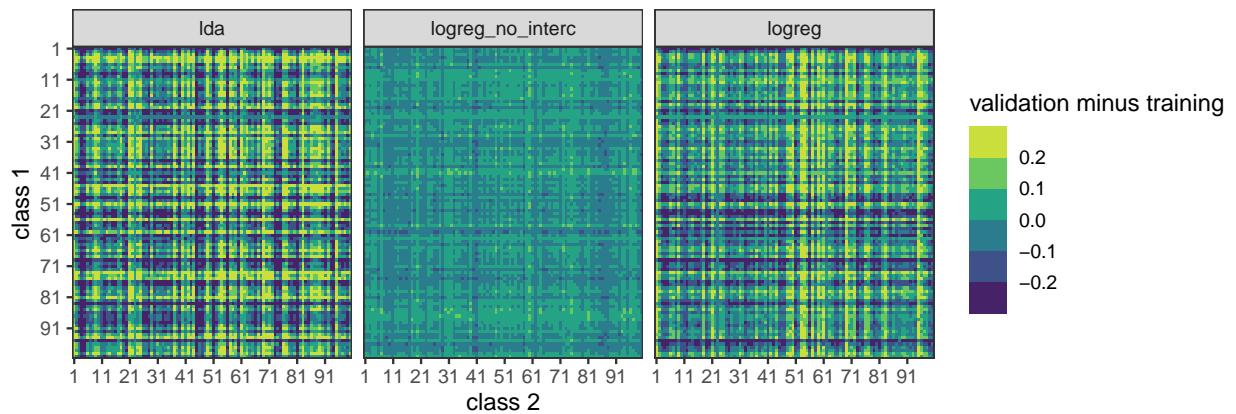
Differences between average pairwise probabilities – class 21



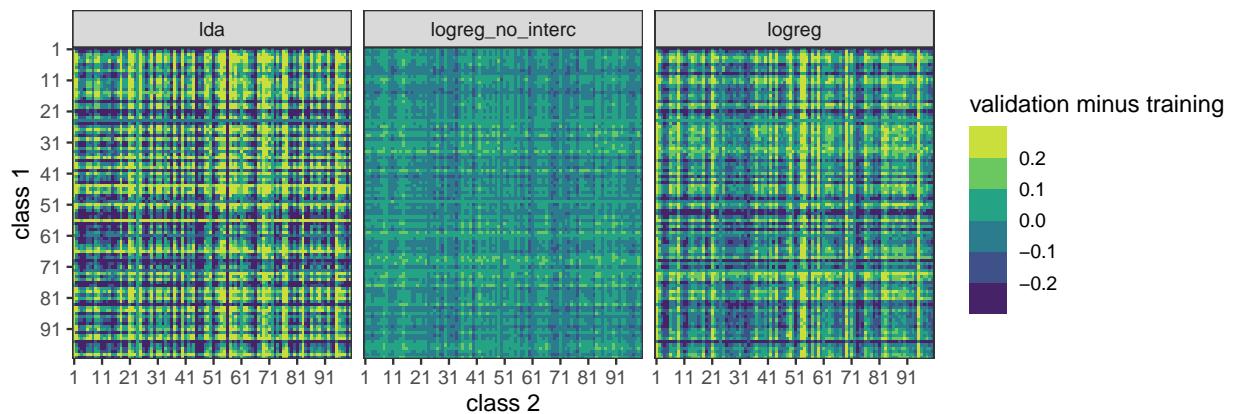
Differences between average pairwise probabilities – class 22



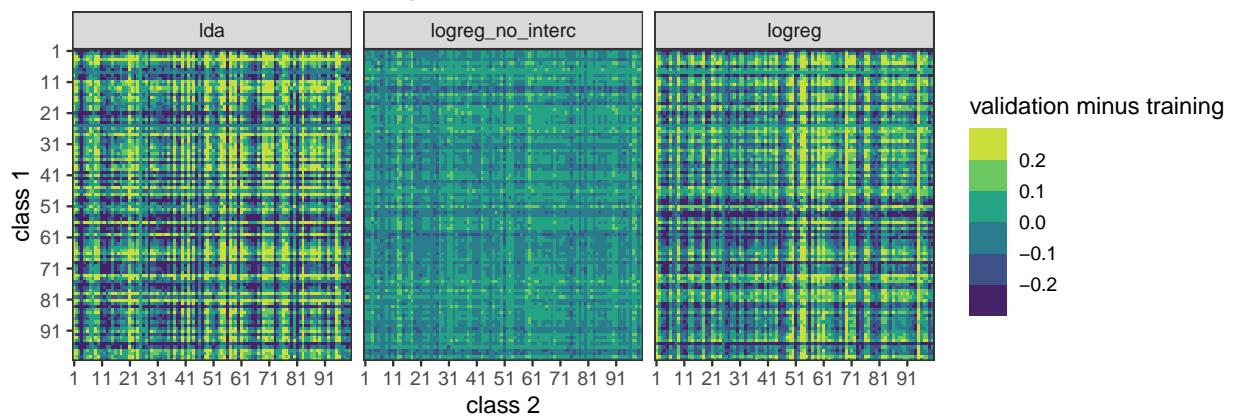
Differences between average pairwise probabilities – class 23



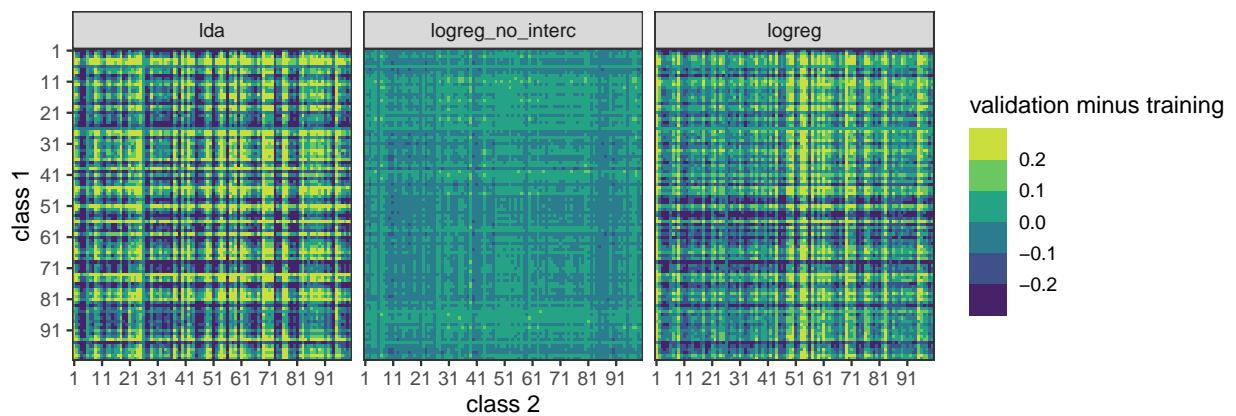
Differences between average pairwise probabilities – class 24



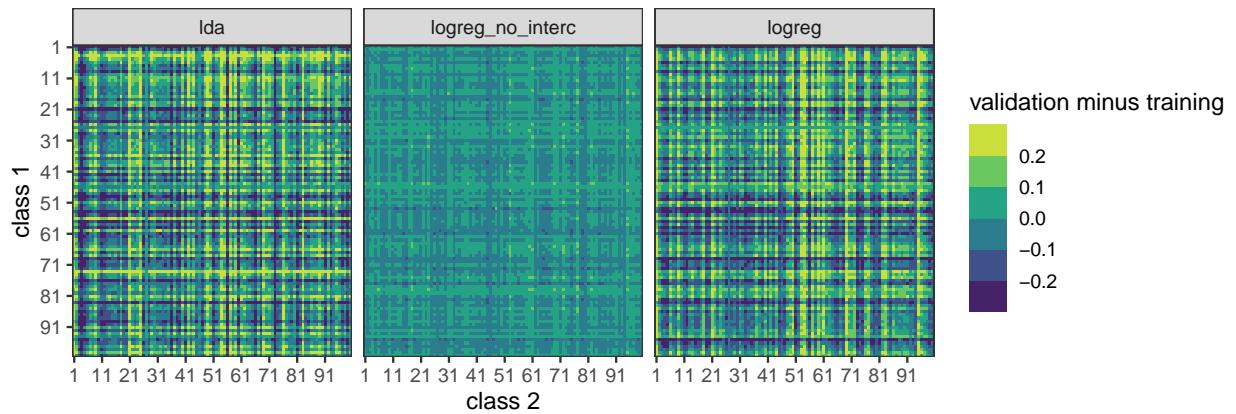
Differences between average pairwise probabilities – class 25



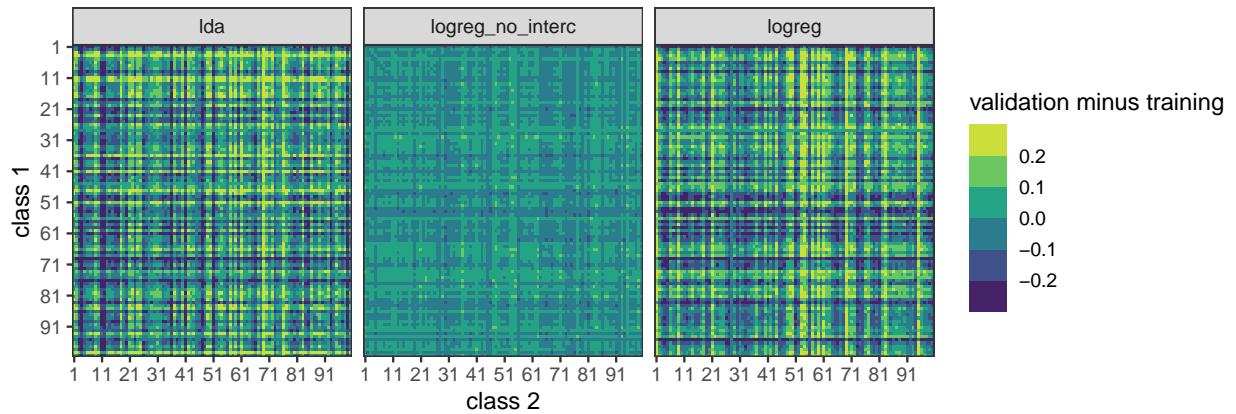
Differences between average pairwise probabilities – class 26



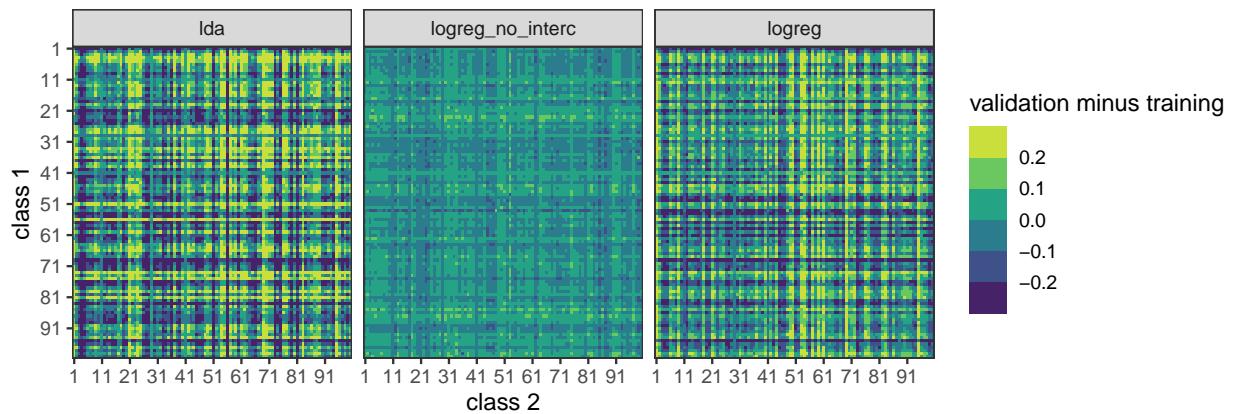
Differences between average pairwise probabilities – class 27



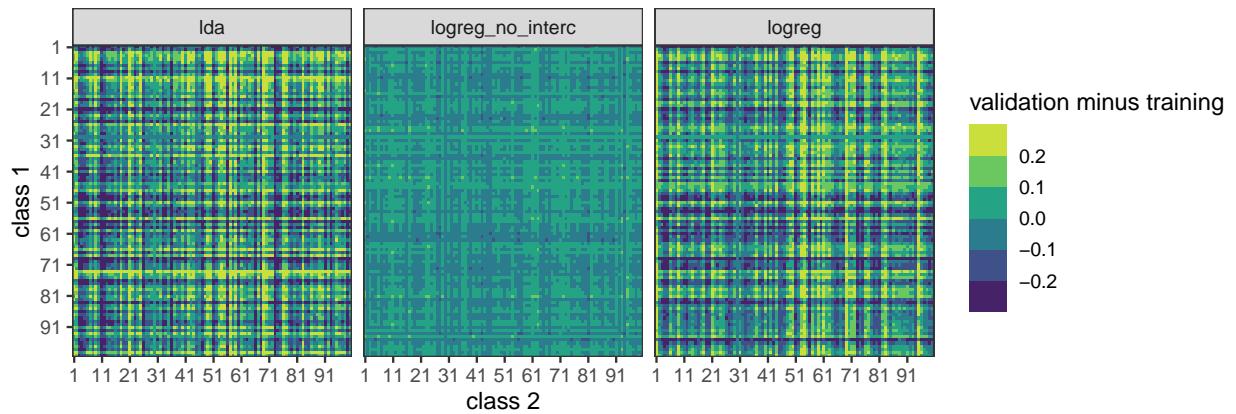
Differences between average pairwise probabilities – class 28



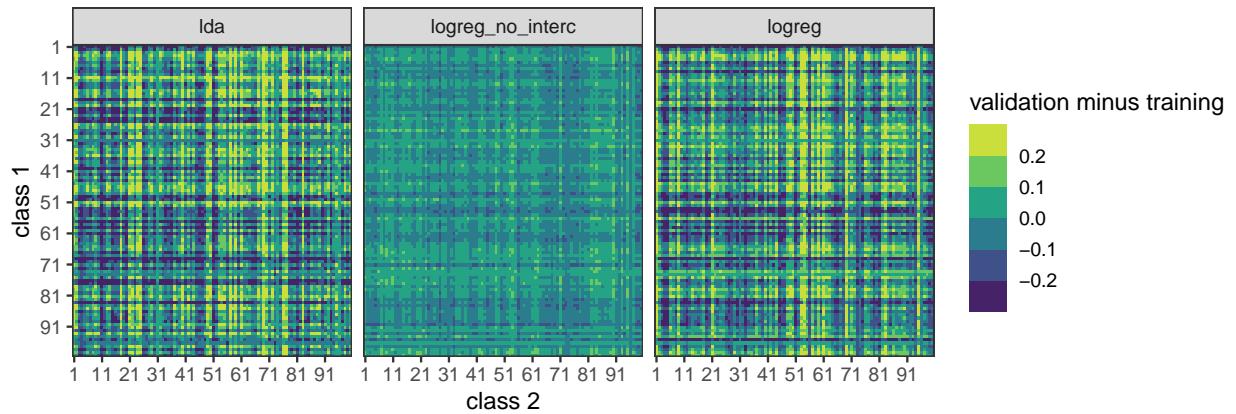
Differences between average pairwise probabilities – class 29



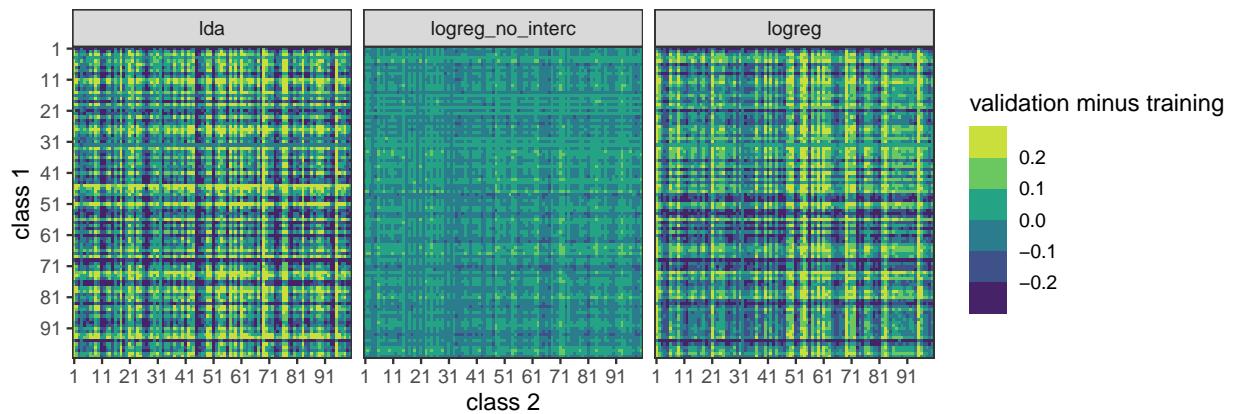
Differences between average pairwise probabilities – class 30



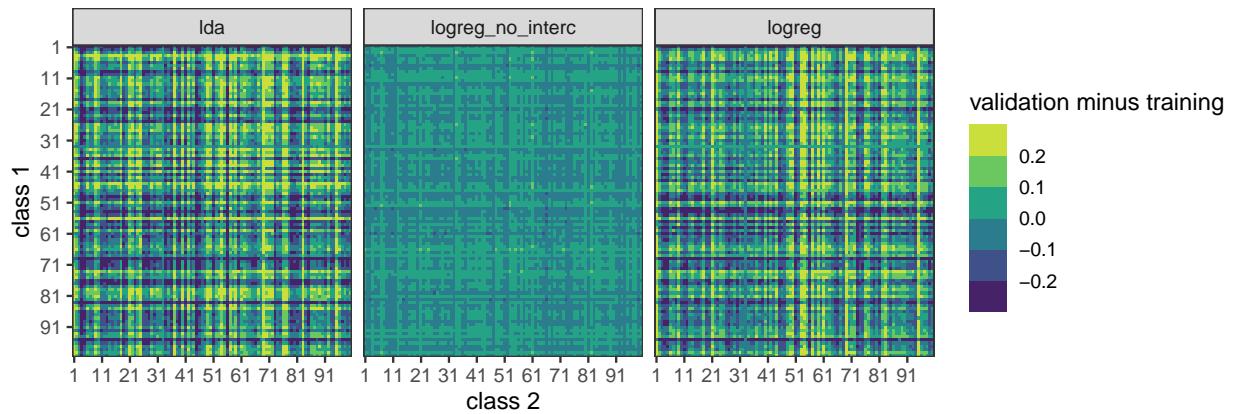
Differences between average pairwise probabilities – class 31



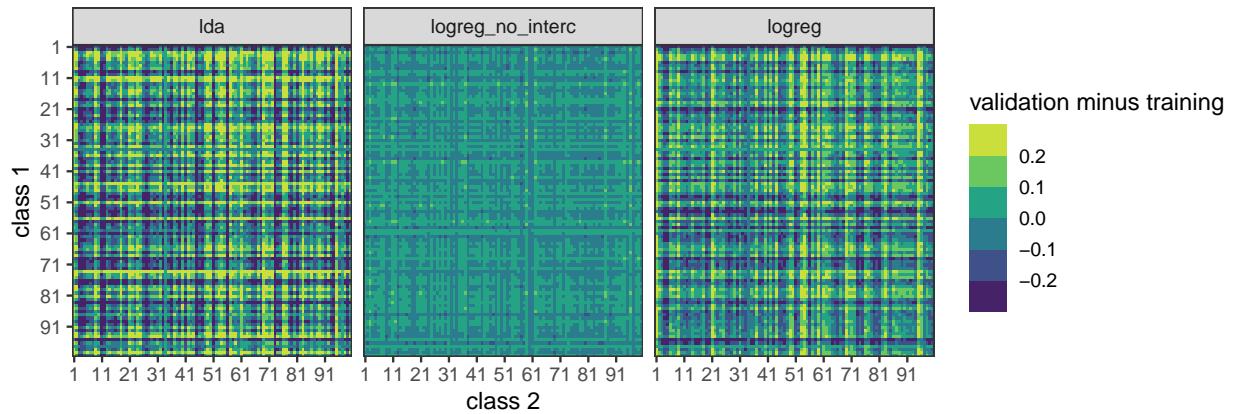
Differences between average pairwise probabilities – class 32



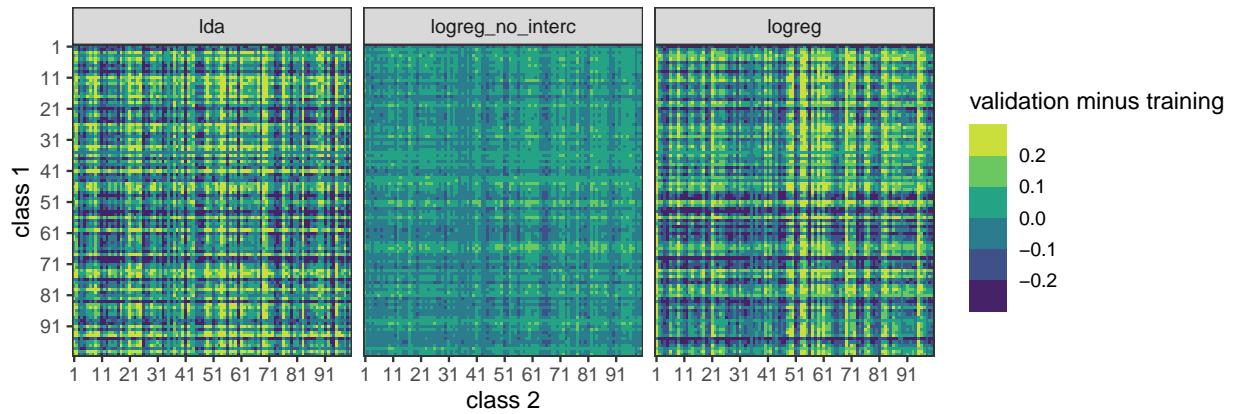
Differences between average pairwise probabilities – class 33



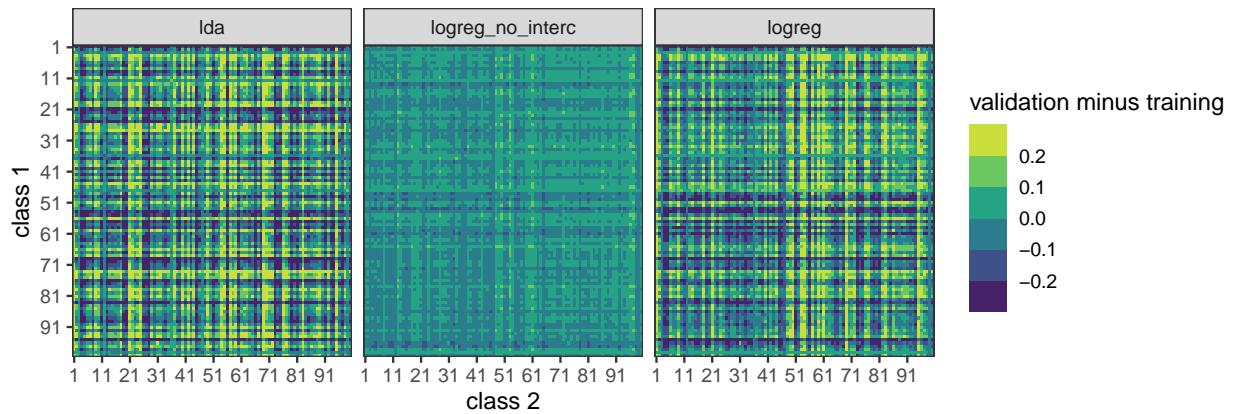
Differences between average pairwise probabilities – class 34



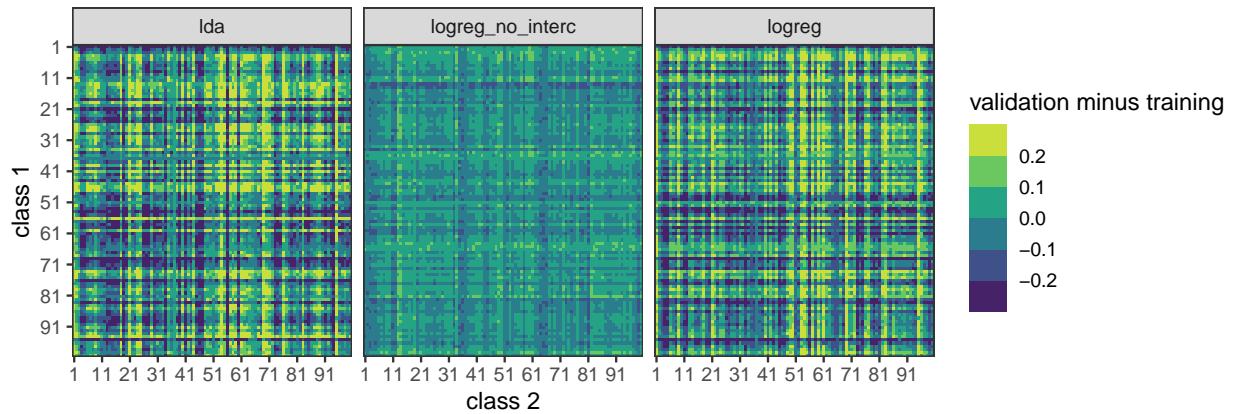
Differences between average pairwise probabilities – class 35



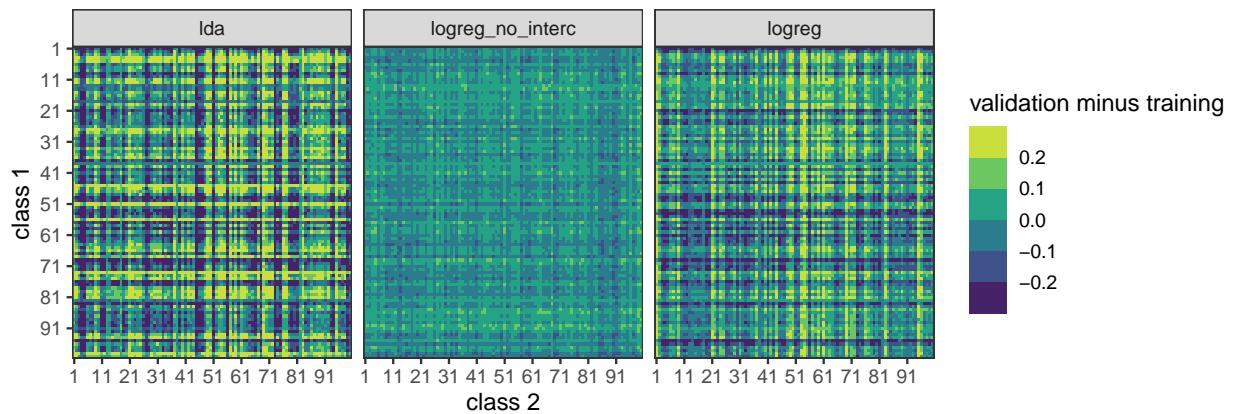
Differences between average pairwise probabilities – class 36



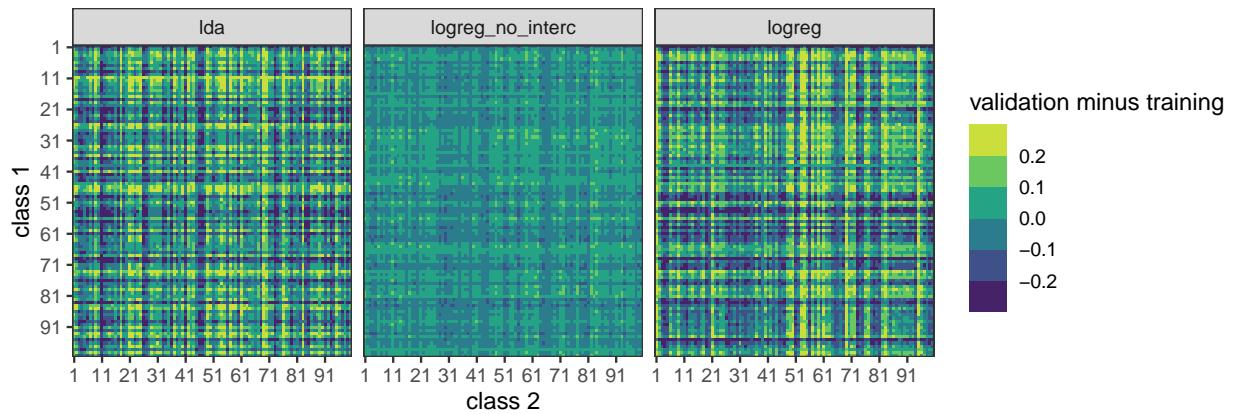
Differences between average pairwise probabilities – class 37



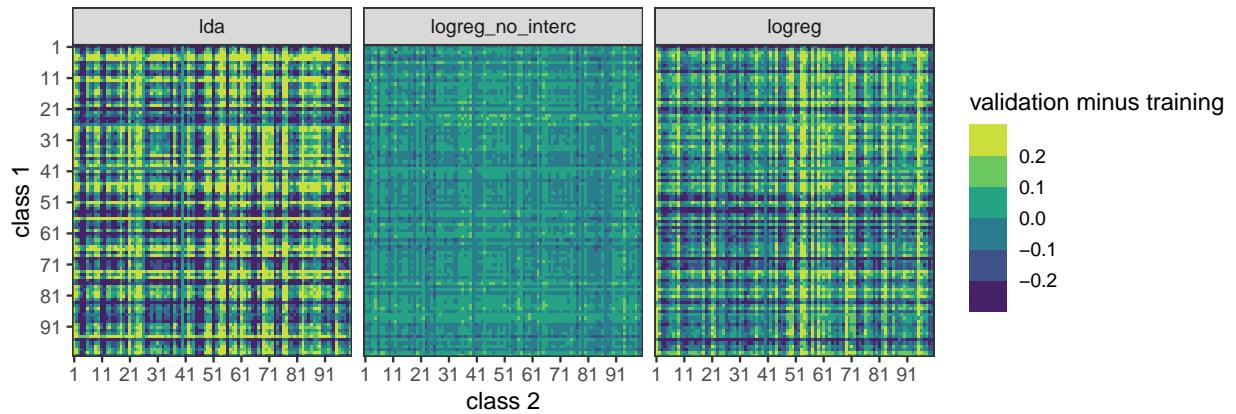
Differences between average pairwise probabilities – class 38



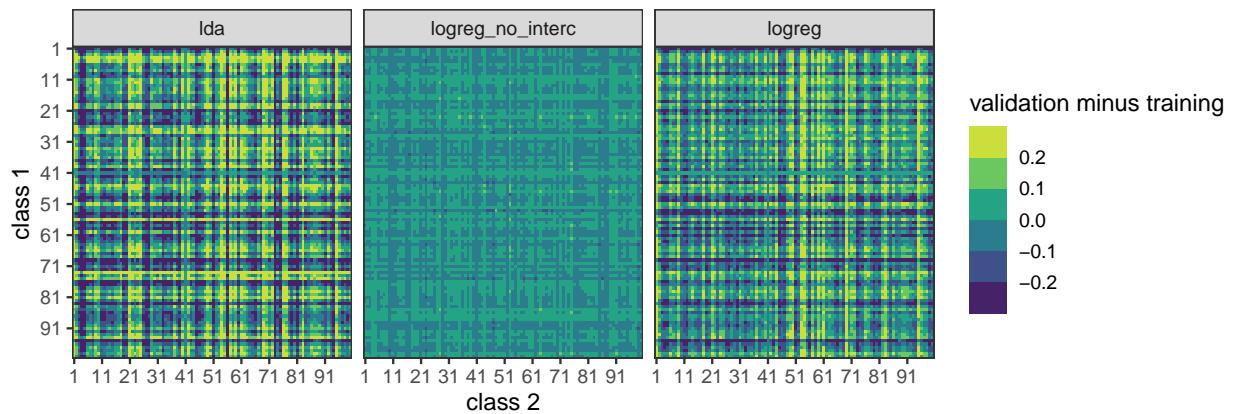
Differences between average pairwise probabilities – class 39



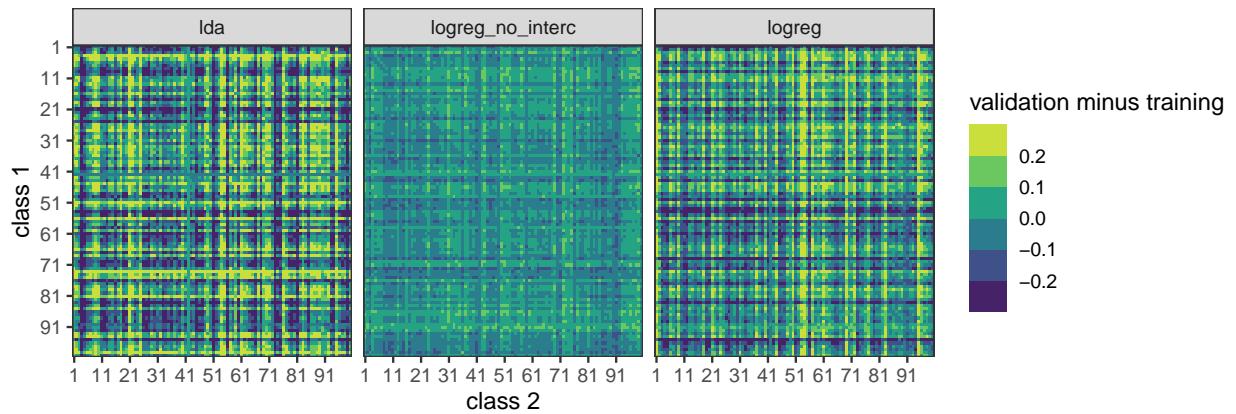
Differences between average pairwise probabilities – class 40



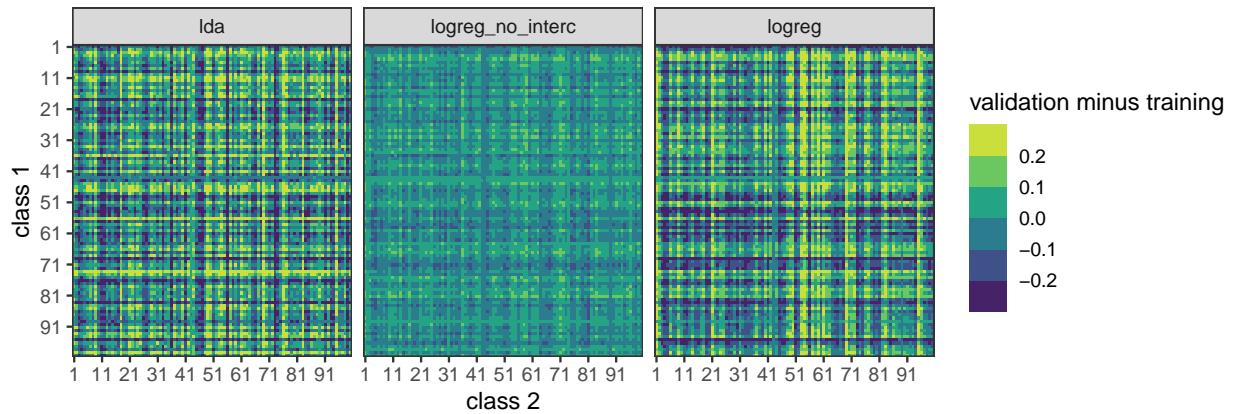
Differences between average pairwise probabilities – class 41



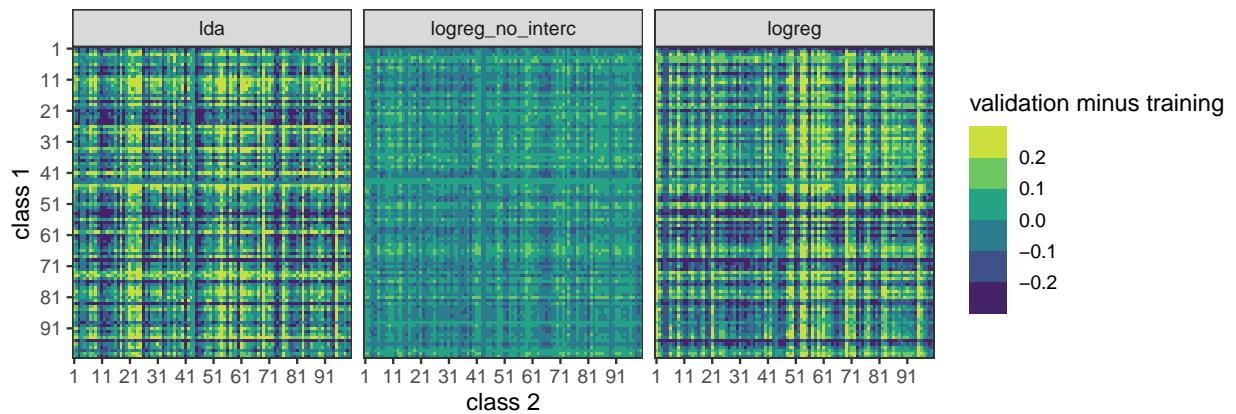
Differences between average pairwise probabilities – class 42



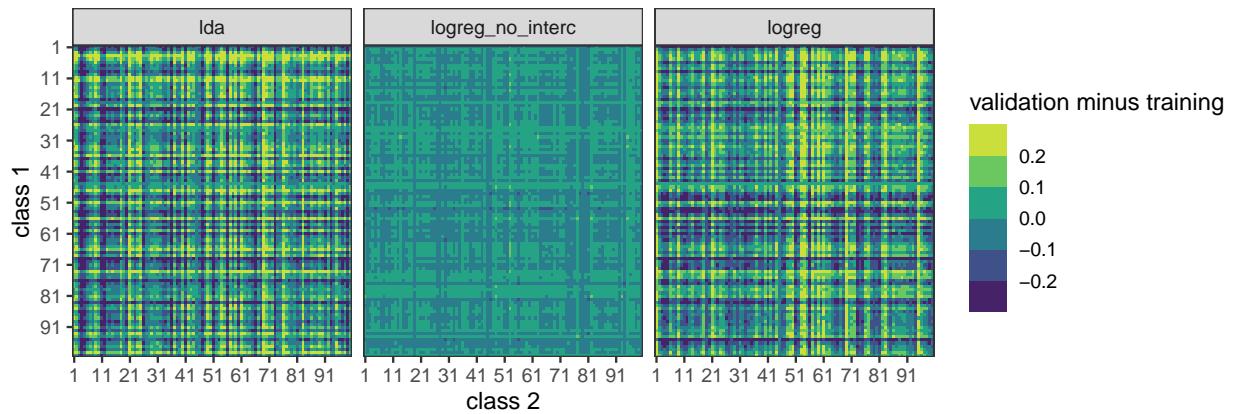
Differences between average pairwise probabilities – class 43



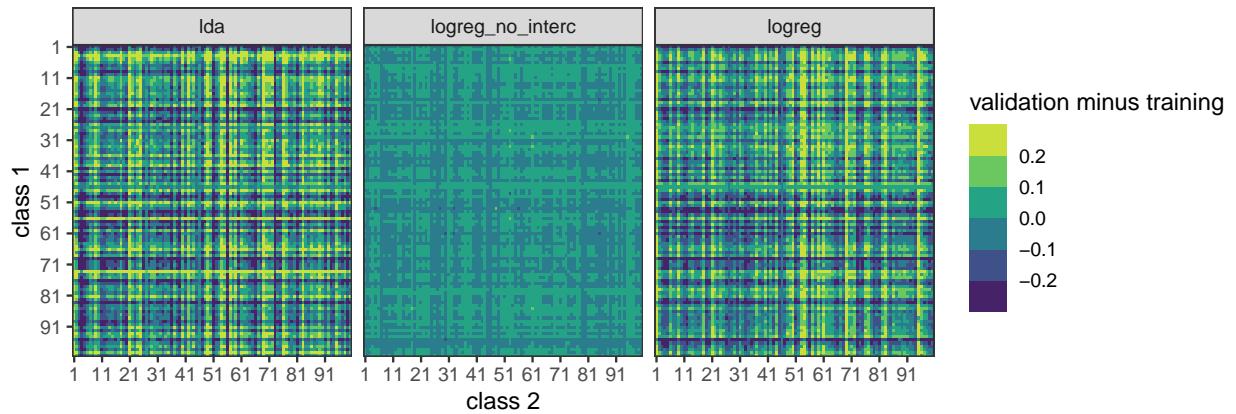
Differences between average pairwise probabilities – class 44



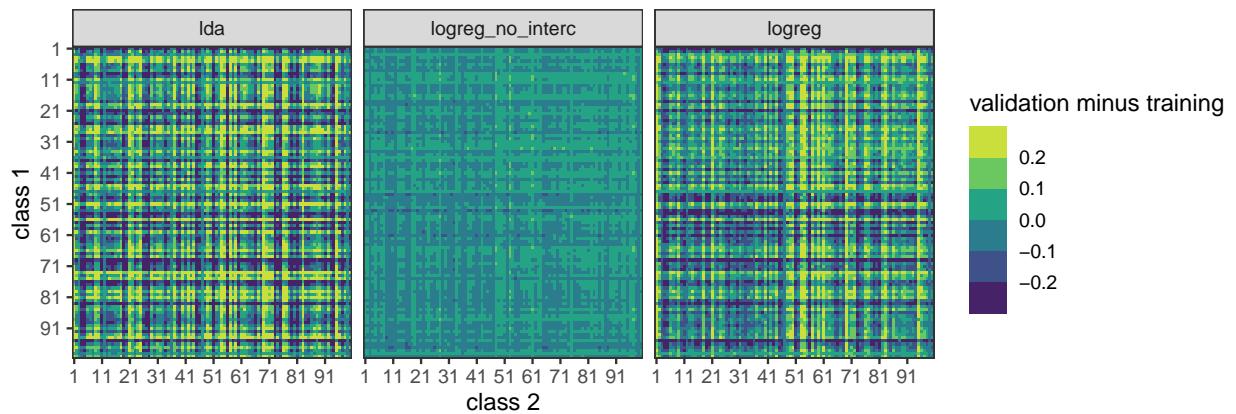
Differences between average pairwise probabilities – class 45



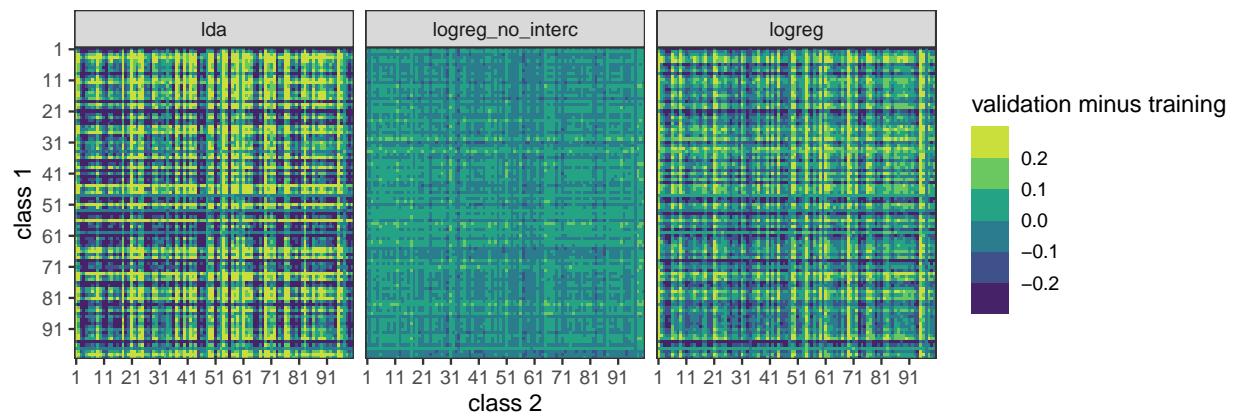
Differences between average pairwise probabilities – class 46



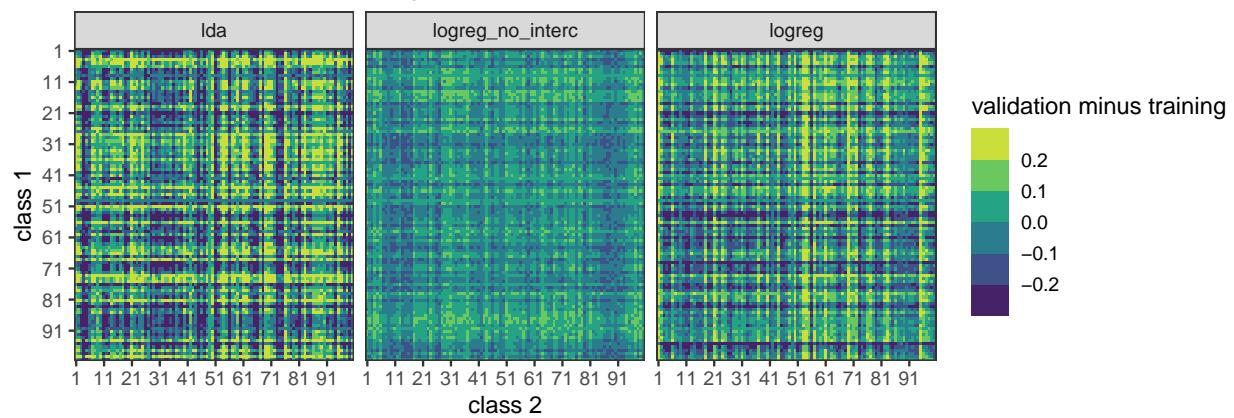
Differences between average pairwise probabilities – class 47



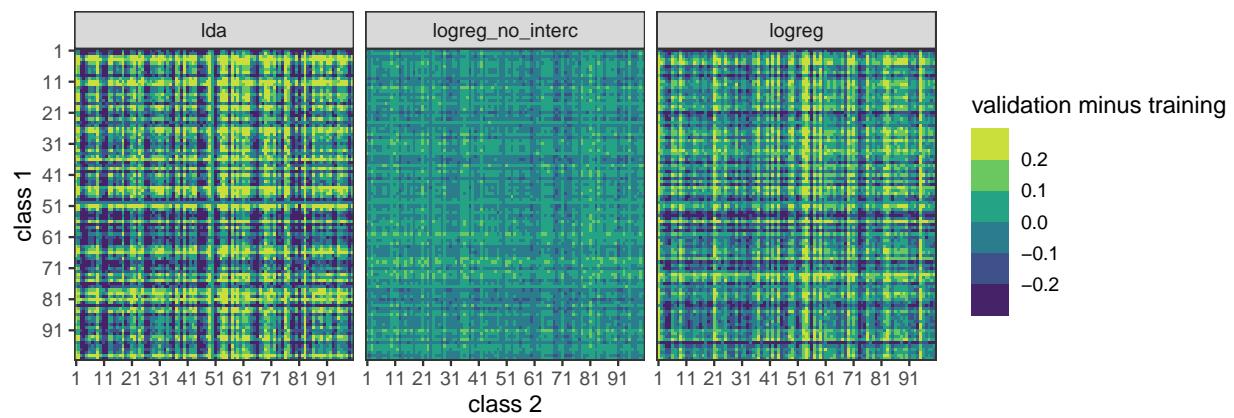
Differences between average pairwise probabilities – class 48



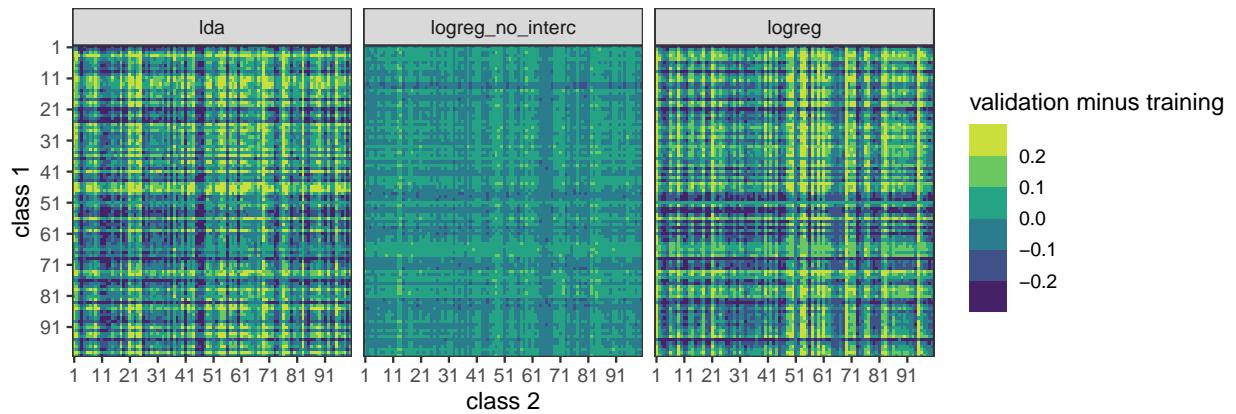
Differences between average pairwise probabilities – class 49



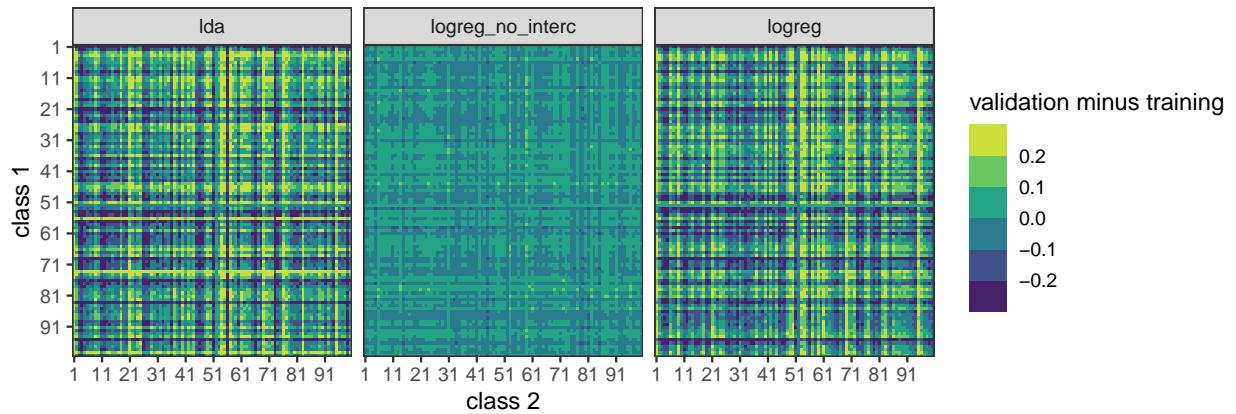
Differences between average pairwise probabilities – class 50



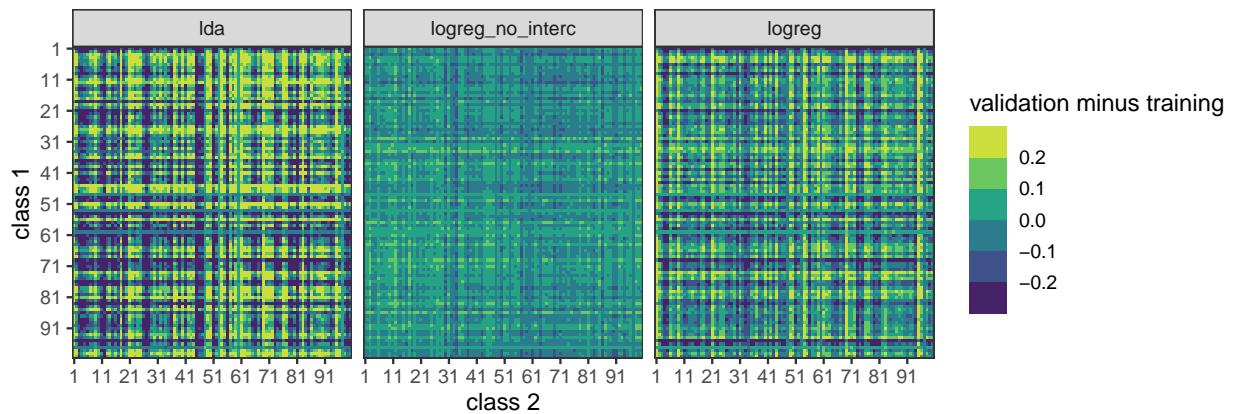
Differences between average pairwise probabilities – class 51



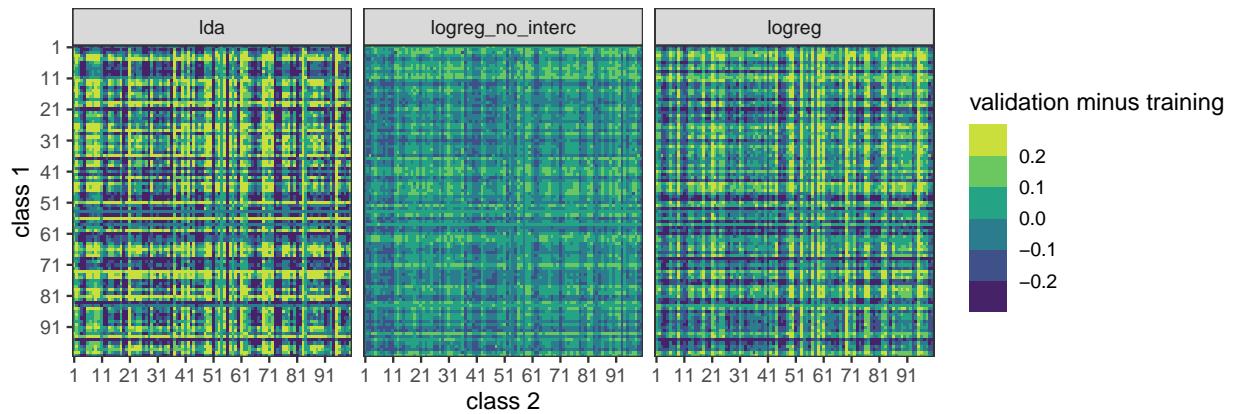
Differences between average pairwise probabilities – class 52



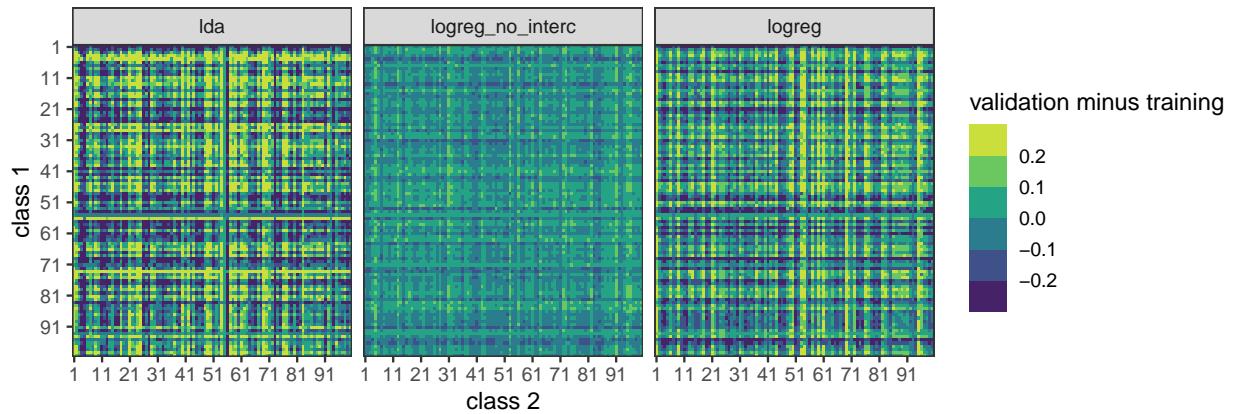
Differences between average pairwise probabilities – class 53



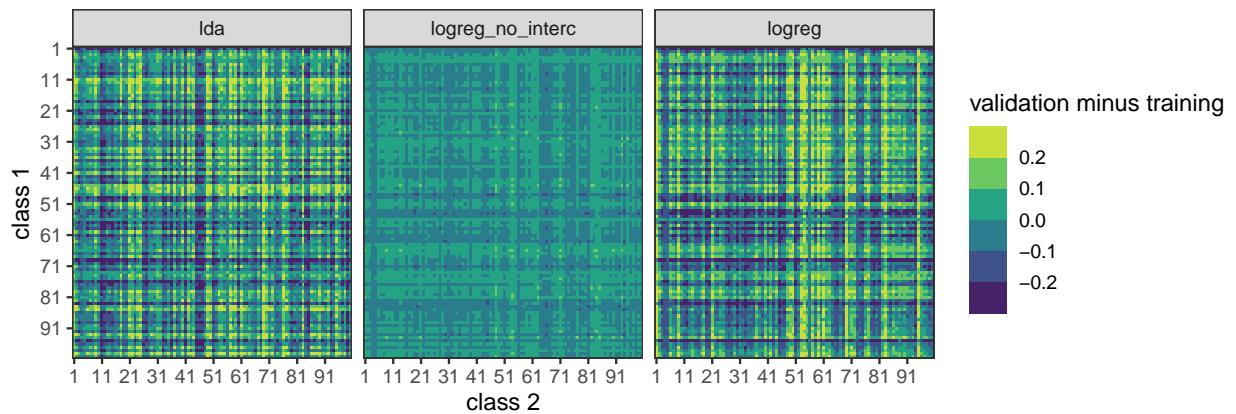
Differences between average pairwise probabilities – class 54



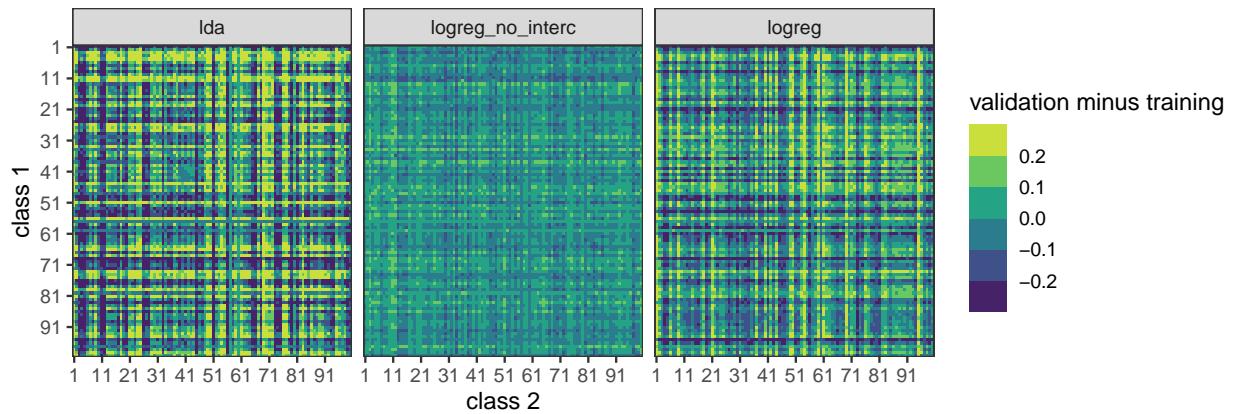
Differences between average pairwise probabilities – class 55



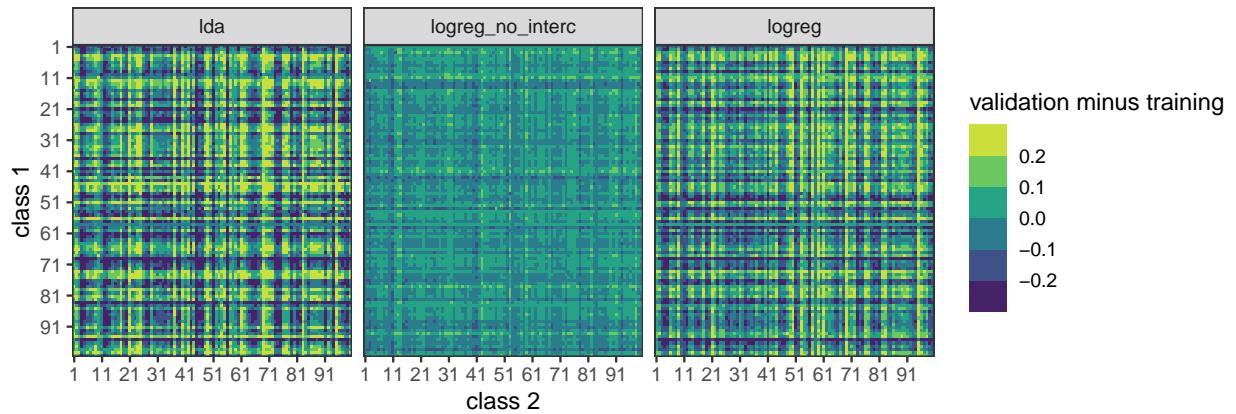
Differences between average pairwise probabilities – class 56



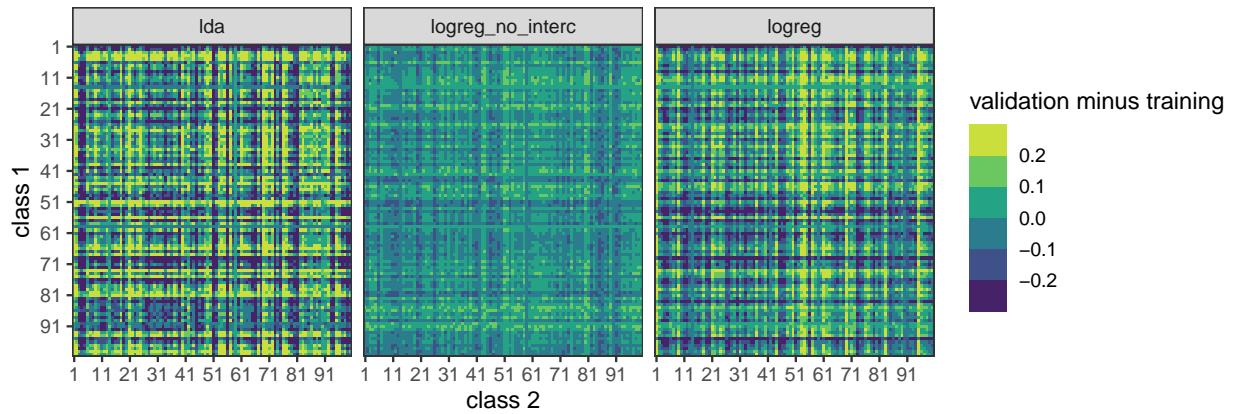
Differences between average pairwise probabilities – class 57



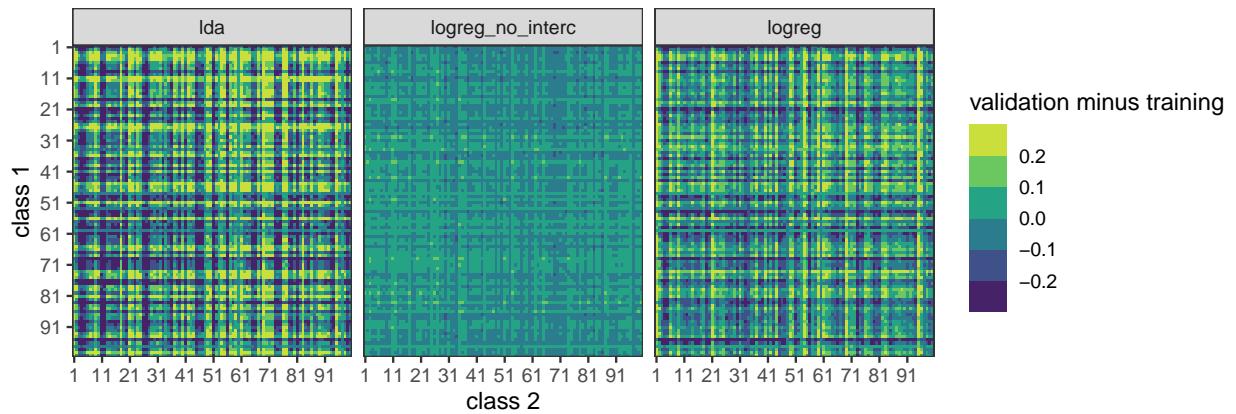
Differences between average pairwise probabilities – class 58



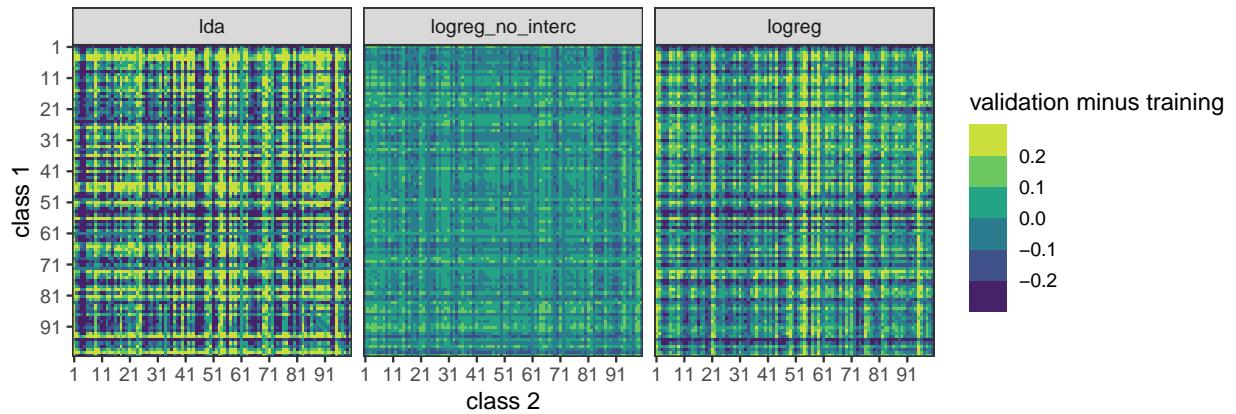
Differences between average pairwise probabilities – class 59



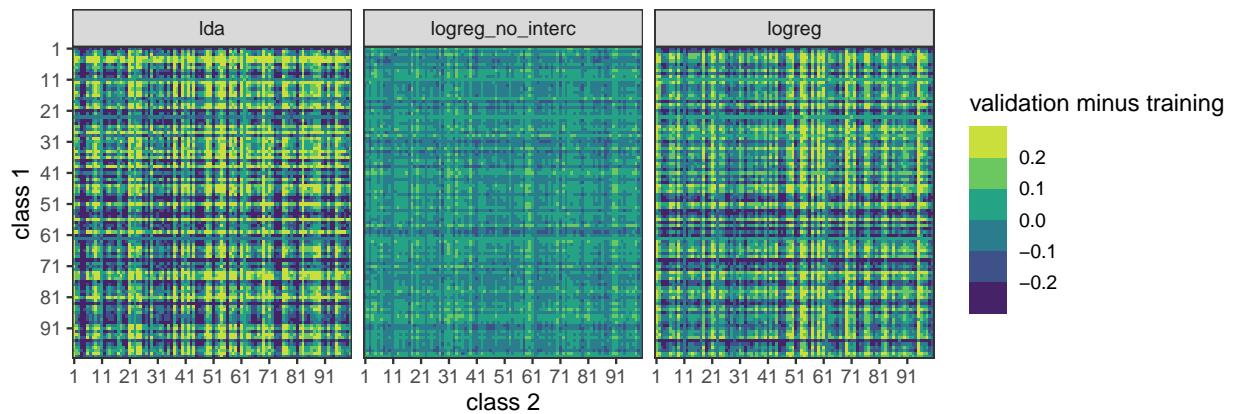
Differences between average pairwise probabilities – class 60



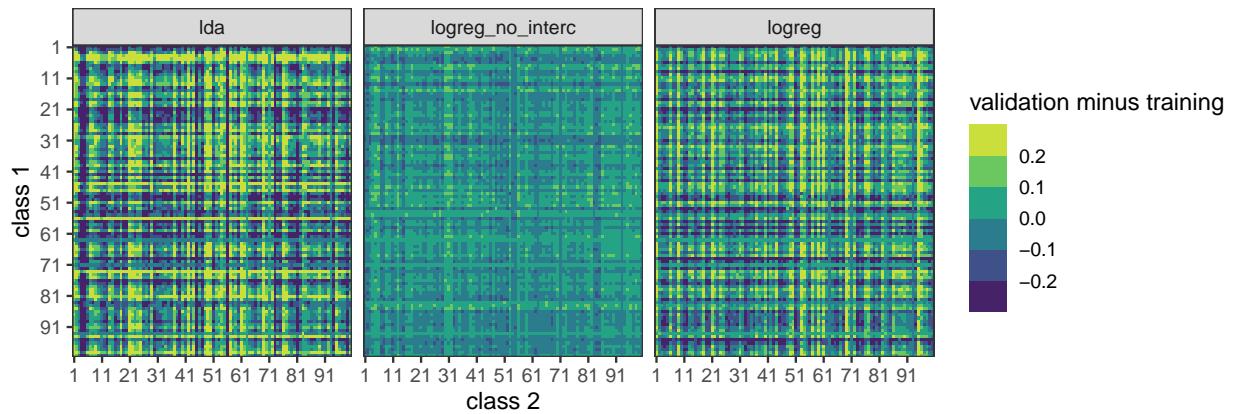
Differences between average pairwise probabilities – class 61



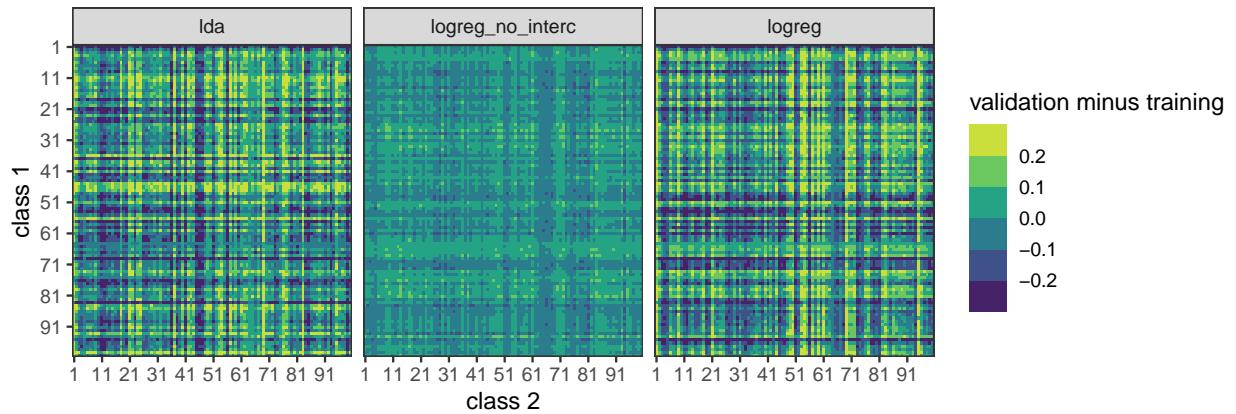
Differences between average pairwise probabilities – class 62



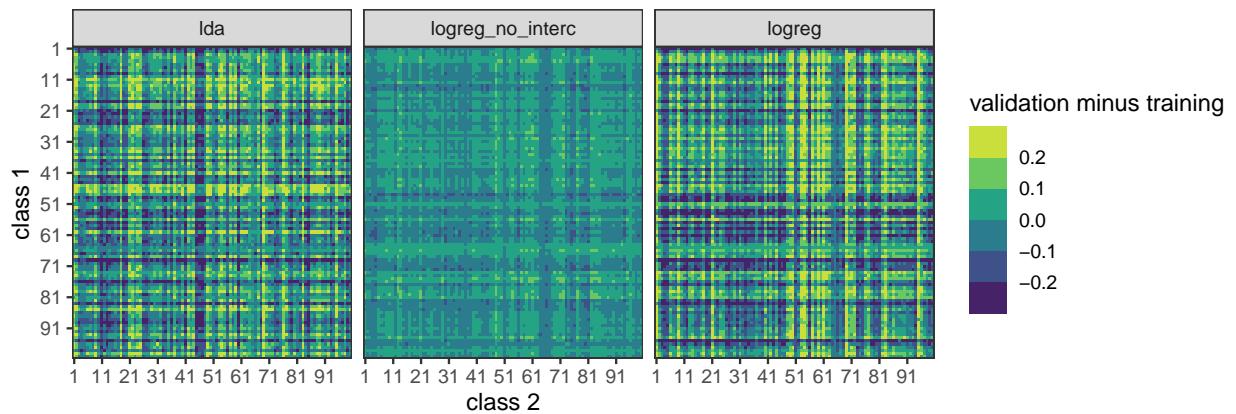
Differences between average pairwise probabilities – class 63



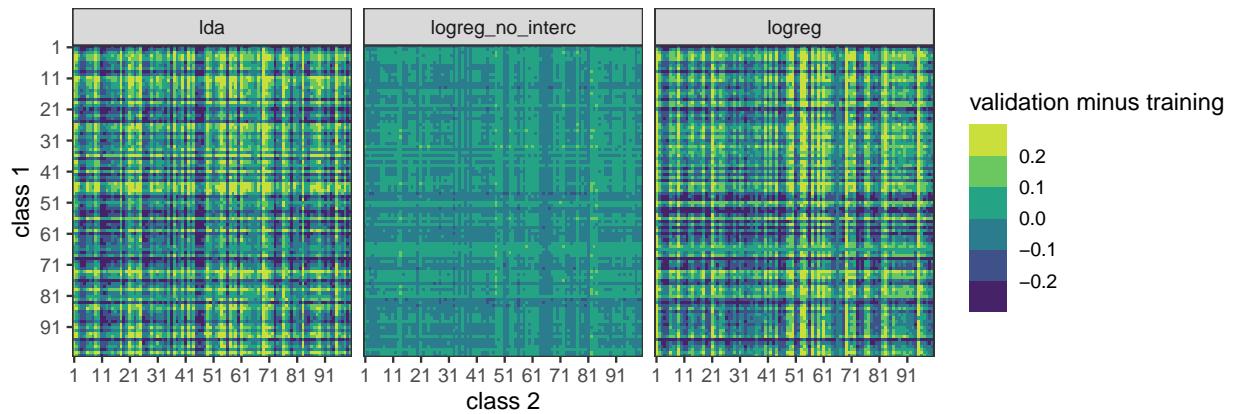
Differences between average pairwise probabilities – class 64



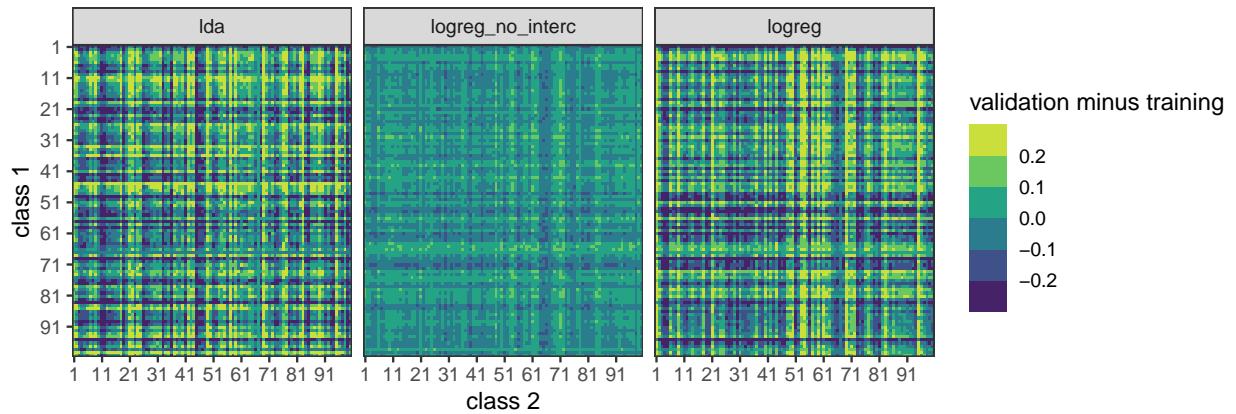
Differences between average pairwise probabilities – class 65



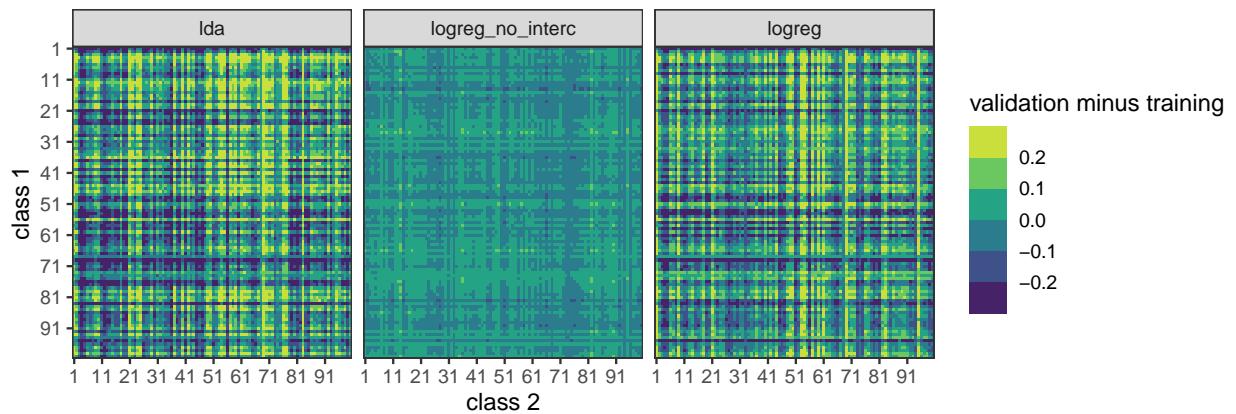
Differences between average pairwise probabilities – class 66



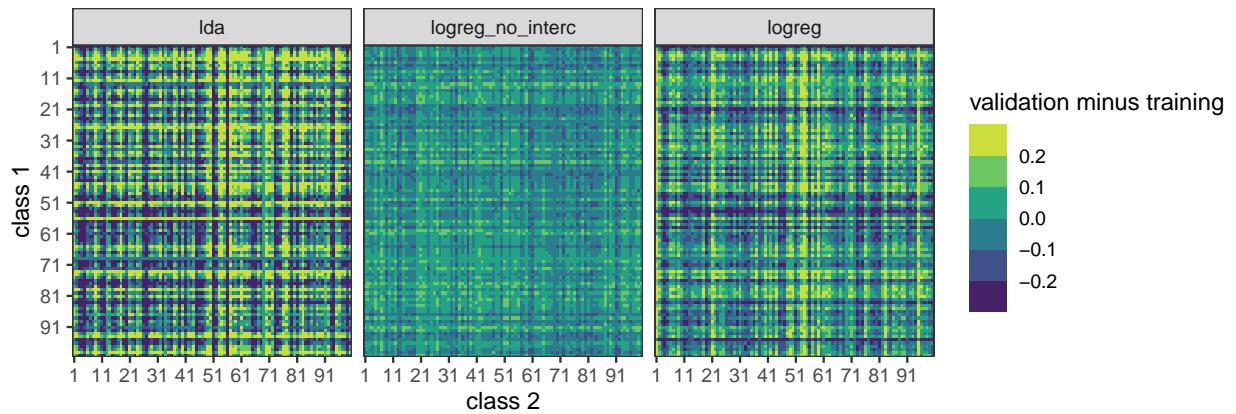
Differences between average pairwise probabilities – class 67



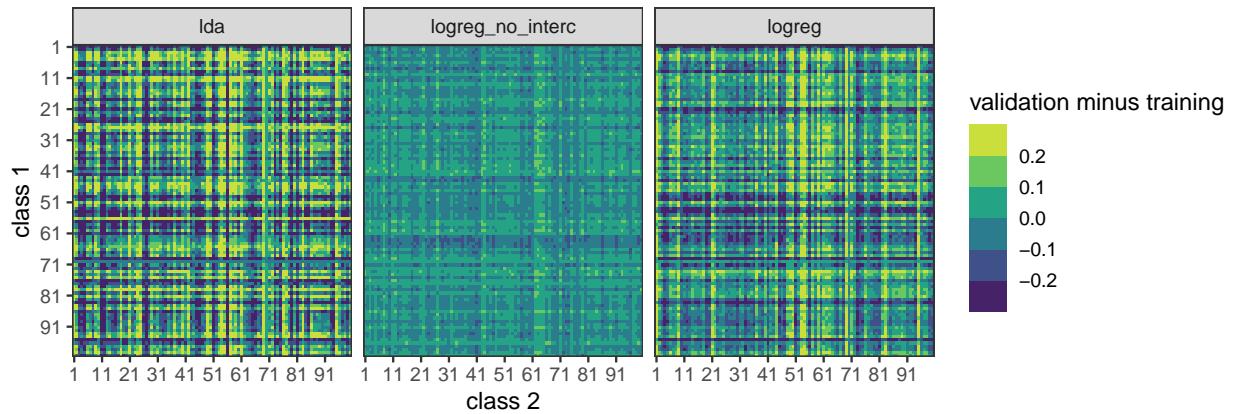
Differences between average pairwise probabilities – class 68



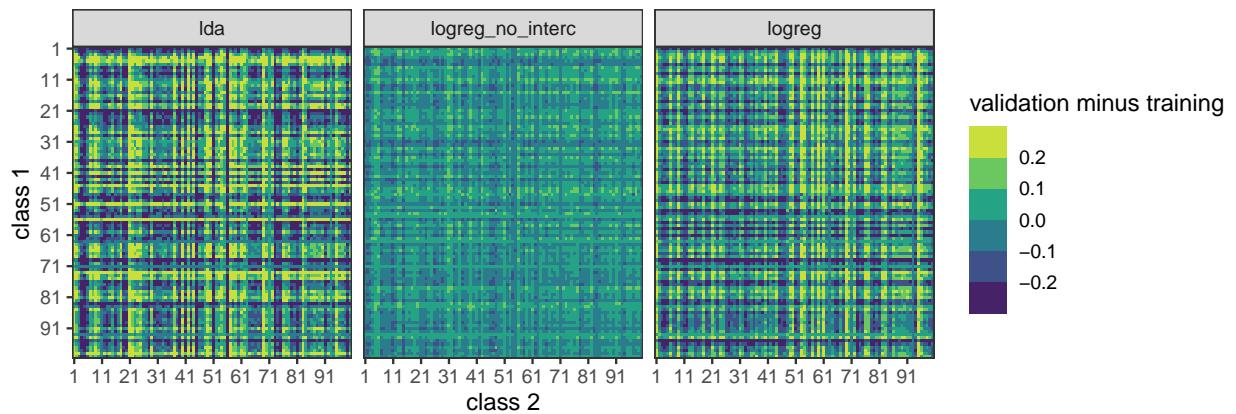
Differences between average pairwise probabilities – class 69



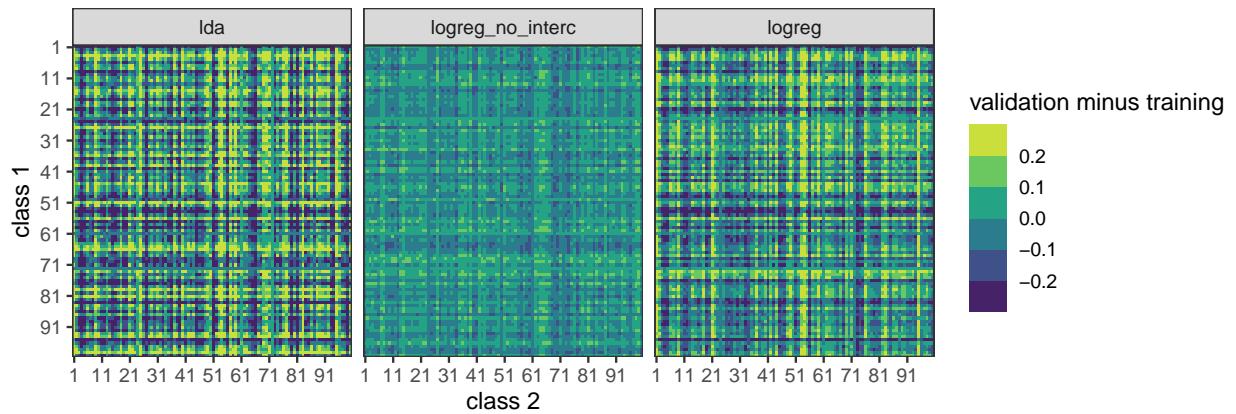
Differences between average pairwise probabilities – class 70



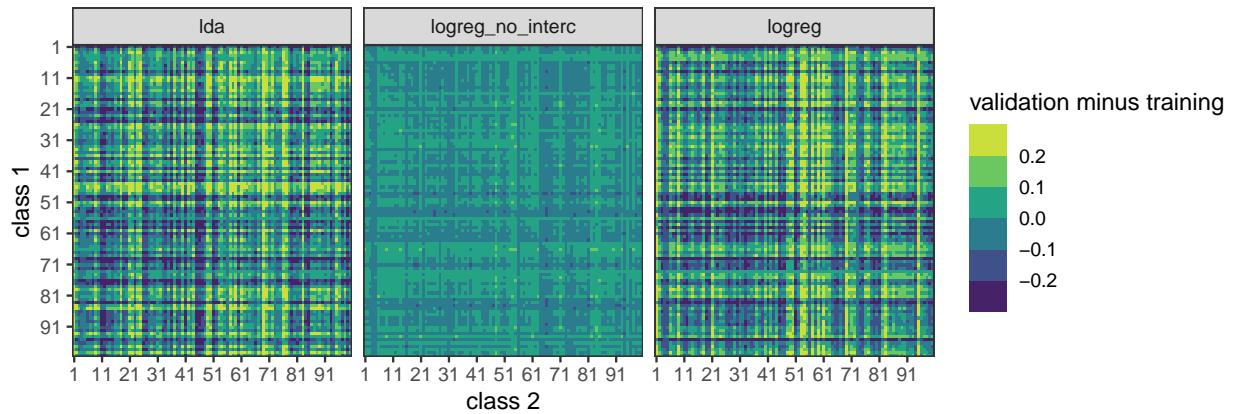
Differences between average pairwise probabilities – class 71



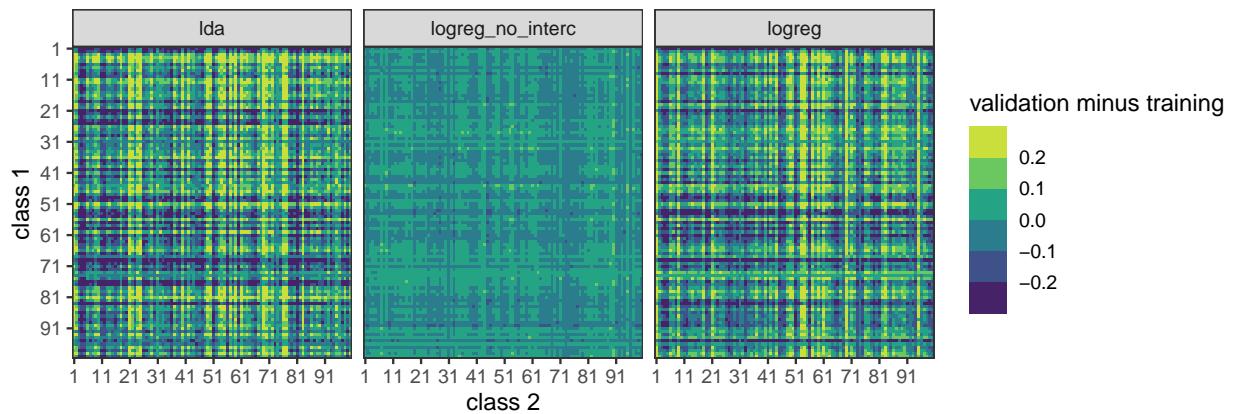
Differences between average pairwise probabilities – class 72



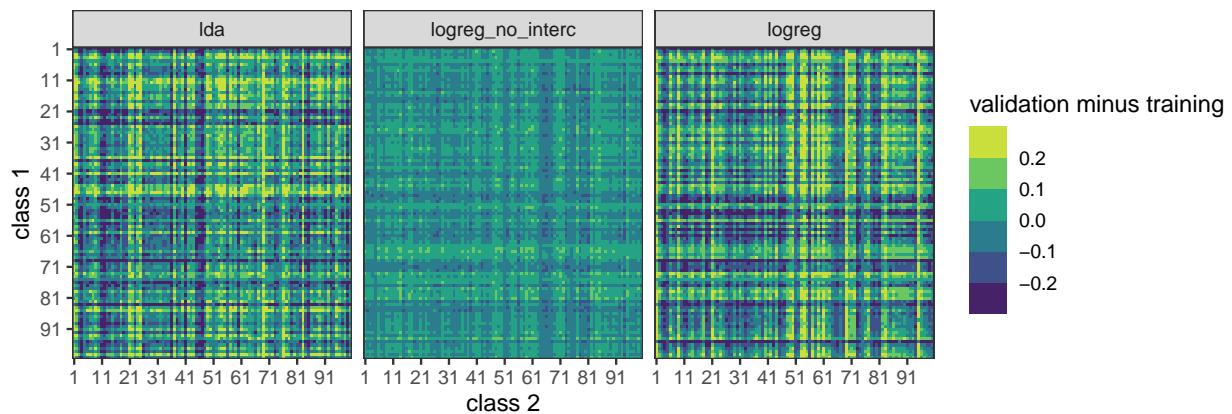
Differences between average pairwise probabilities – class 73



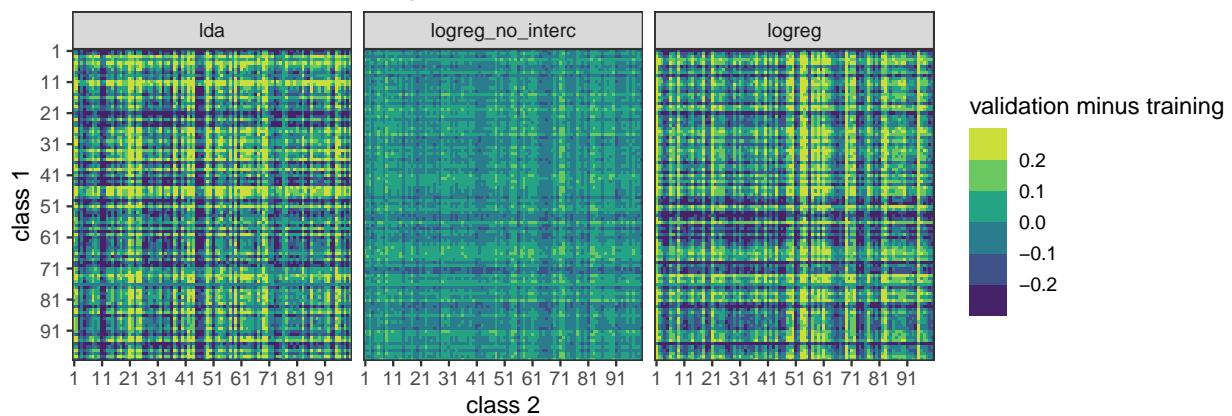
Differences between average pairwise probabilities – class 74



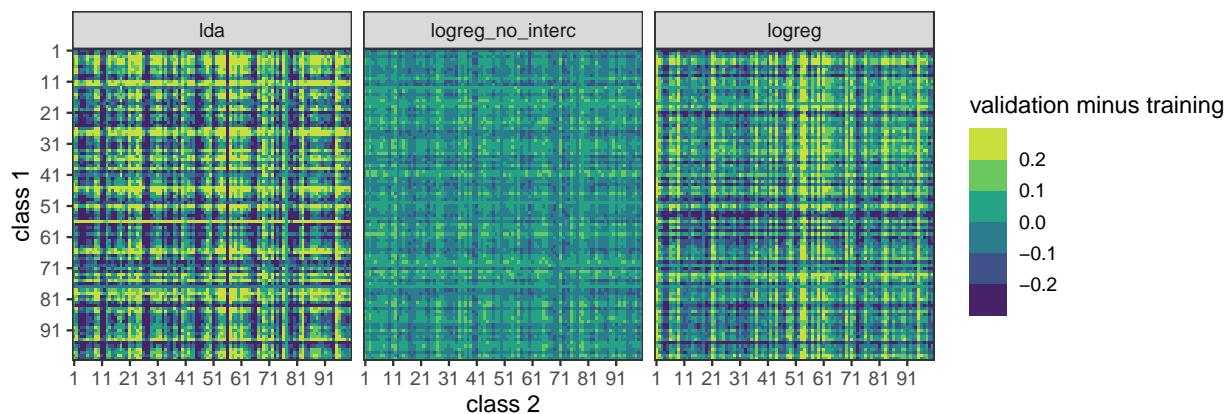
Differences between average pairwise probabilities – class 75



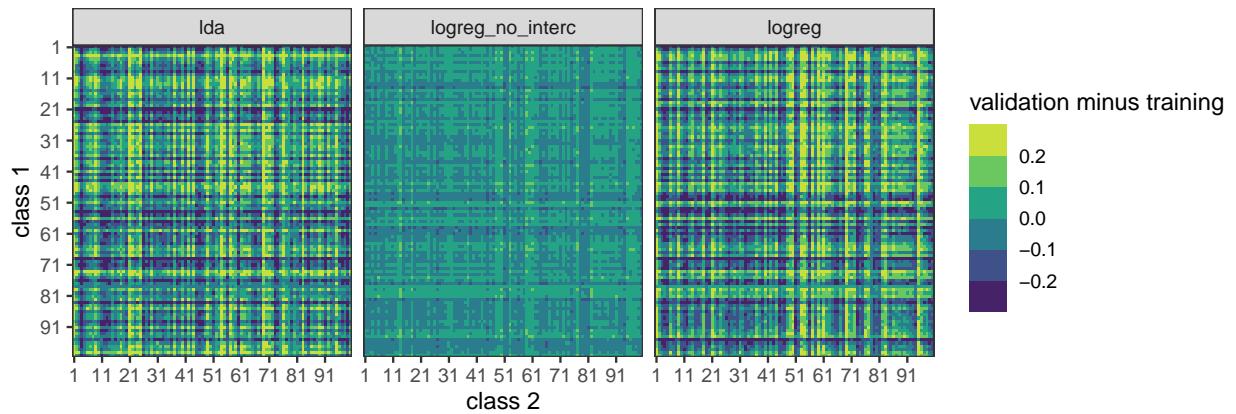
Differences between average pairwise probabilities – class 76



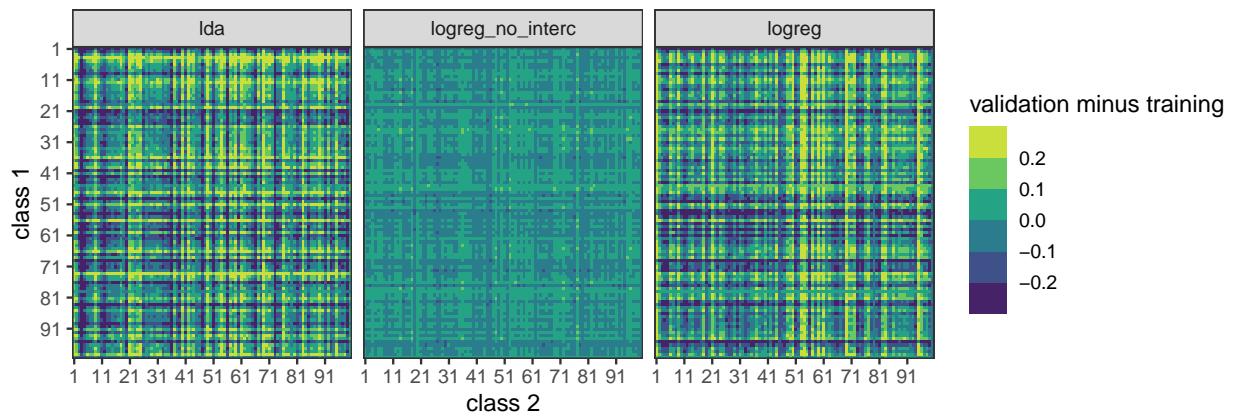
Differences between average pairwise probabilities – class 77



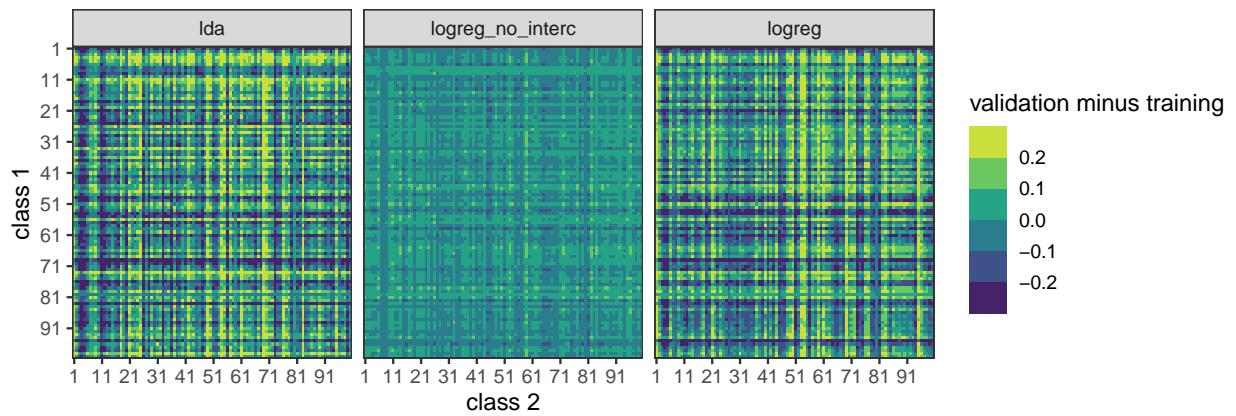
Differences between average pairwise probabilities – class 78



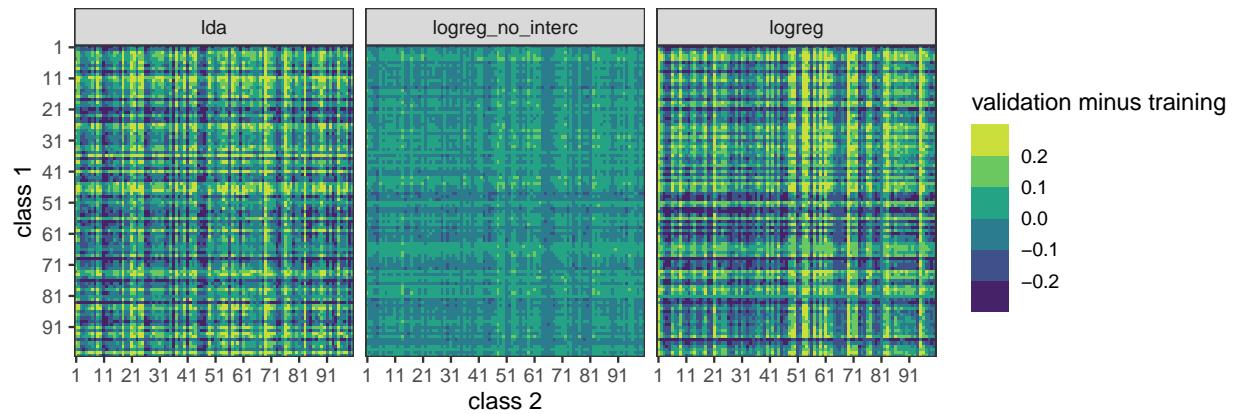
Differences between average pairwise probabilities – class 79



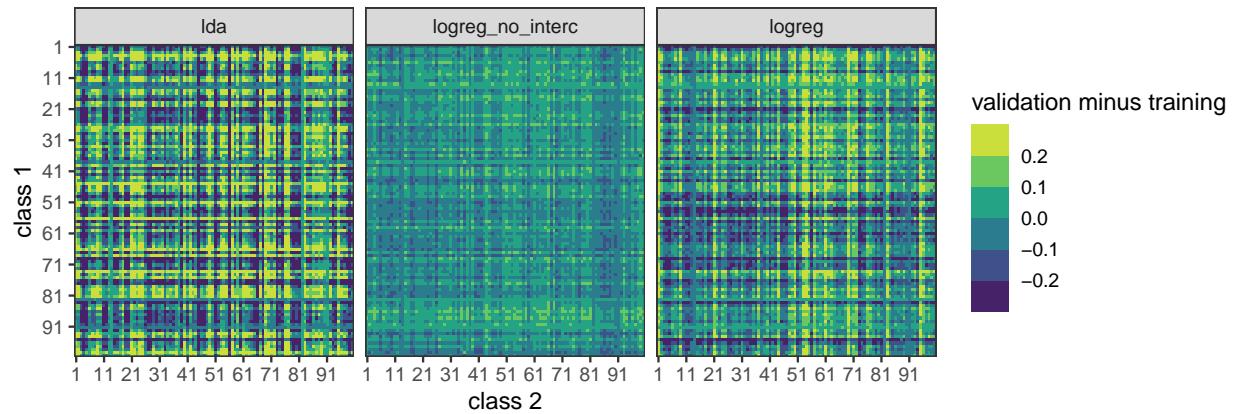
Differences between average pairwise probabilities – class 80



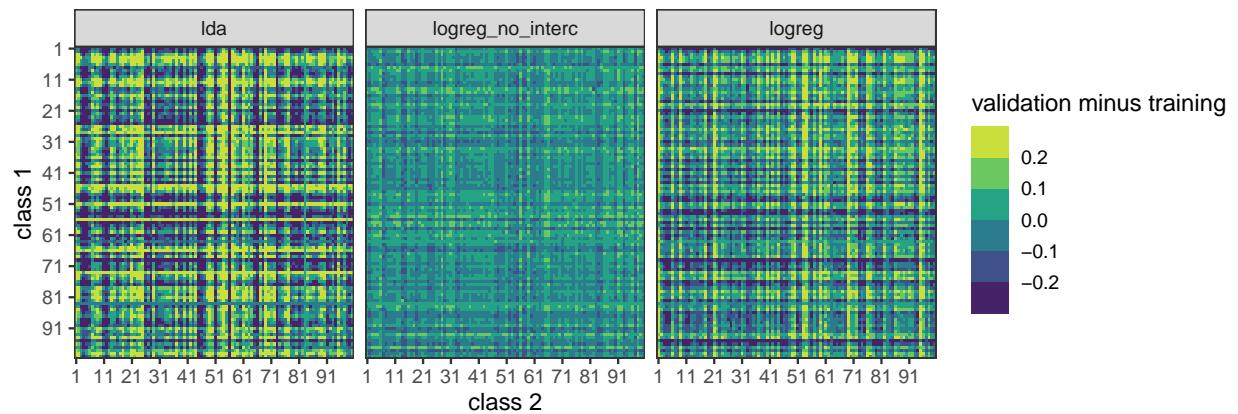
Differences between average pairwise probabilities – class 81



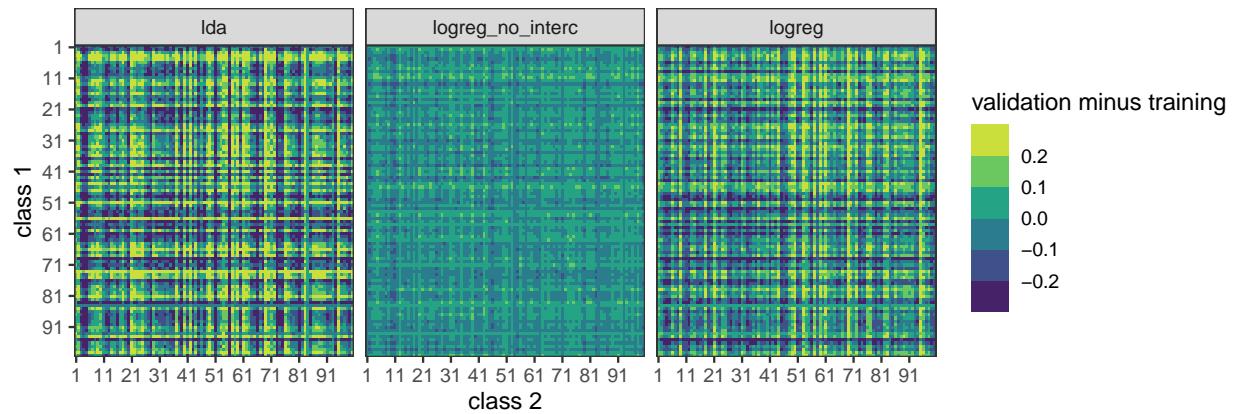
Differences between average pairwise probabilities – class 82



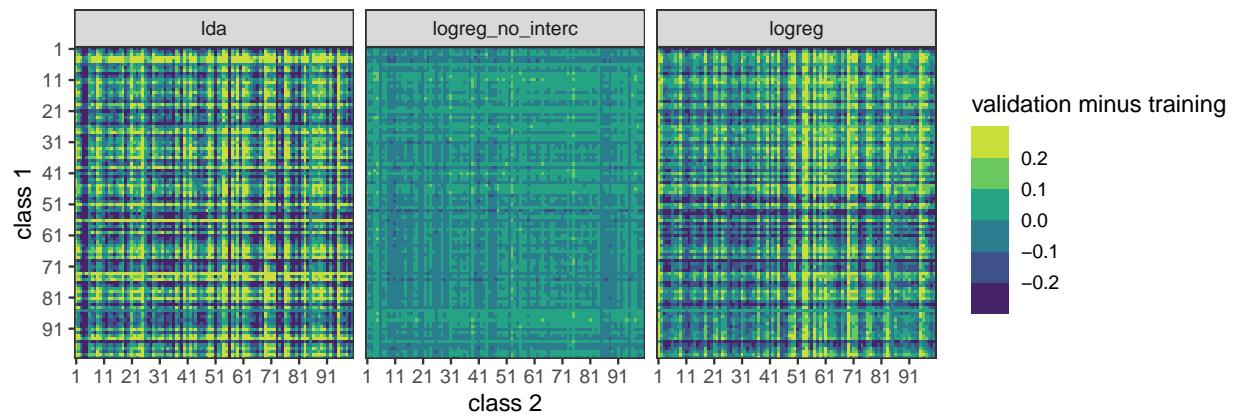
Differences between average pairwise probabilities – class 83



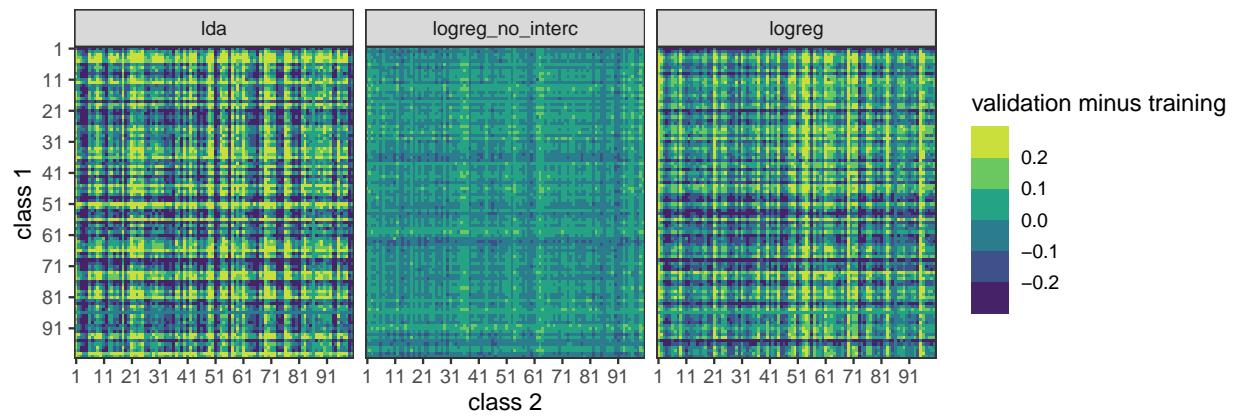
Differences between average pairwise probabilities – class 84



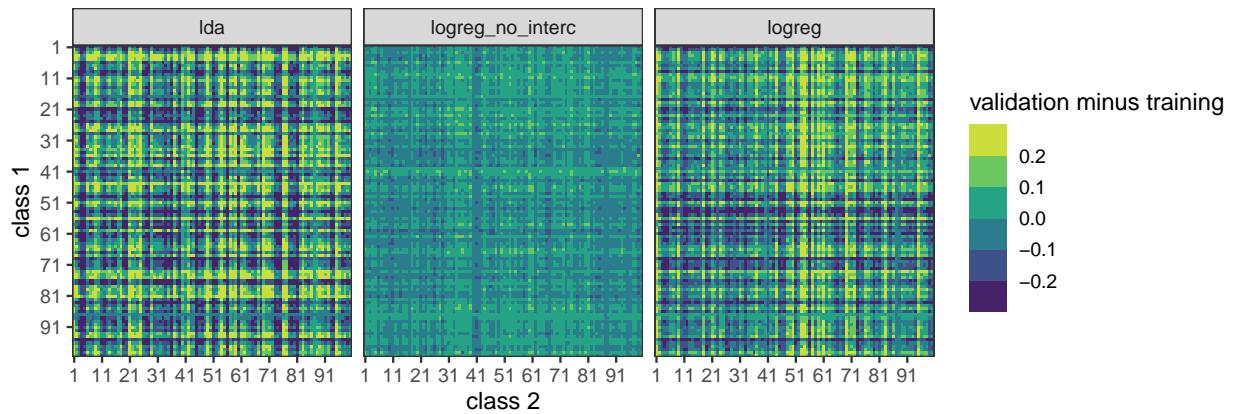
Differences between average pairwise probabilities – class 85



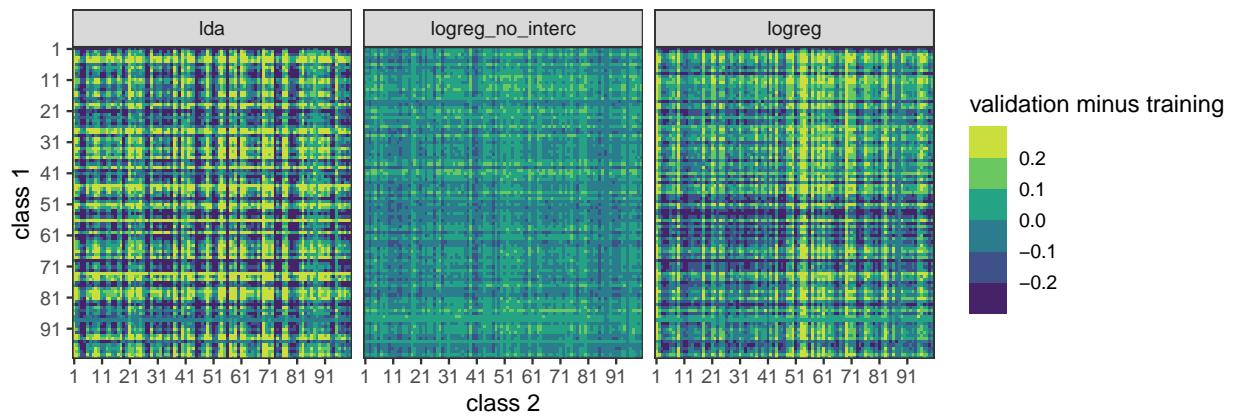
Differences between average pairwise probabilities – class 86



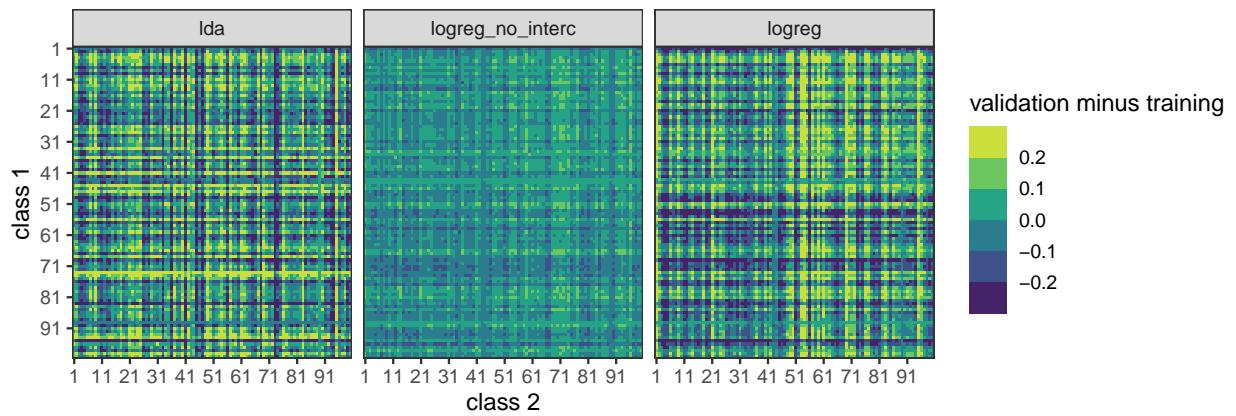
Differences between average pairwise probabilities – class 87



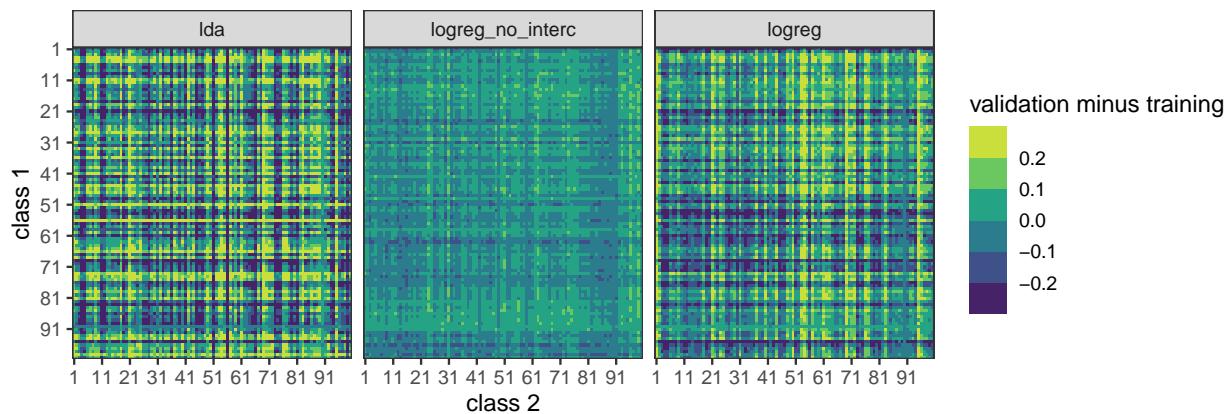
Differences between average pairwise probabilities – class 88



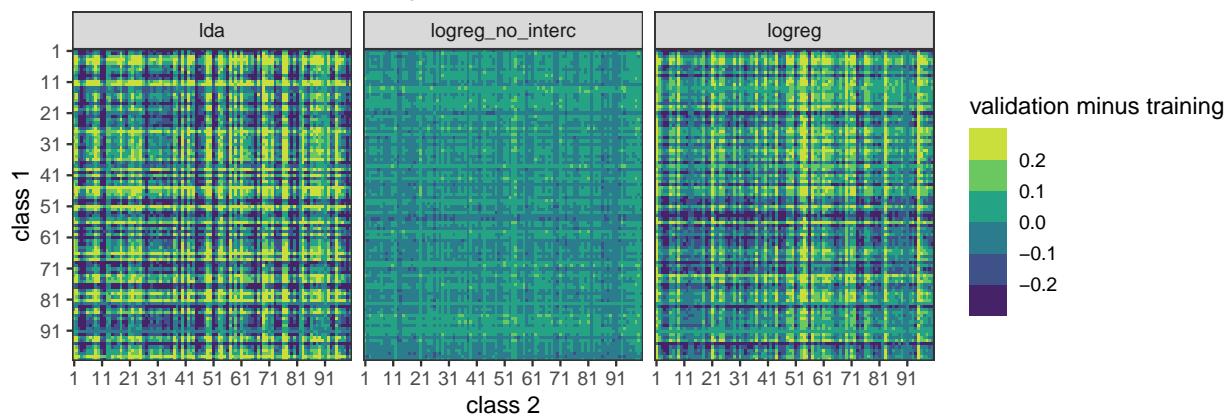
Differences between average pairwise probabilities – class 89



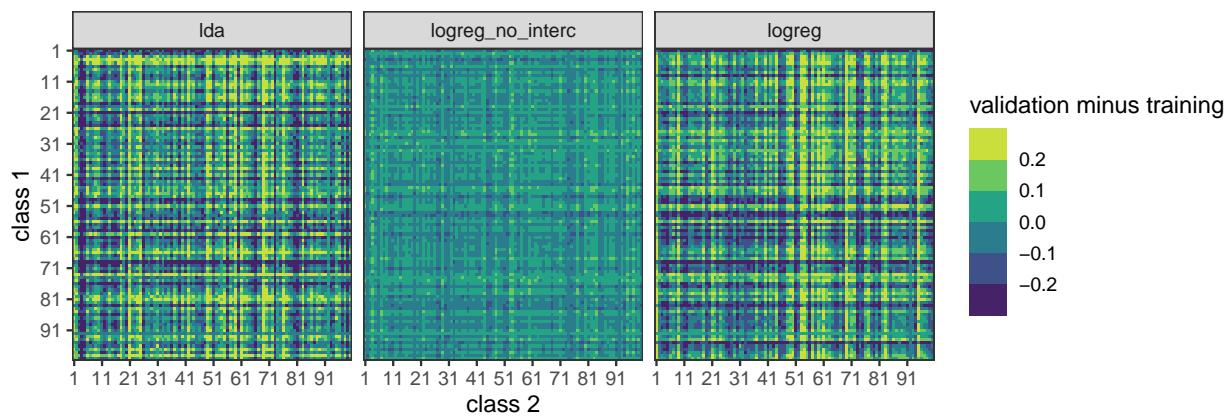
Differences between average pairwise probabilities – class 90



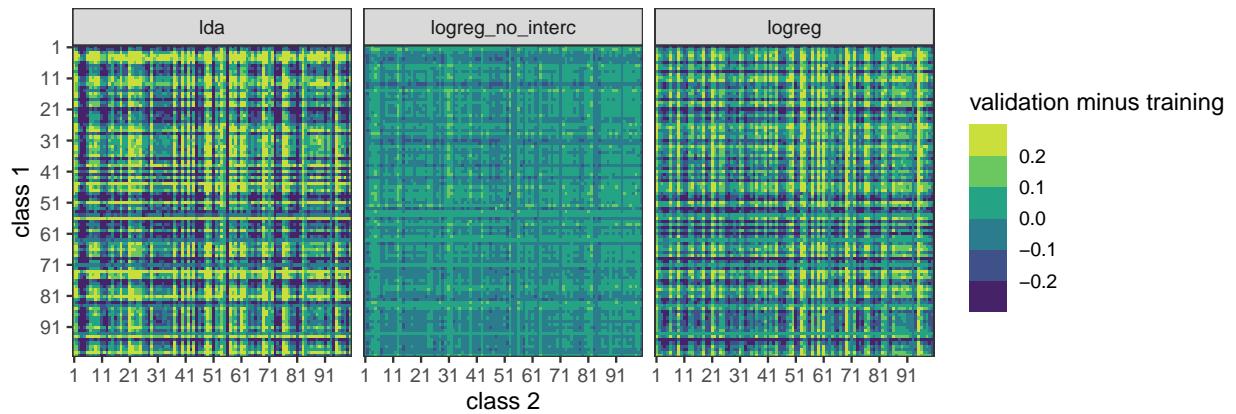
Differences between average pairwise probabilities – class 91



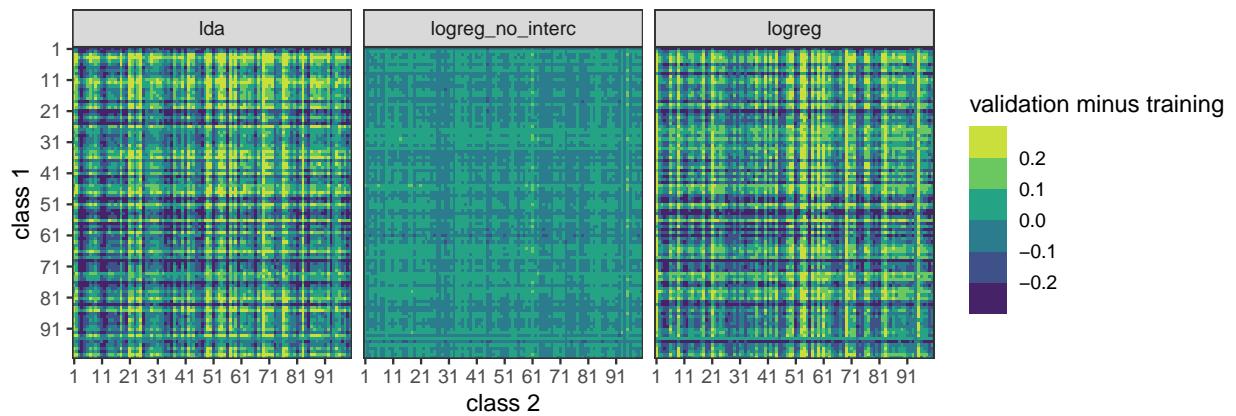
Differences between average pairwise probabilities – class 92



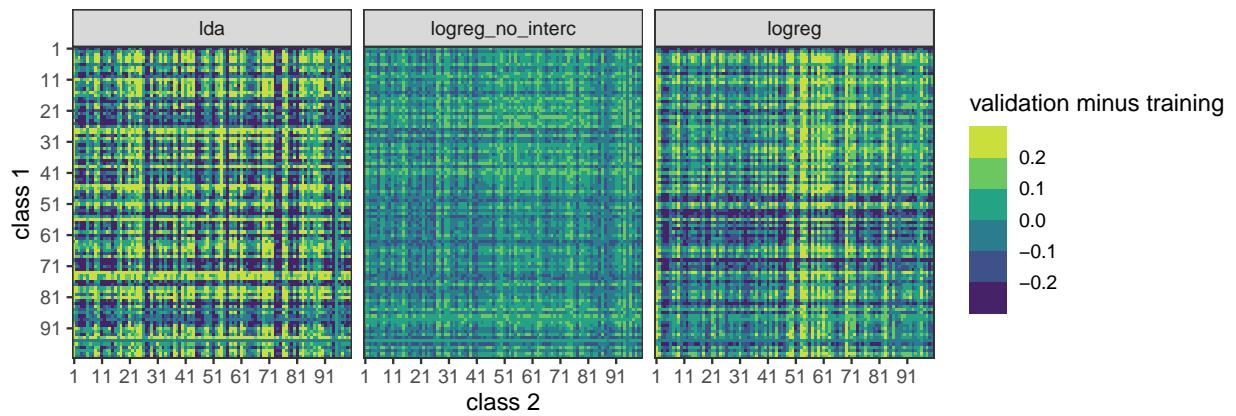
Differences between average pairwise probabilities – class 93



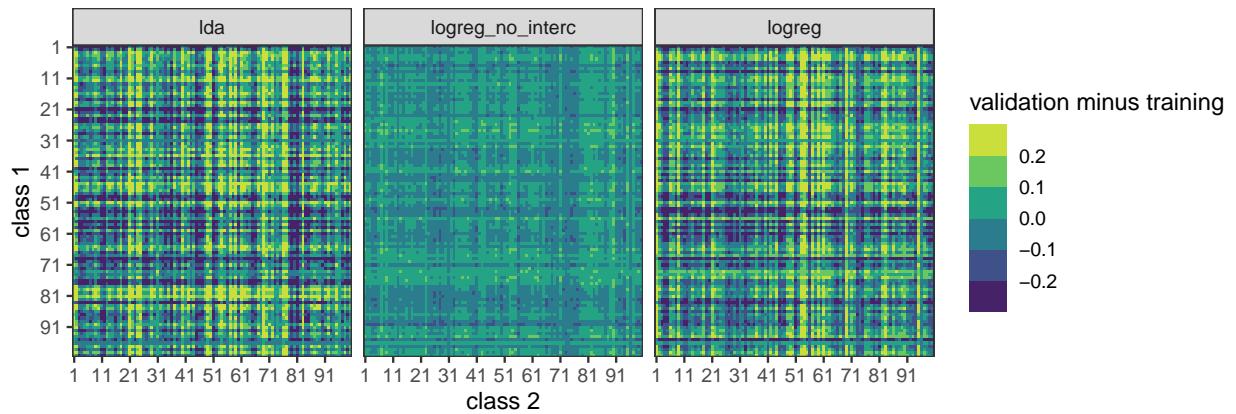
Differences between average pairwise probabilities – class 94



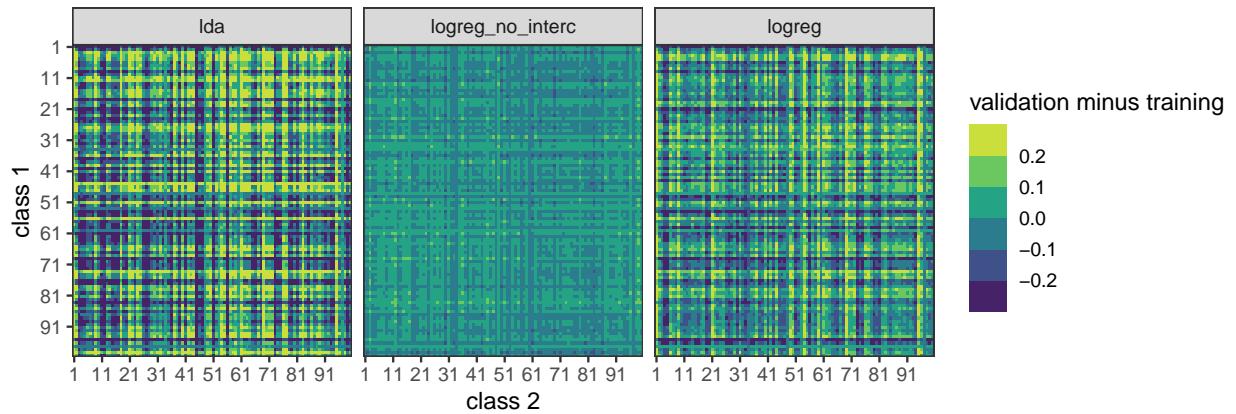
Differences between average pairwise probabilities – class 95



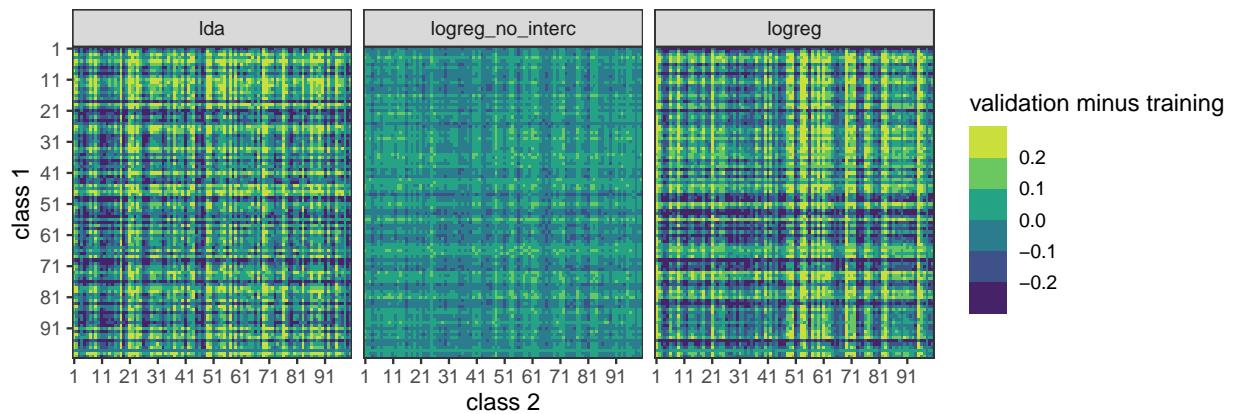
Differences between average pairwise probabilities – class 96



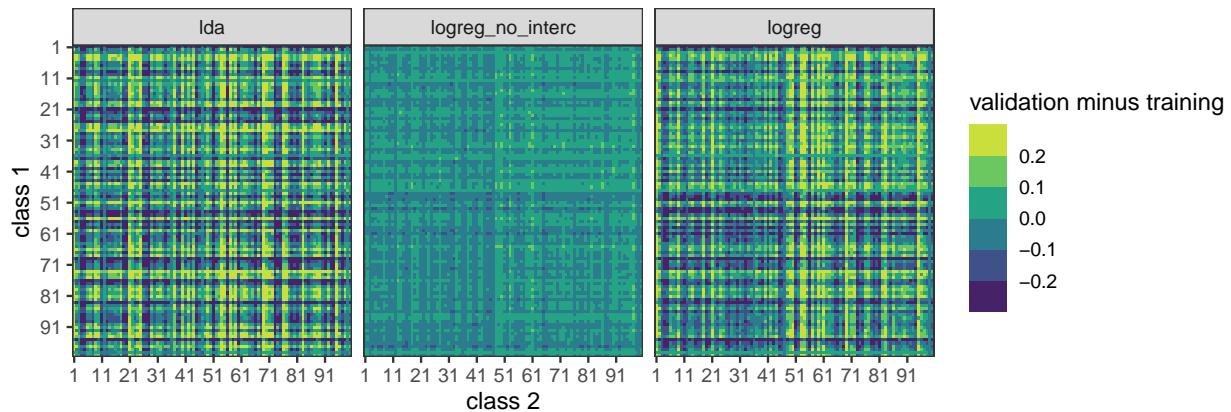
Differences between average pairwise probabilities – class 97



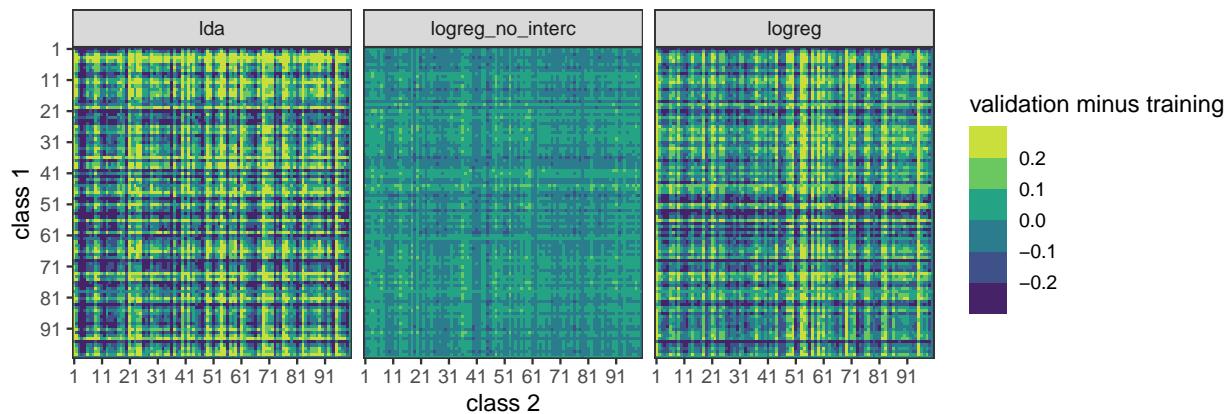
Differences between average pairwise probabilities – class 98



Differences between average pairwise probabilities – class 99



Differences between average pairwise probabilities – class 100

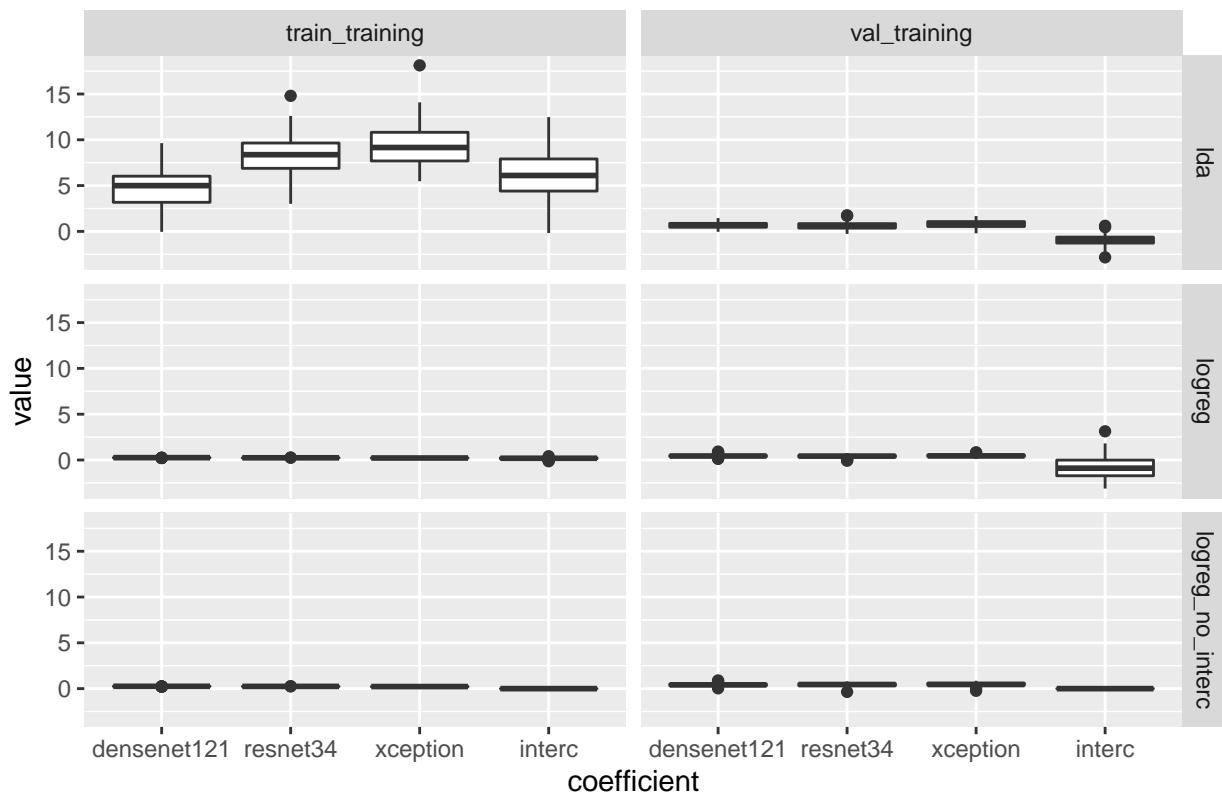


Logistic regression without intercept has lower differences between tt and vt R matrices than other combining methods.

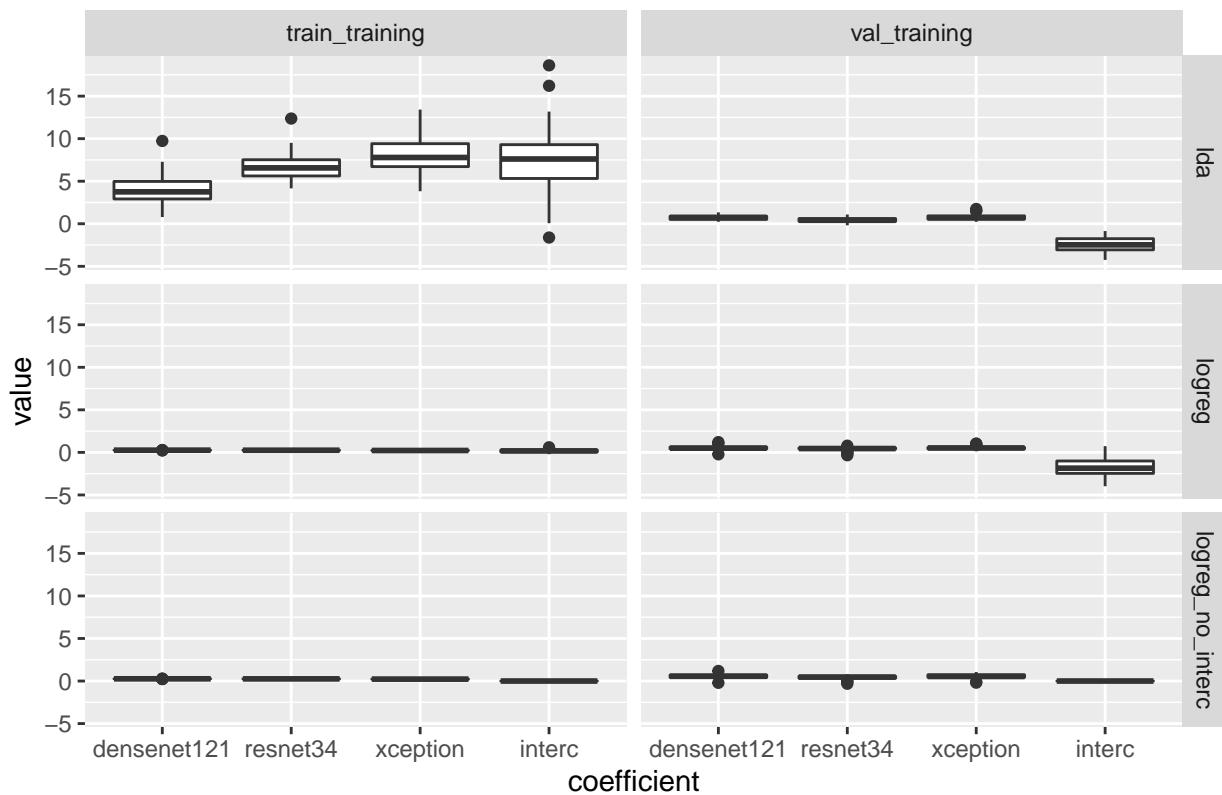
```
combiner_coefs <- load_combiner_coefs(base_dir, repls, folds)
```

```
for (cl1 in 1:(10 - 1))
{
  for (cl2 in (cl1 + 1):10)
  {
    cur_plt <- combiner_coefs %>% filter(class1 == cl1 & class2 == cl2) %>% ggplot() + geom_boxplot(aes
      facet_grid(cols=vars(train_type), rows=vars(combining_method)) + ggtitle(paste("Coefficients for "
    print(cur_plt)
  }
}
```

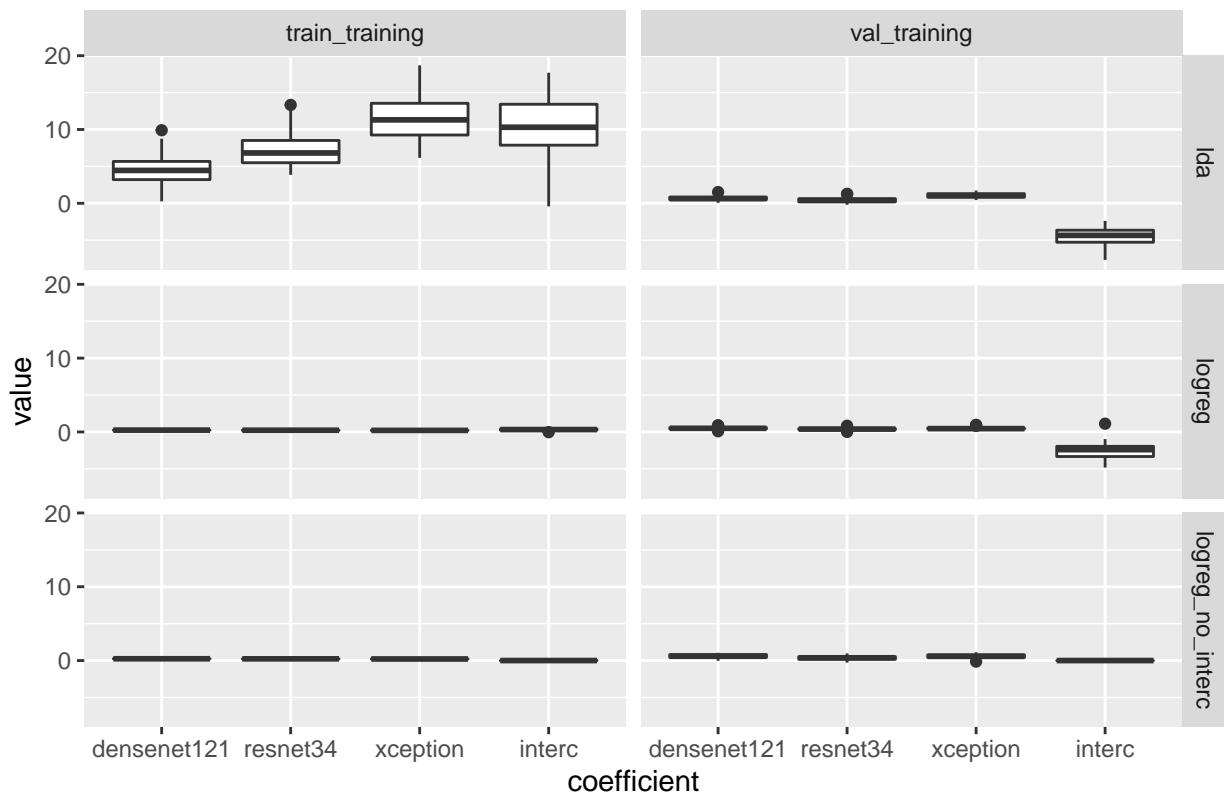
### Coefficients for class 1 vs 2



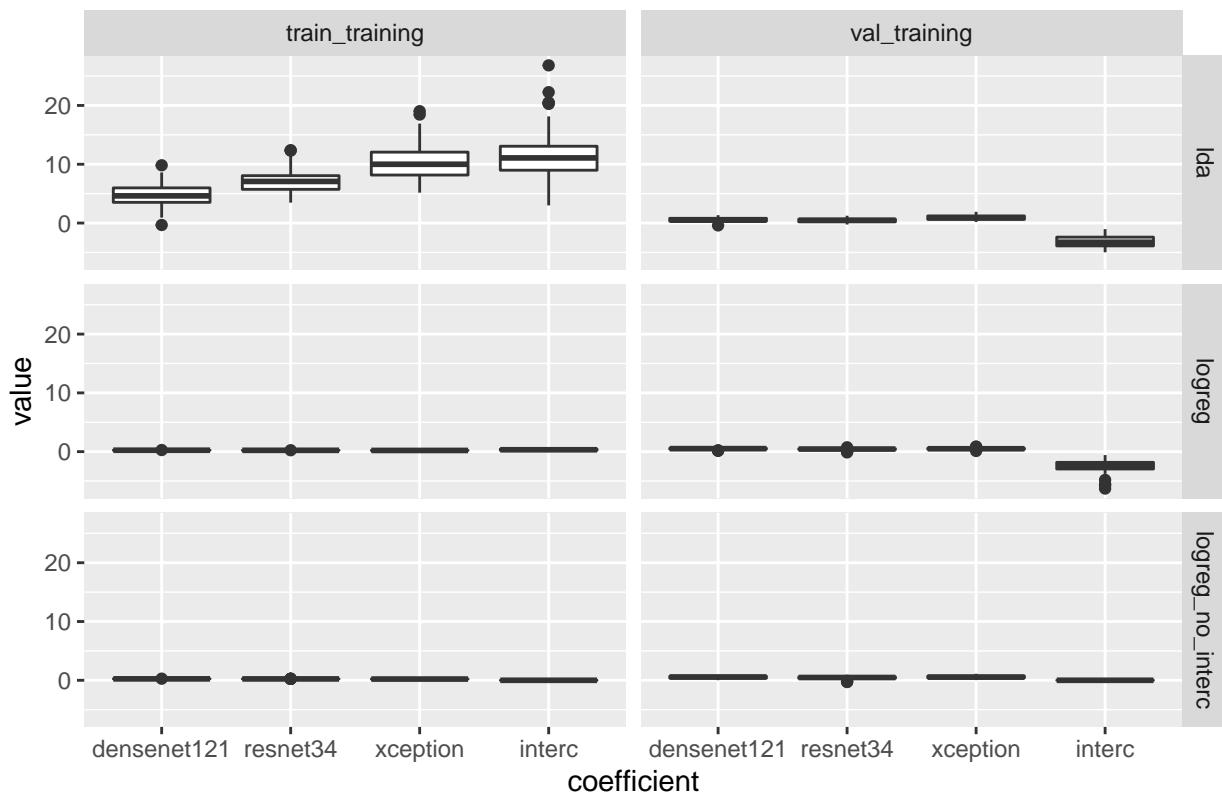
### Coefficients for class 1 vs 3



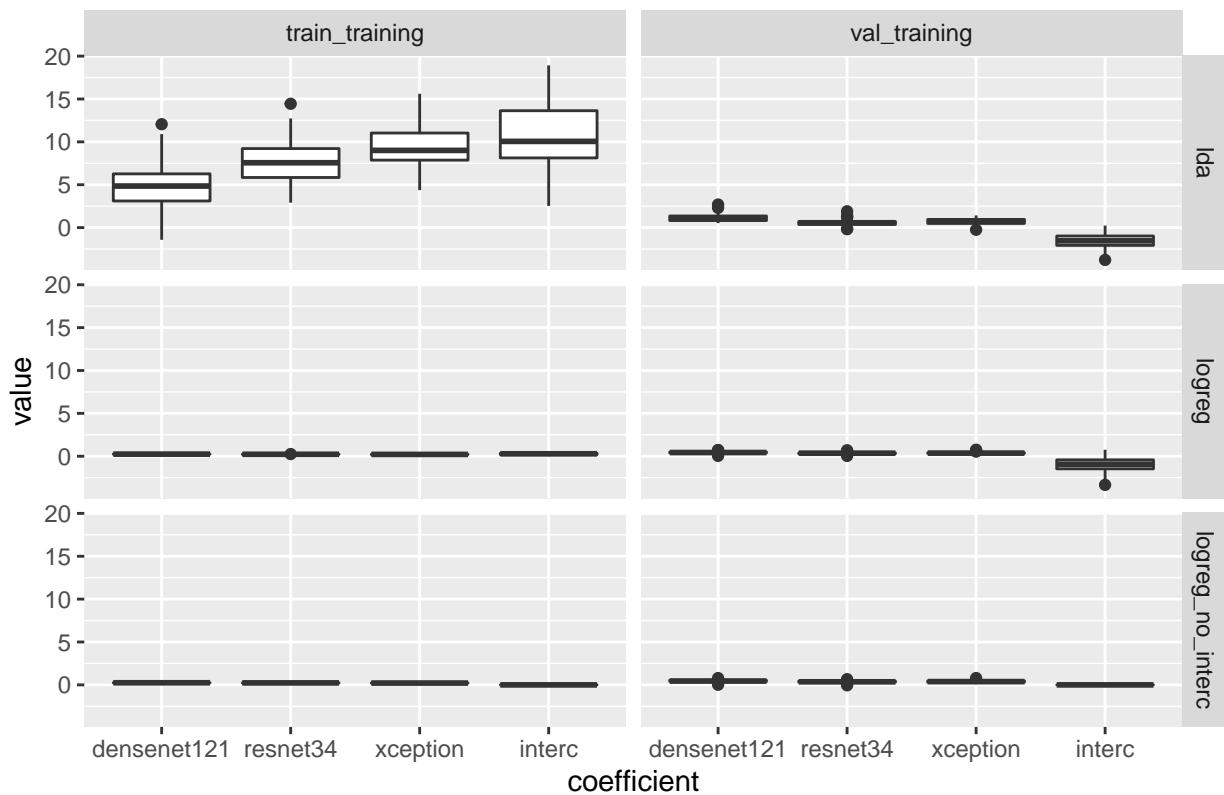
### Coefficients for class 1 vs 4



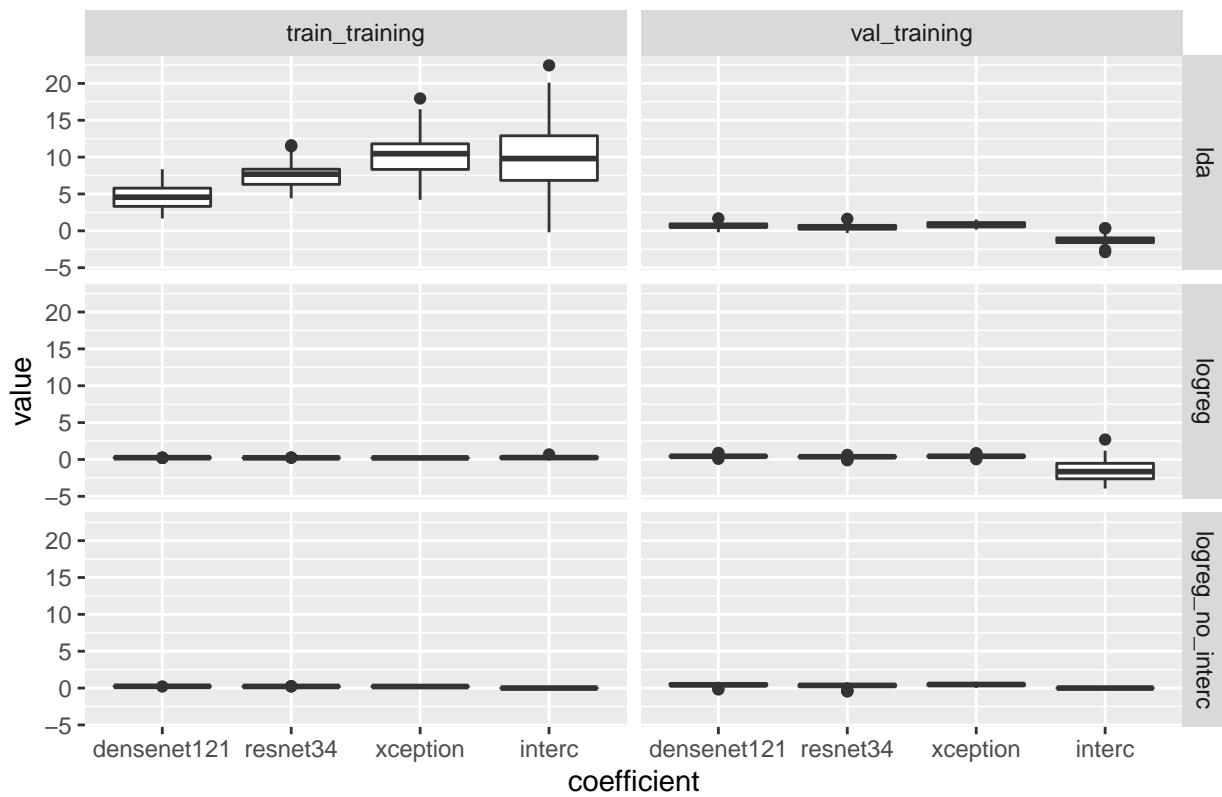
### Coefficients for class 1 vs 5



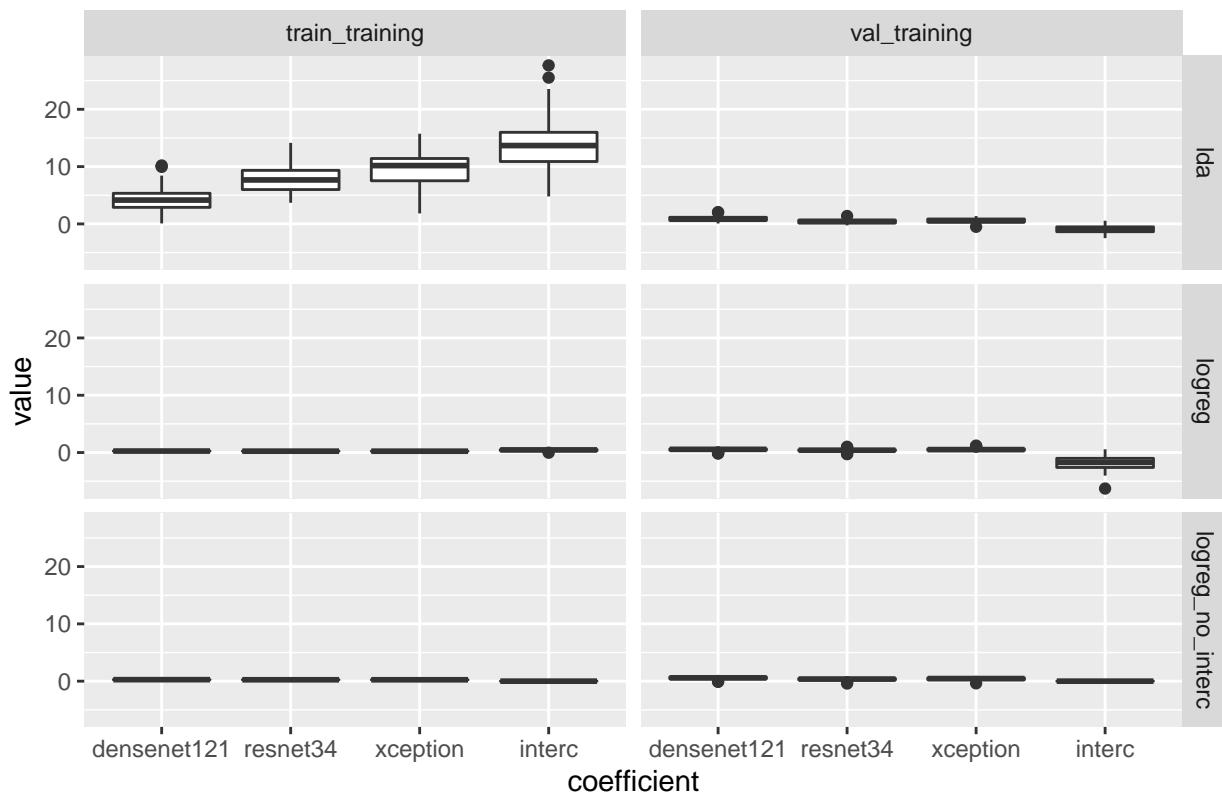
### Coefficients for class 1 vs 6



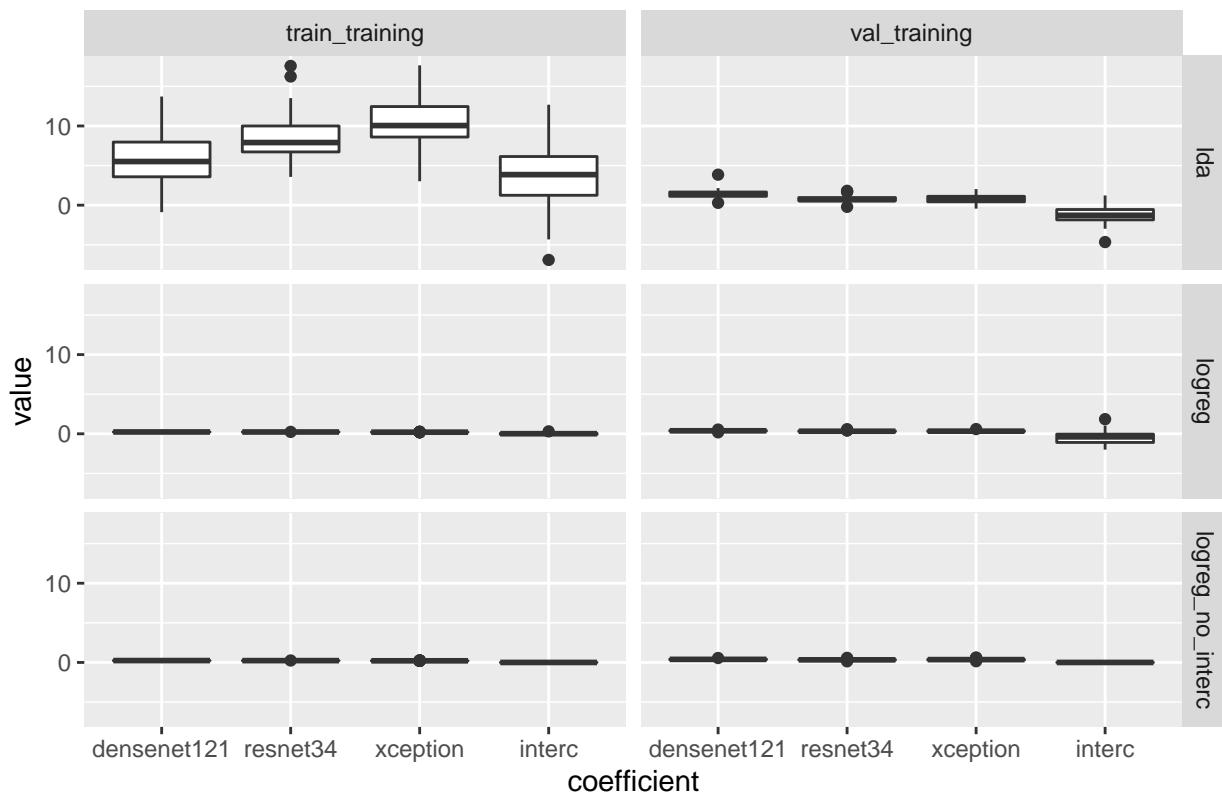
### Coefficients for class 1 vs 7



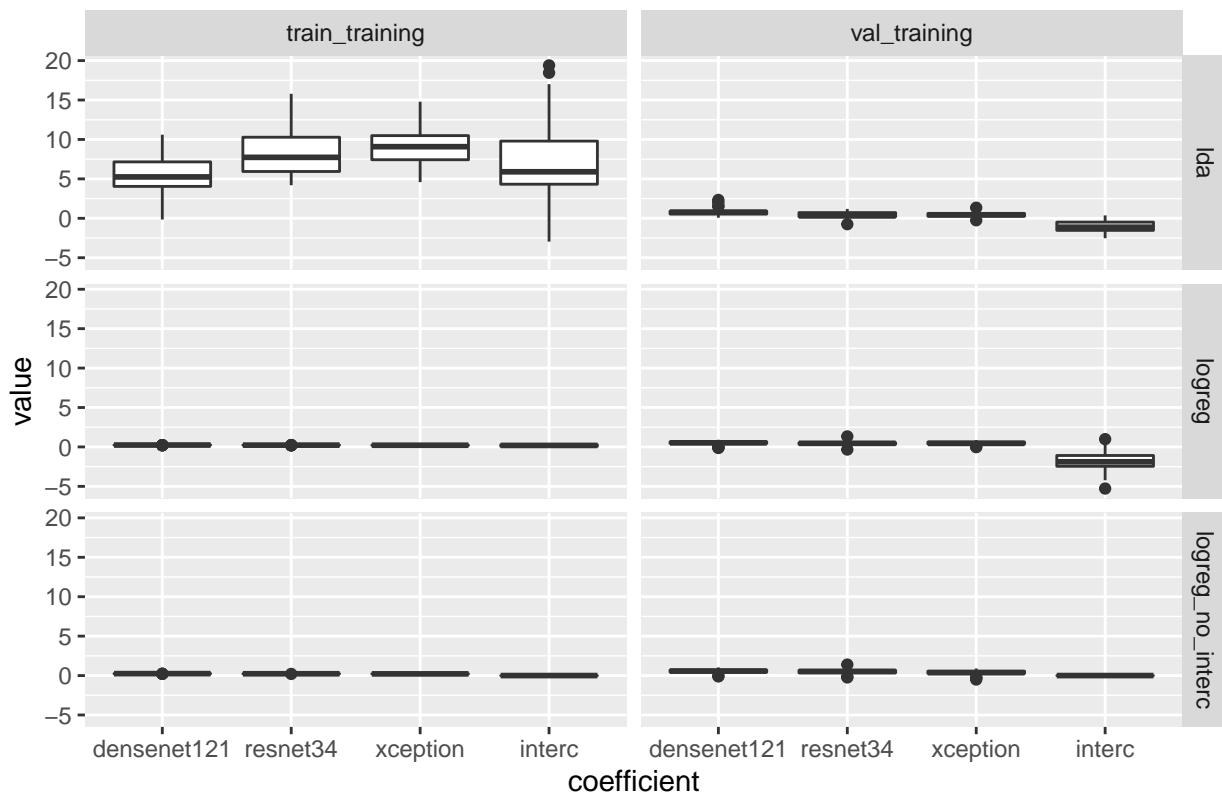
### Coefficients for class 1 vs 8



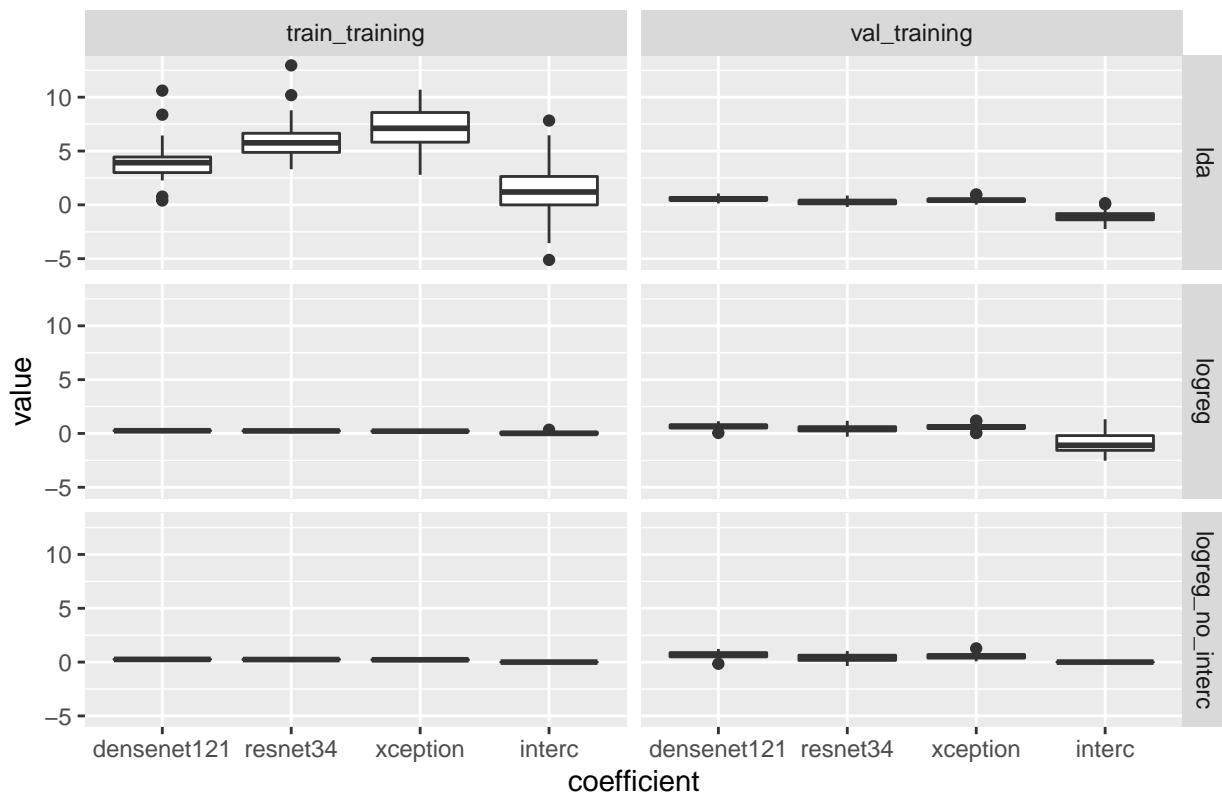
### Coefficients for class 1 vs 9



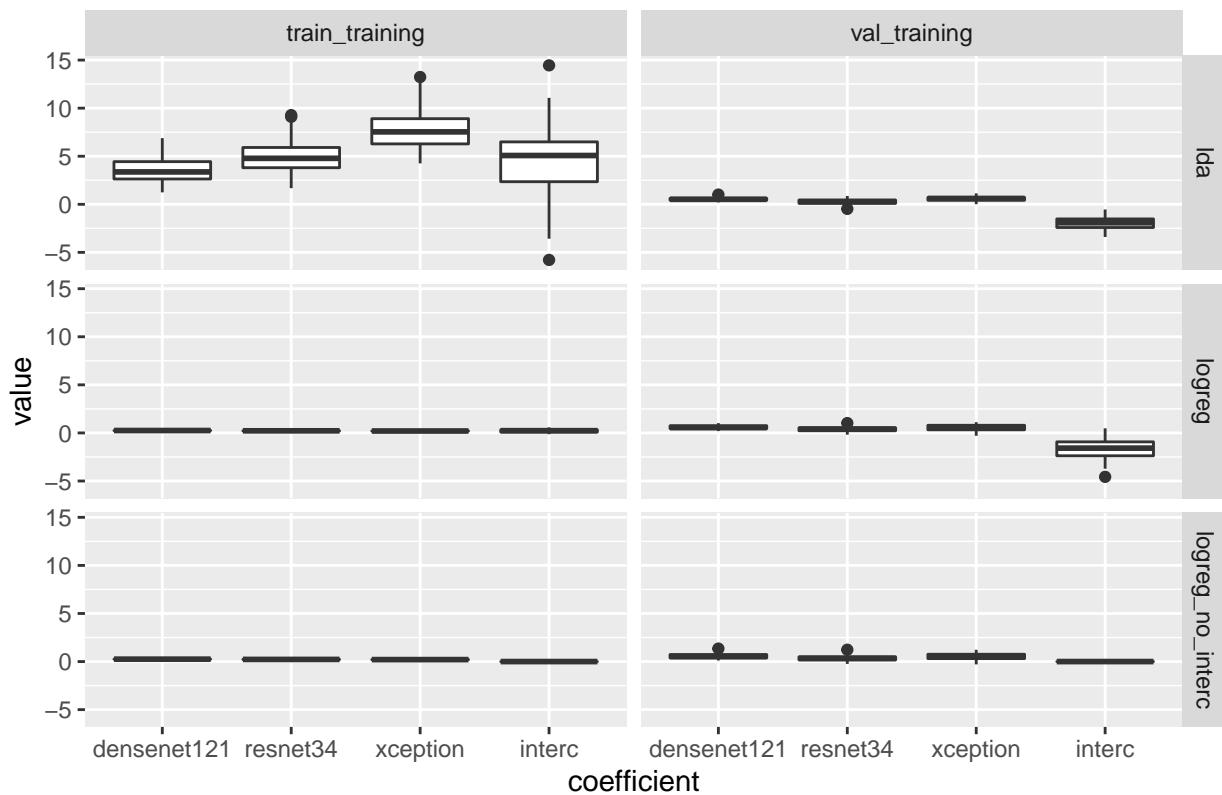
### Coefficients for class 1 vs 10



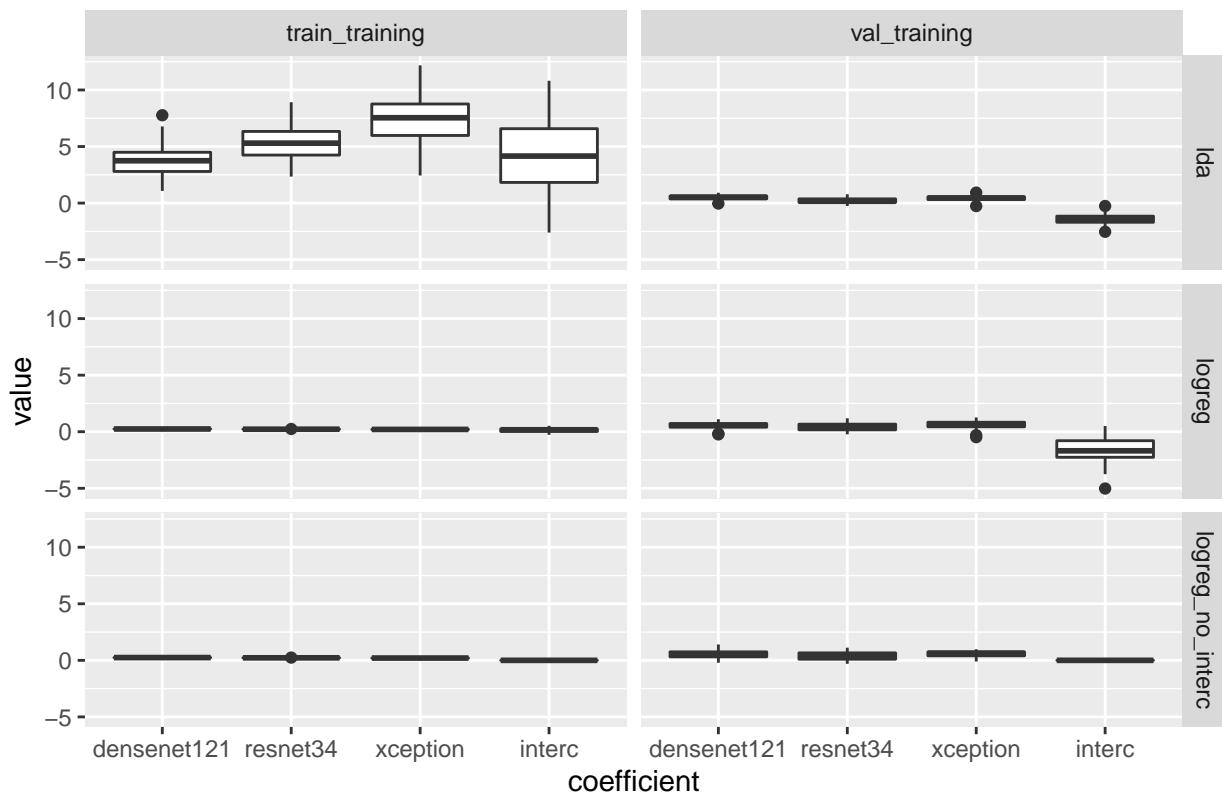
### Coefficients for class 2 vs 3



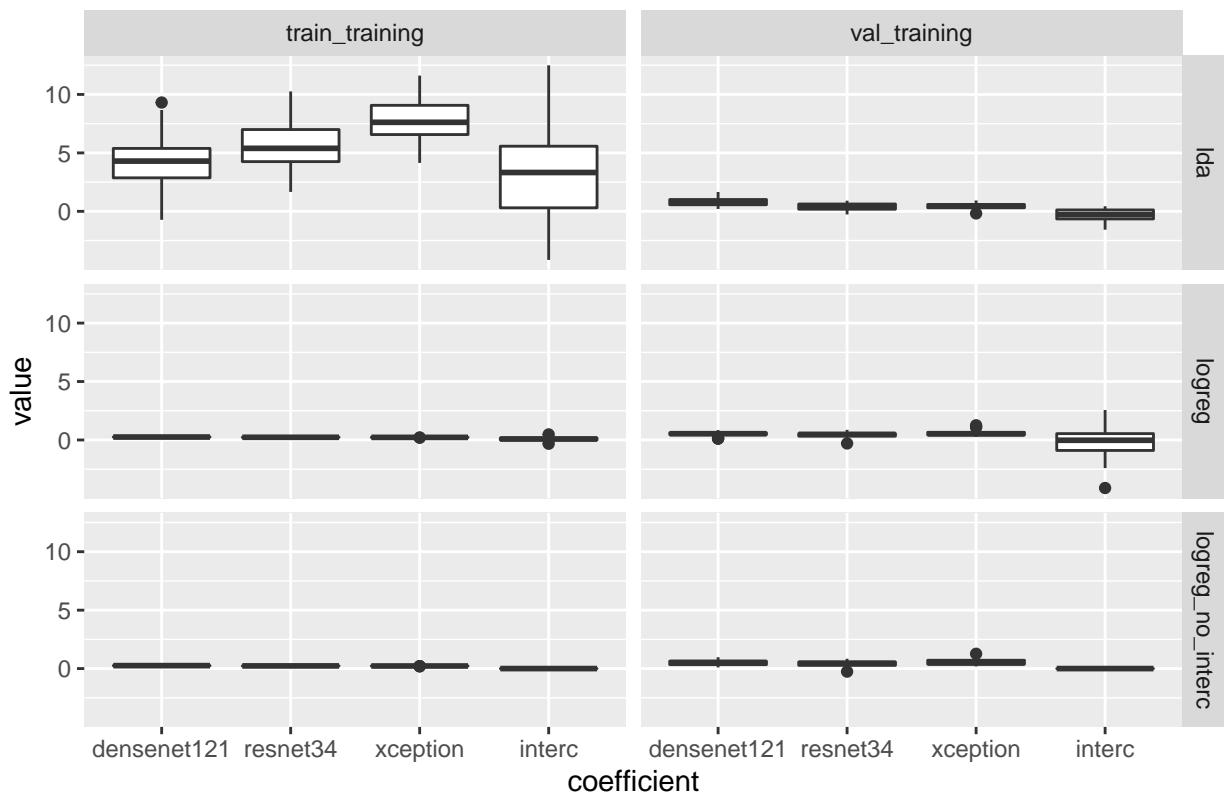
### Coefficients for class 2 vs 4



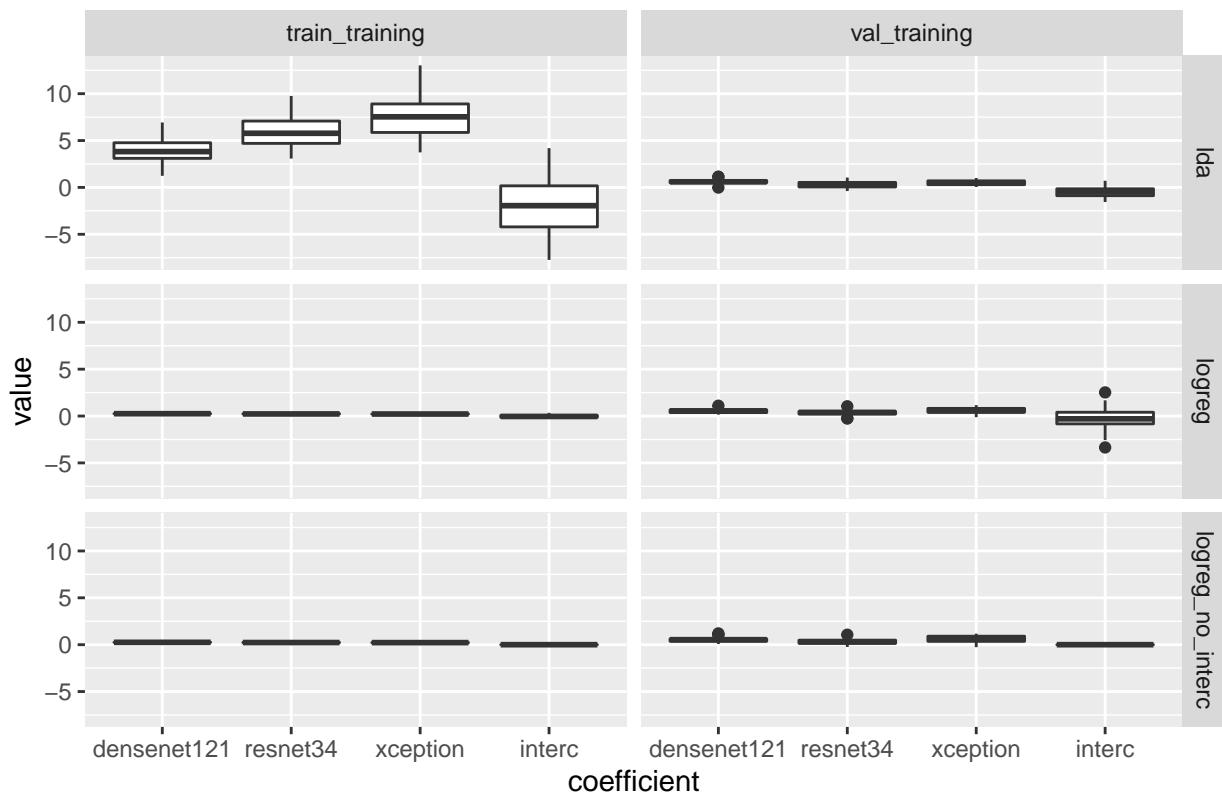
### Coefficients for class 2 vs 5



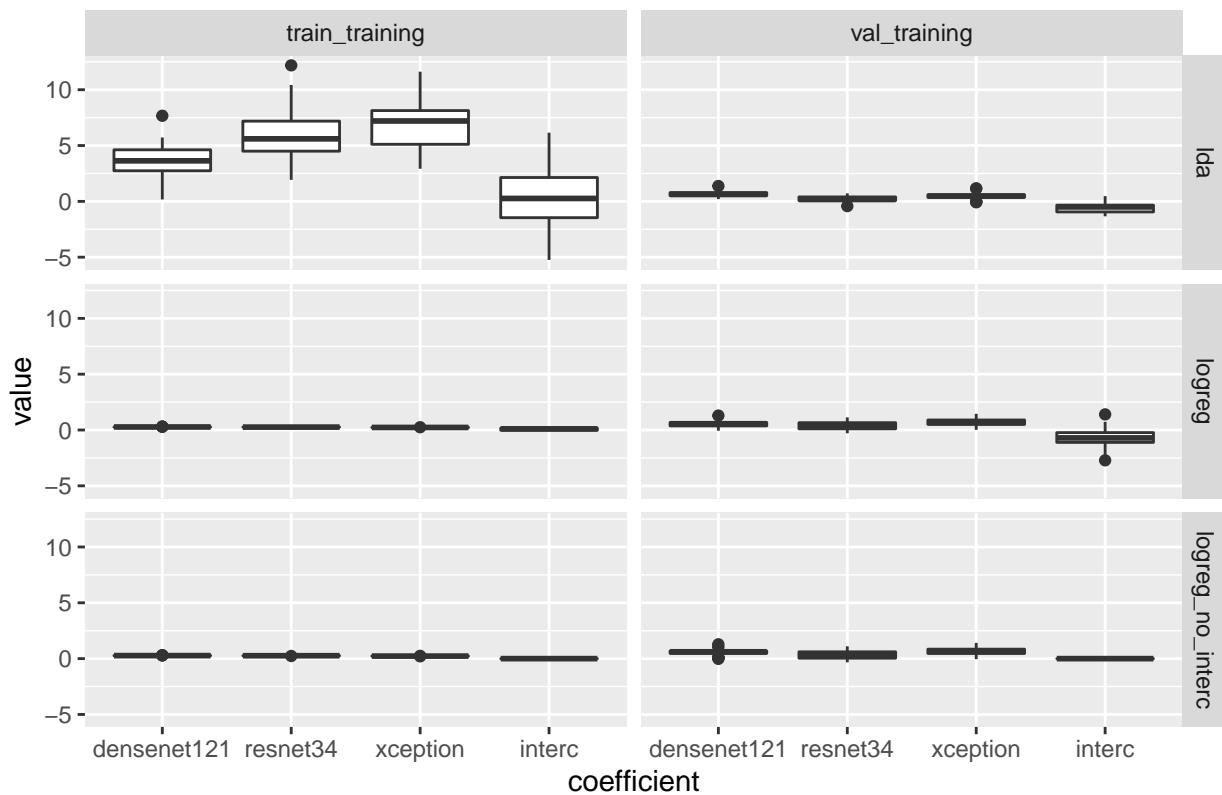
### Coefficients for class 2 vs 6



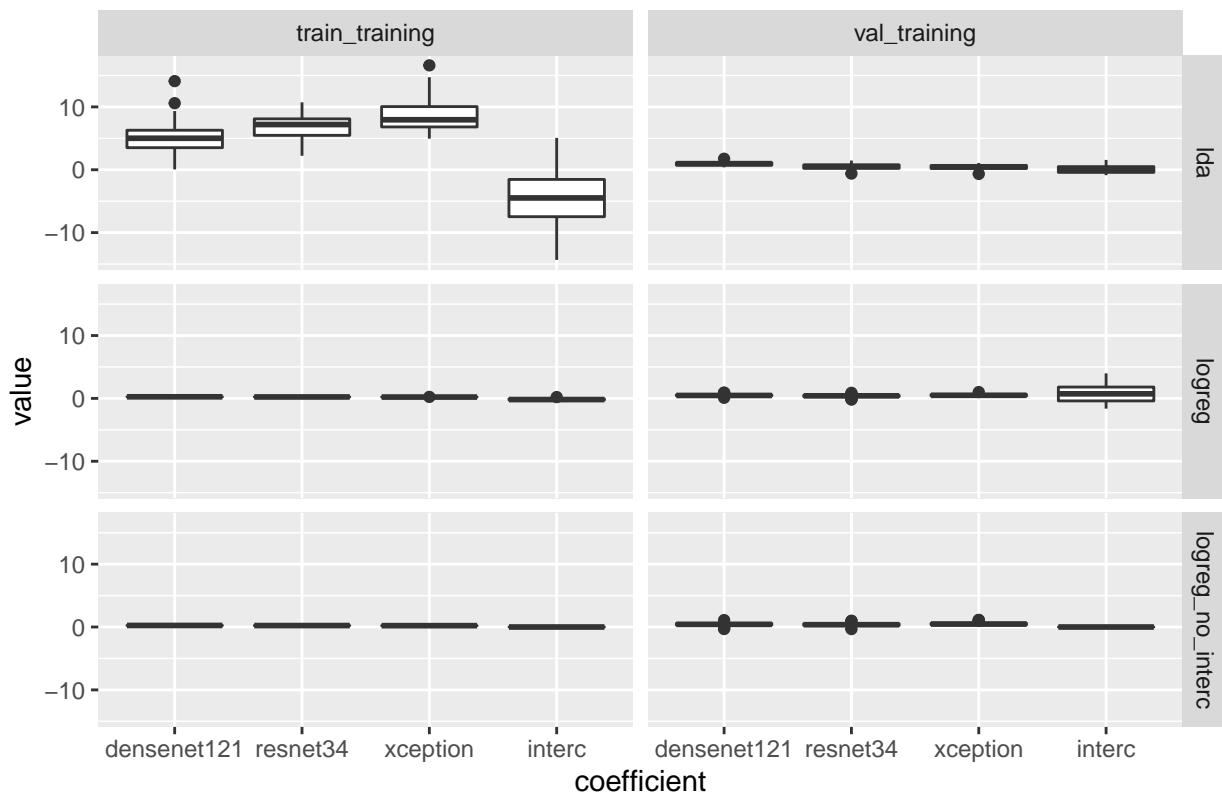
### Coefficients for class 2 vs 7



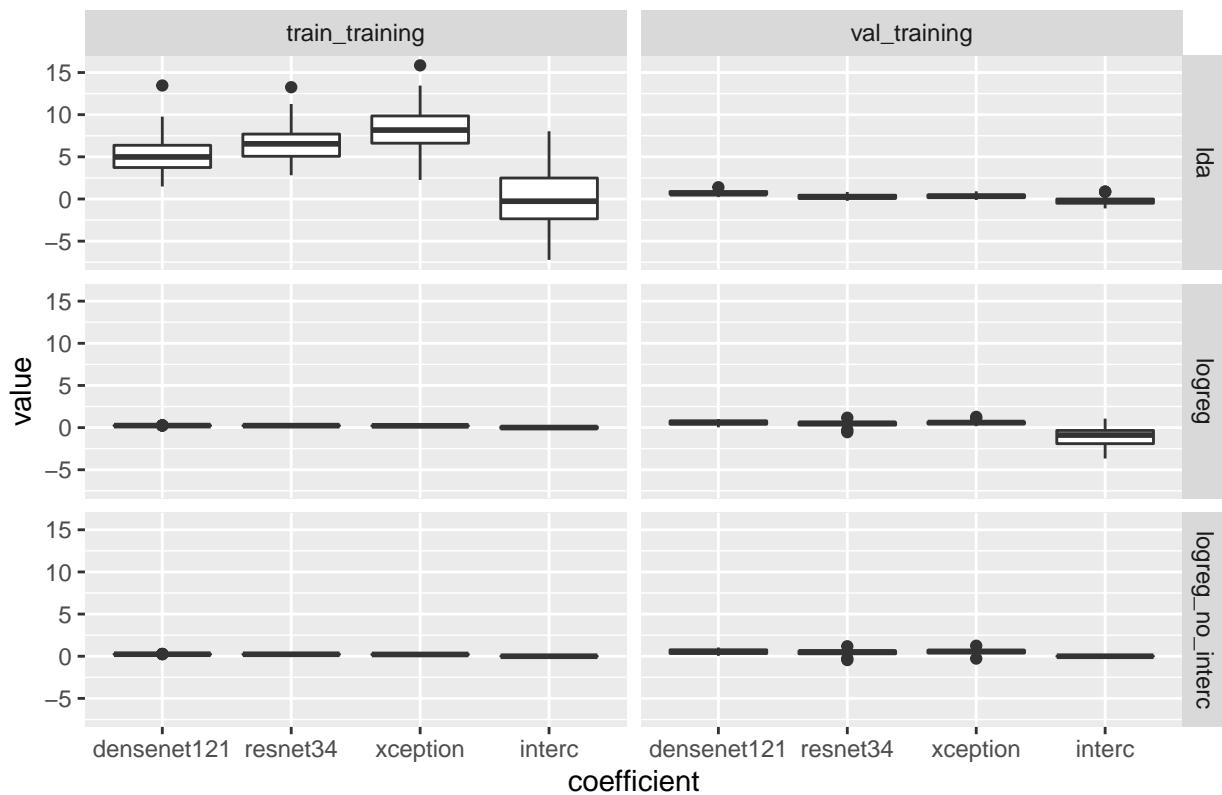
### Coefficients for class 2 vs 8



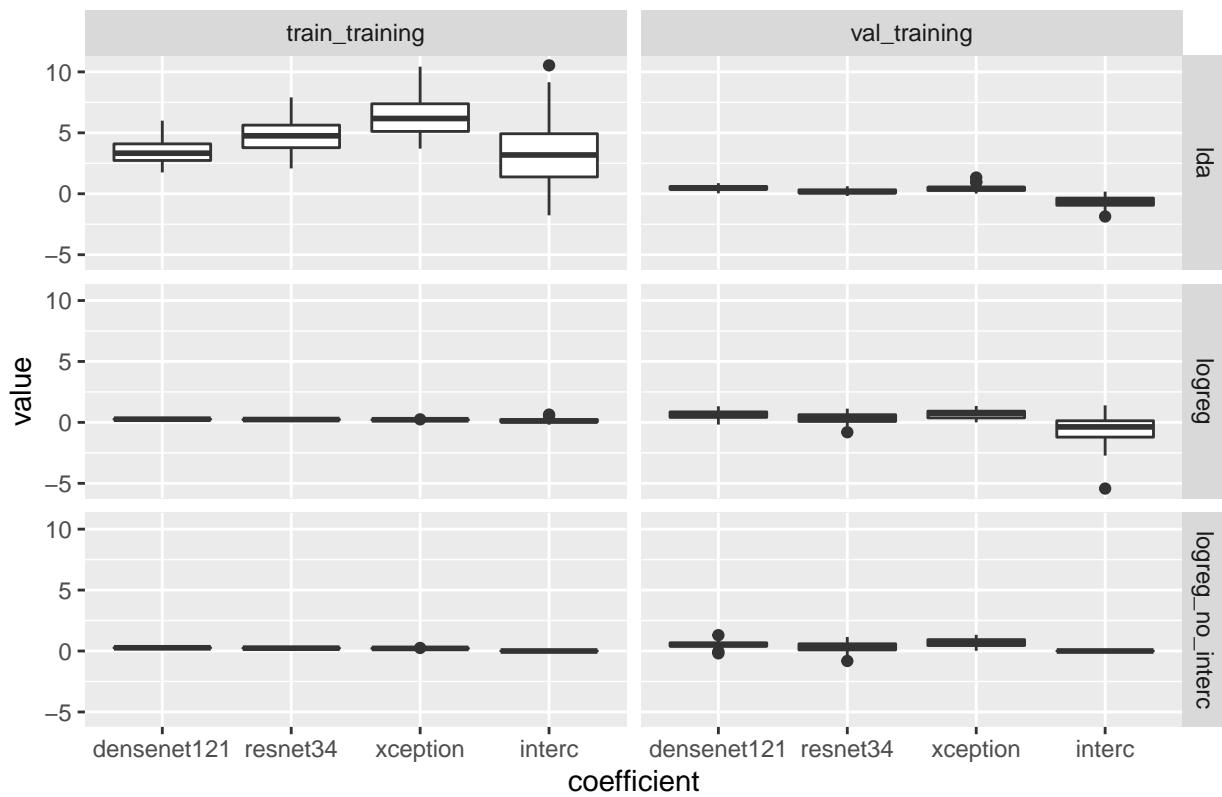
## Coefficients for class 2 vs 9



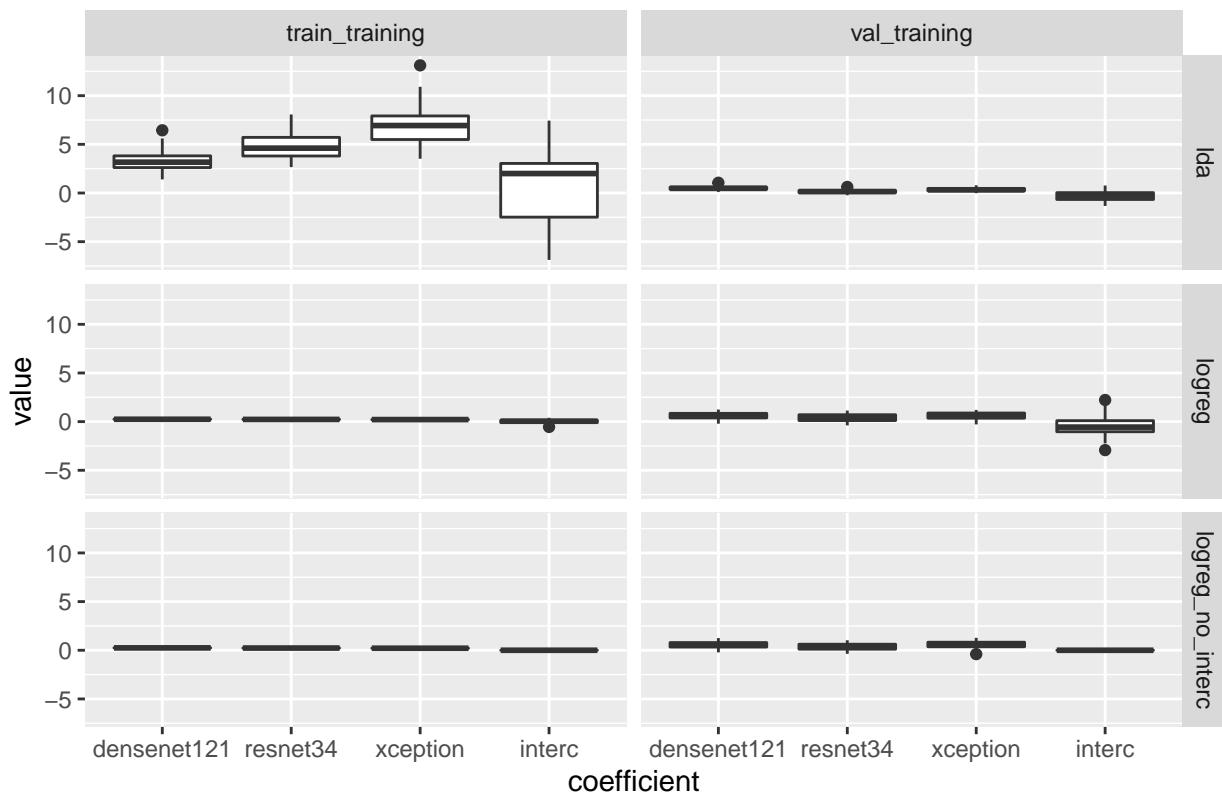
### Coefficients for class 2 vs 10



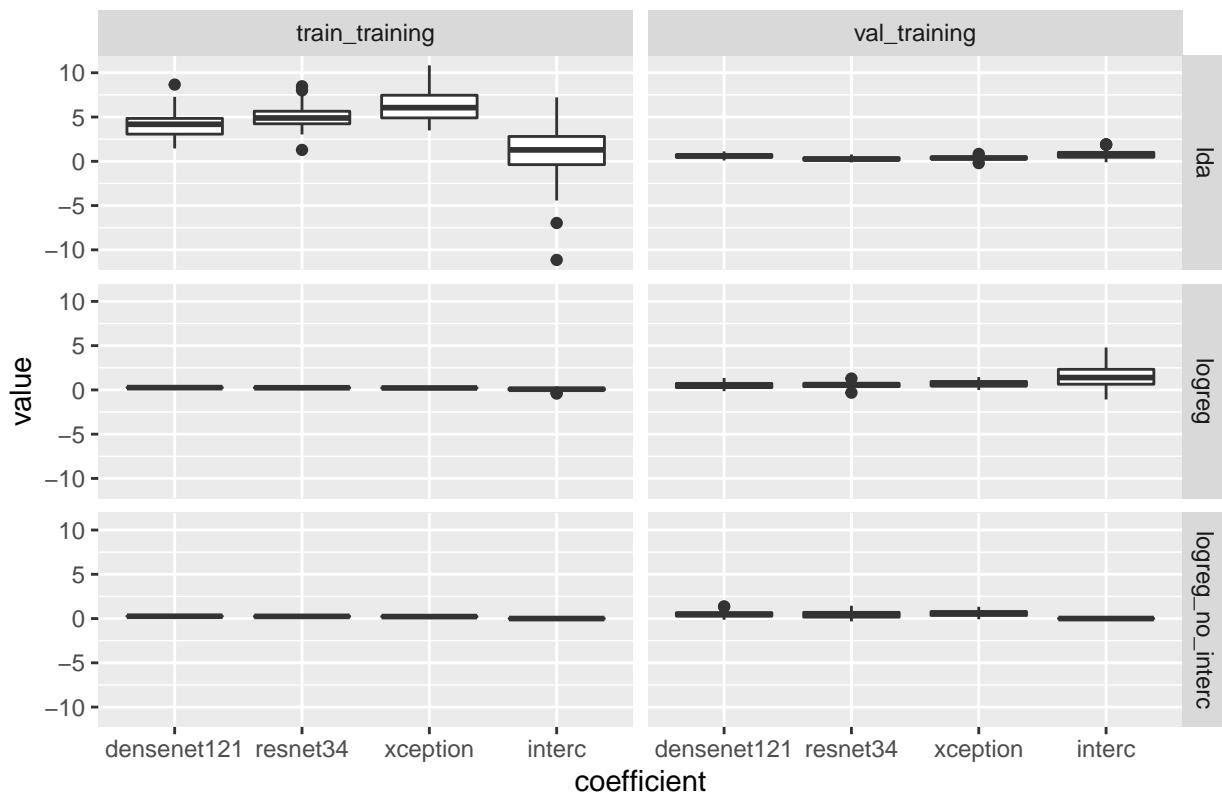
### Coefficients for class 3 vs 4



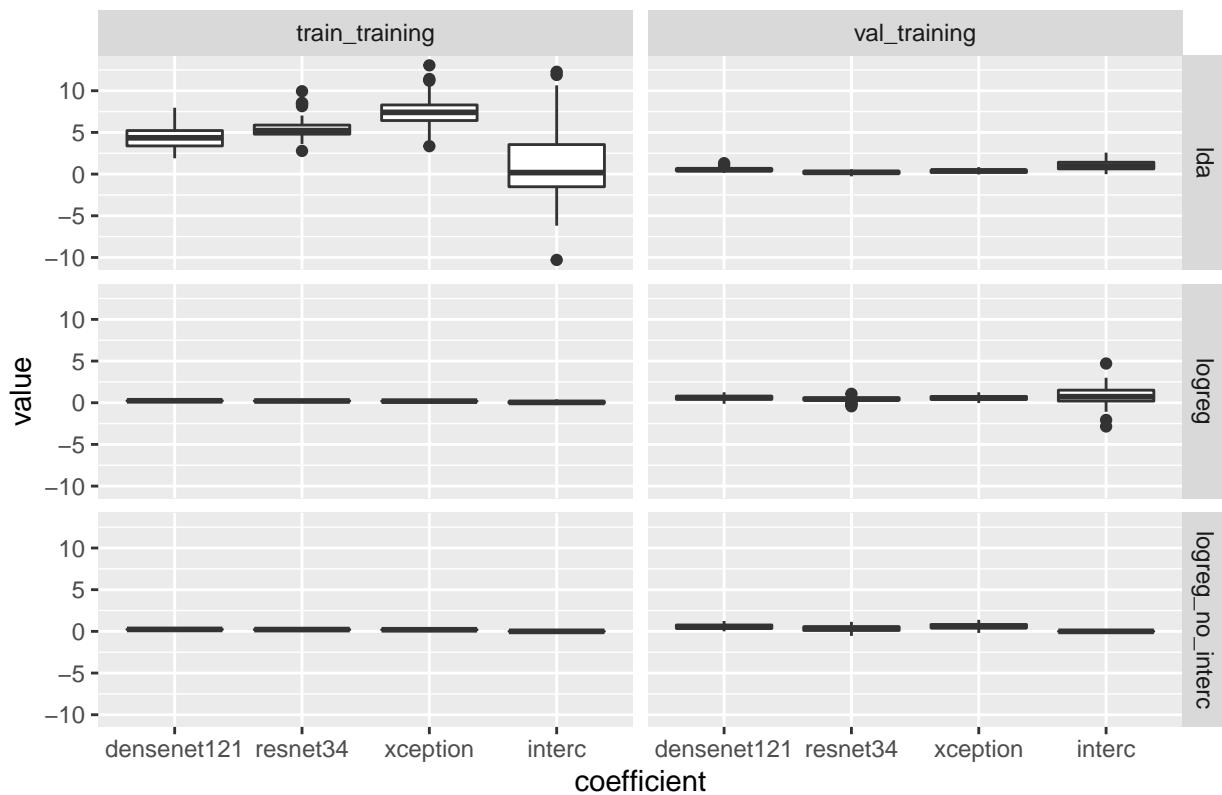
### Coefficients for class 3 vs 5



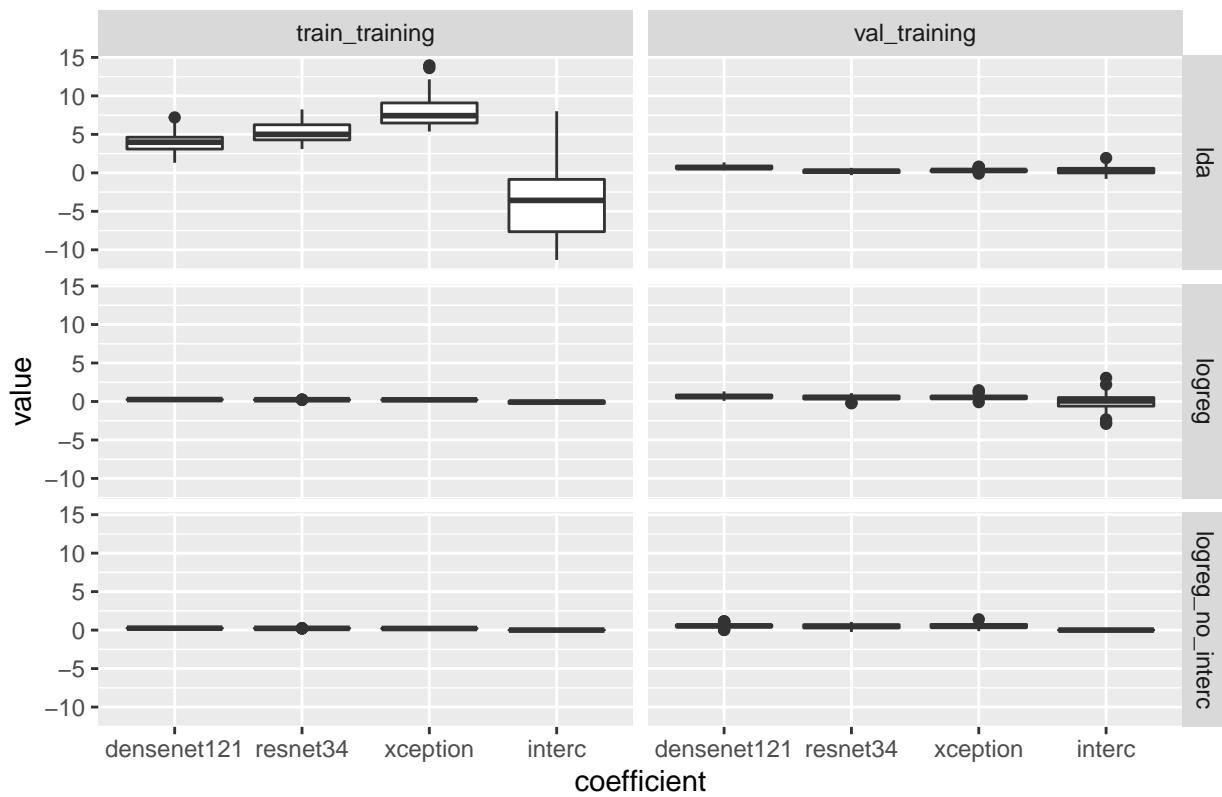
### Coefficients for class 3 vs 6



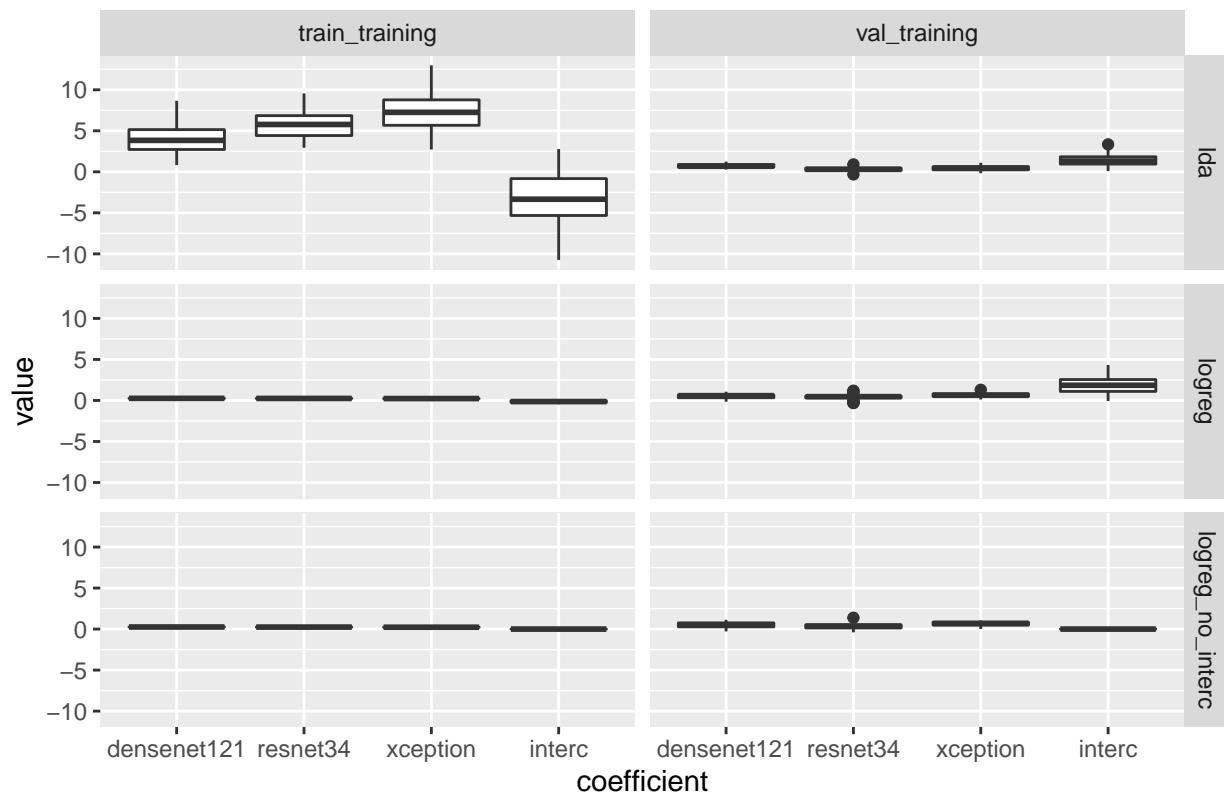
### Coefficients for class 3 vs 7



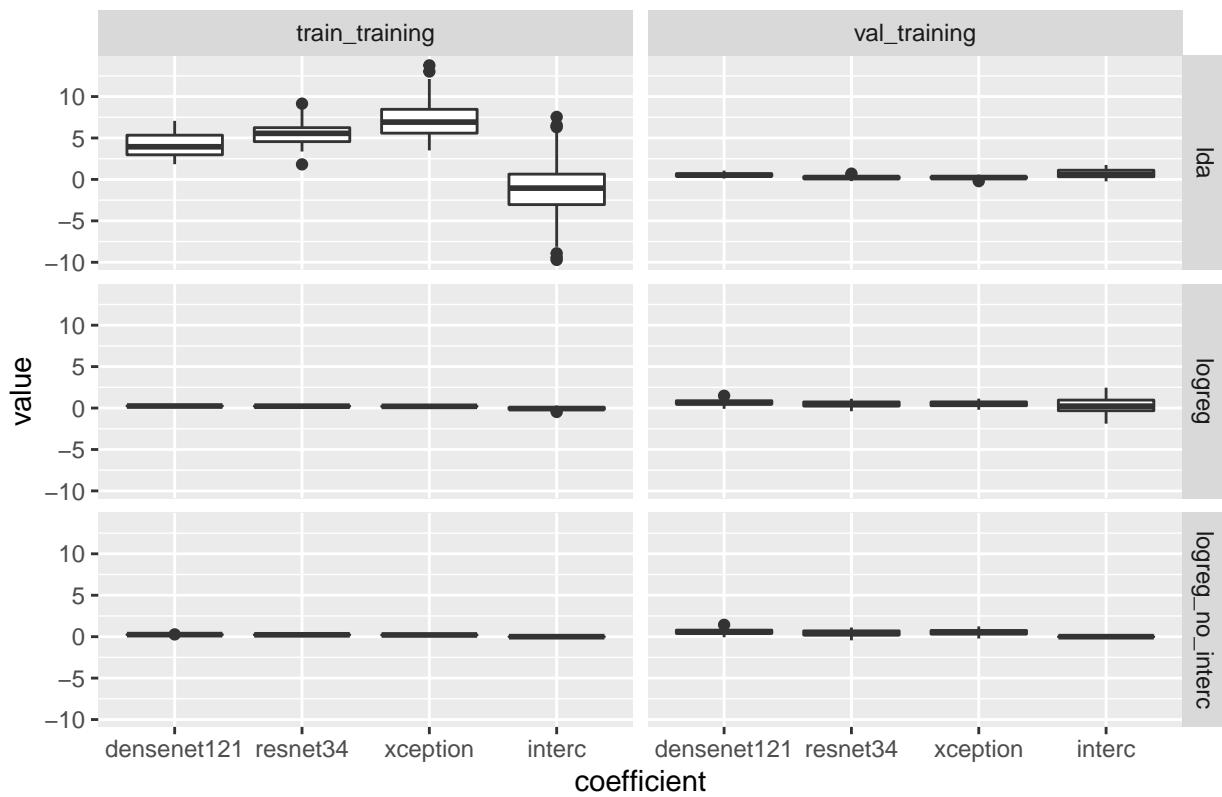
### Coefficients for class 3 vs 8



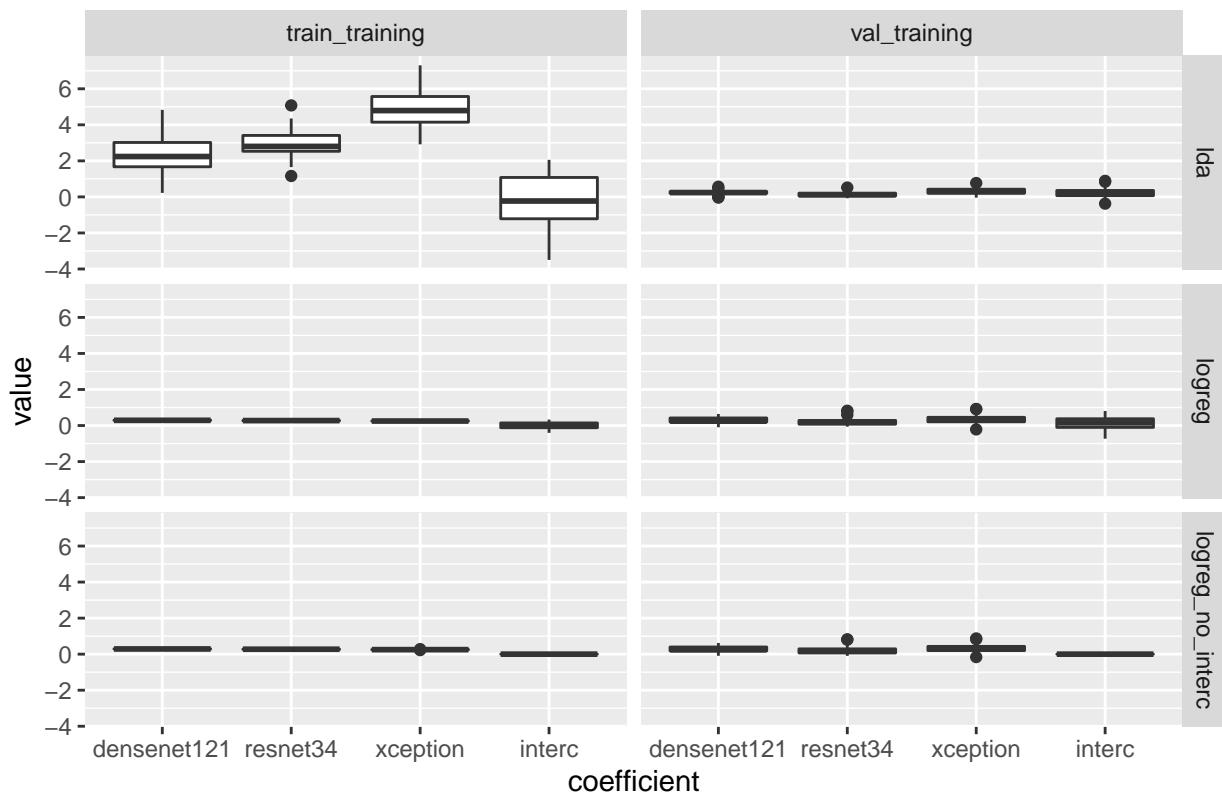
### Coefficients for class 3 vs 9



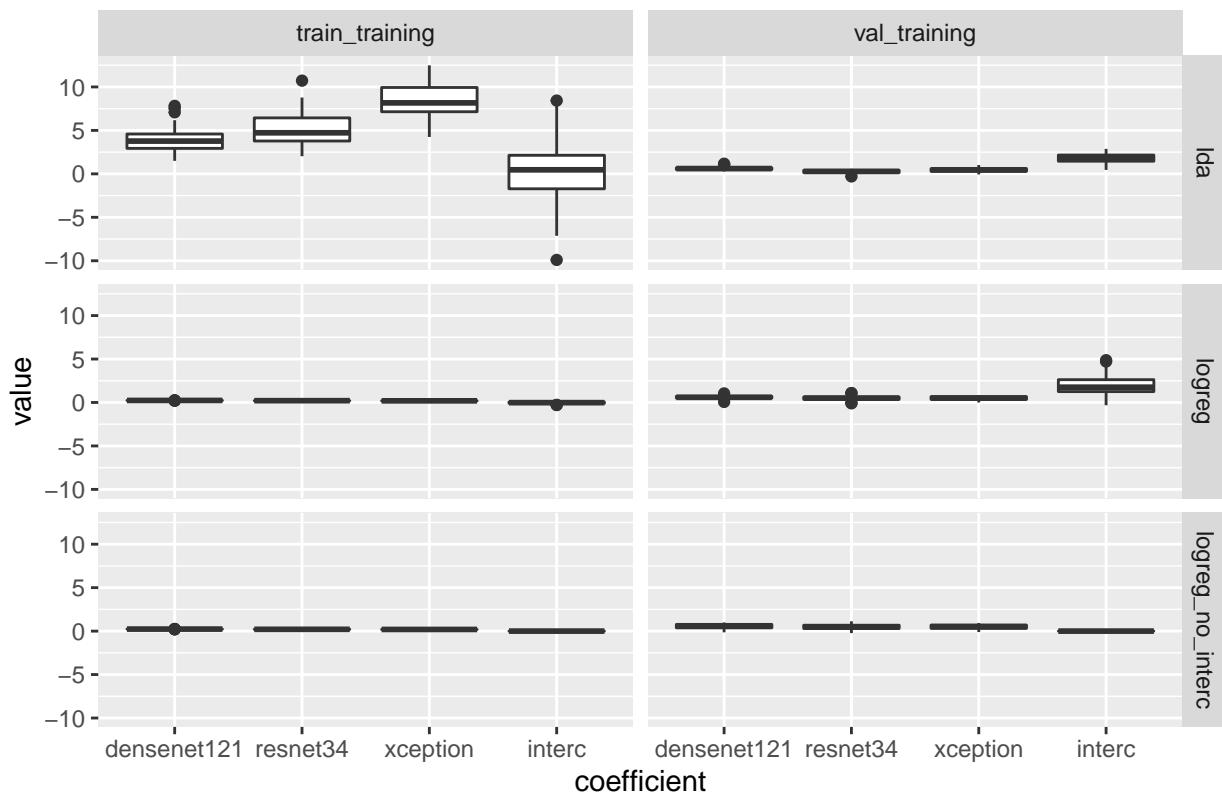
### Coefficients for class 3 vs 10



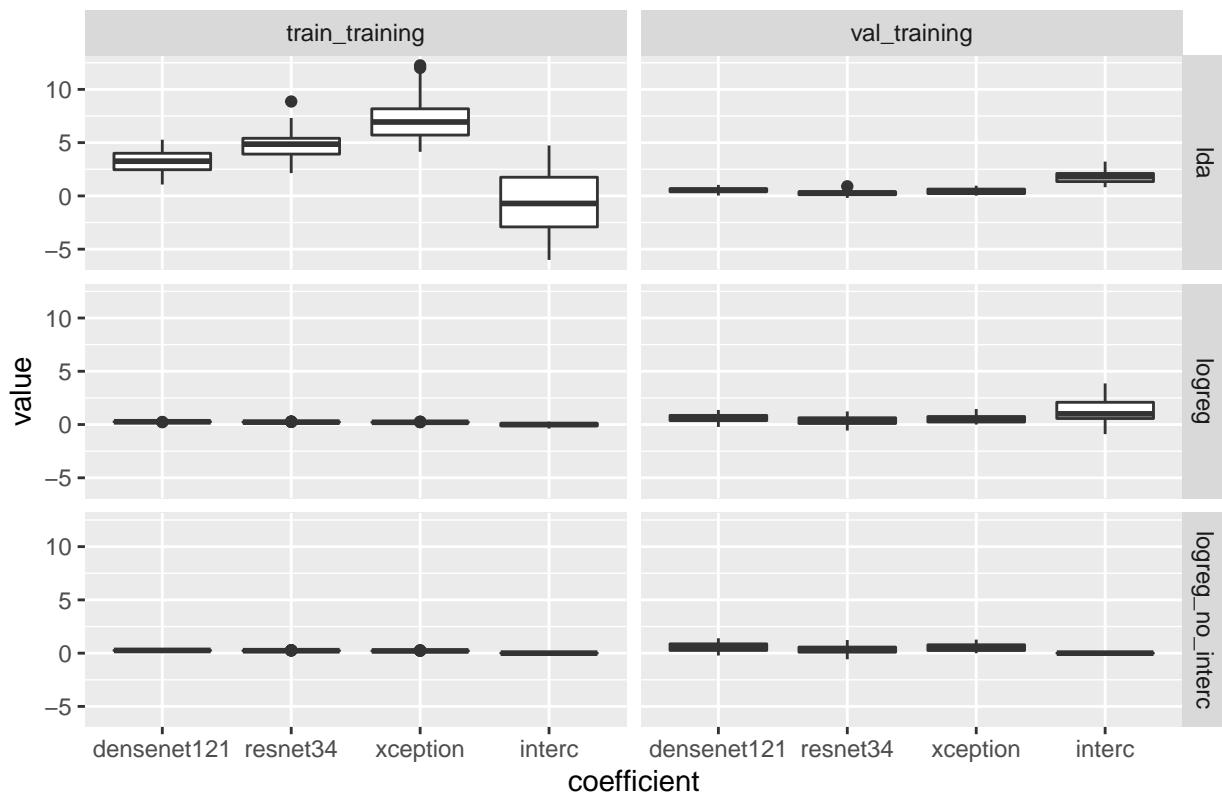
### Coefficients for class 4 vs 5



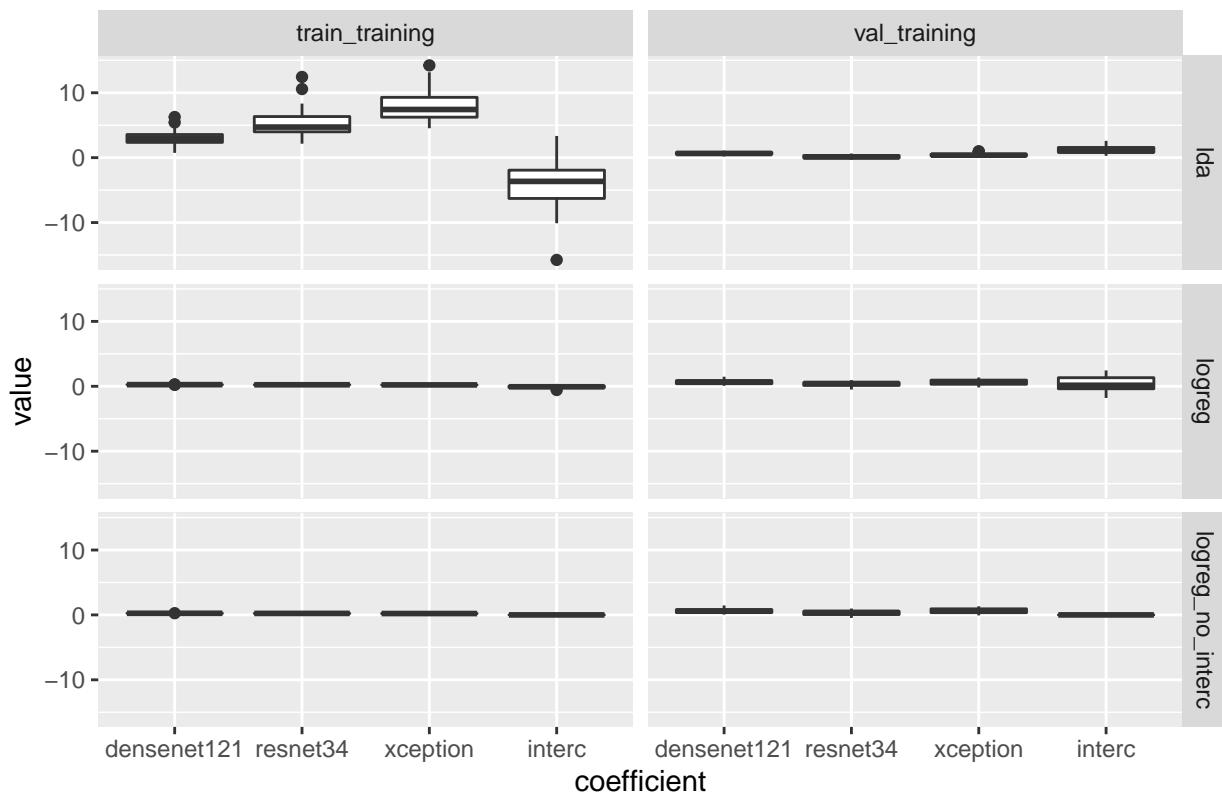
### Coefficients for class 4 vs 6



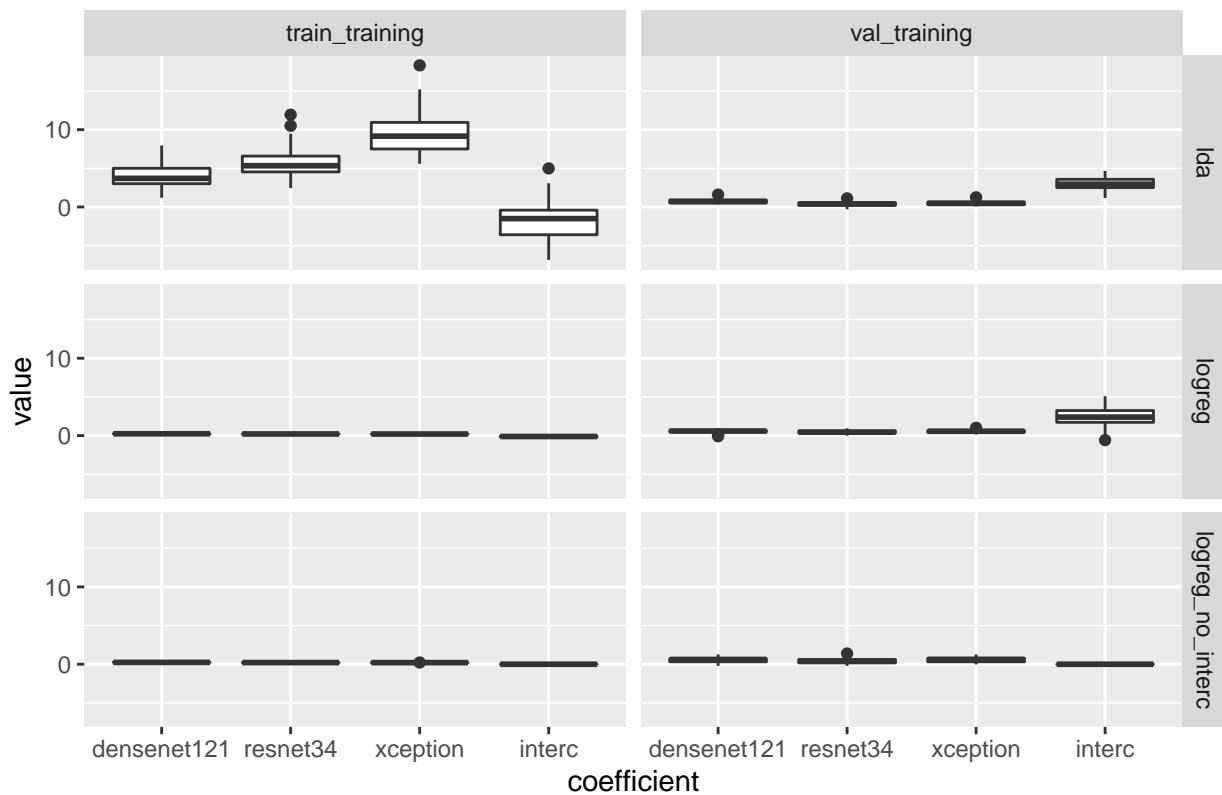
### Coefficients for class 4 vs 7



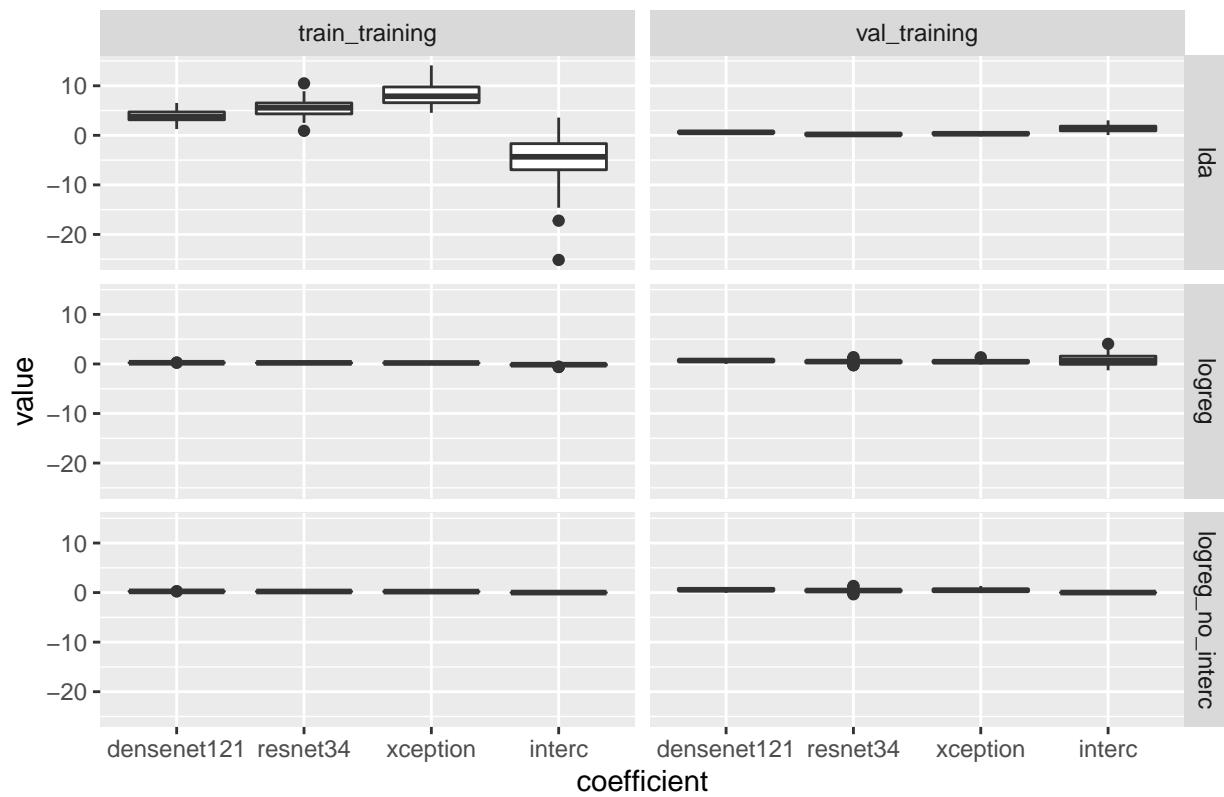
### Coefficients for class 4 vs 8



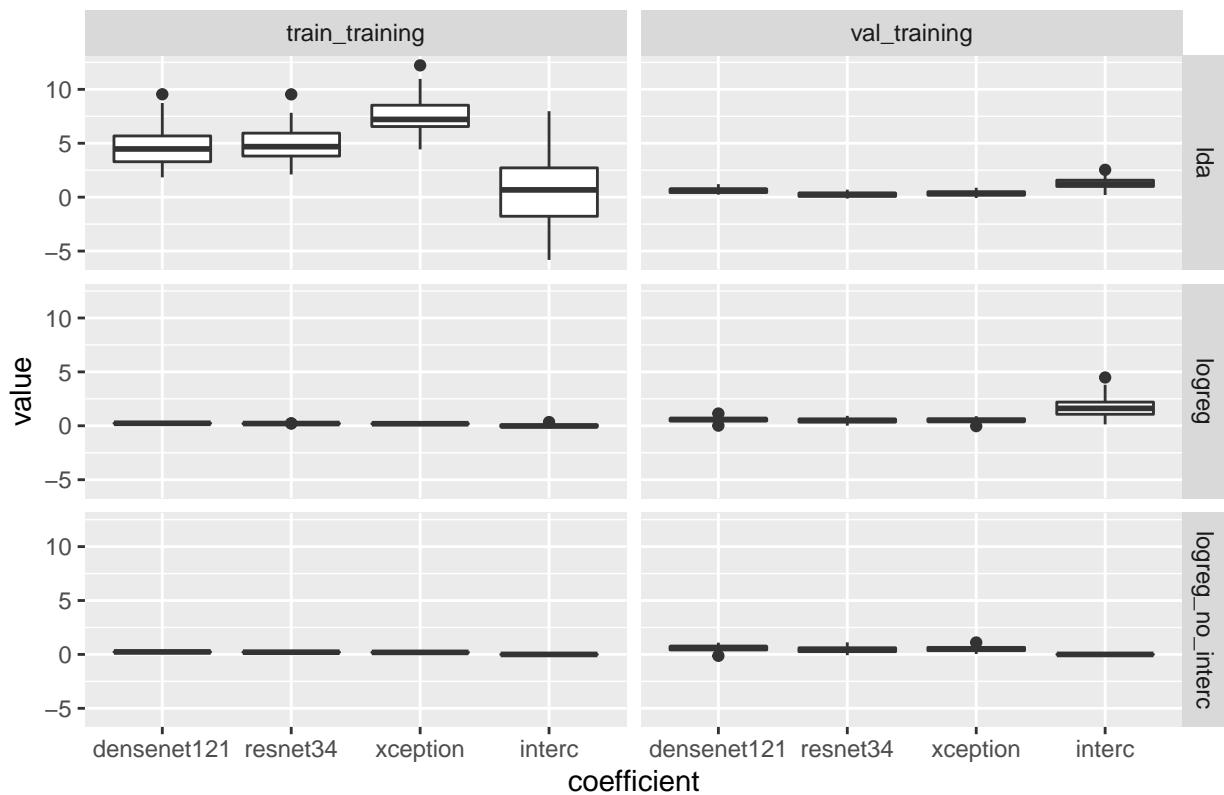
### Coefficients for class 4 vs 9



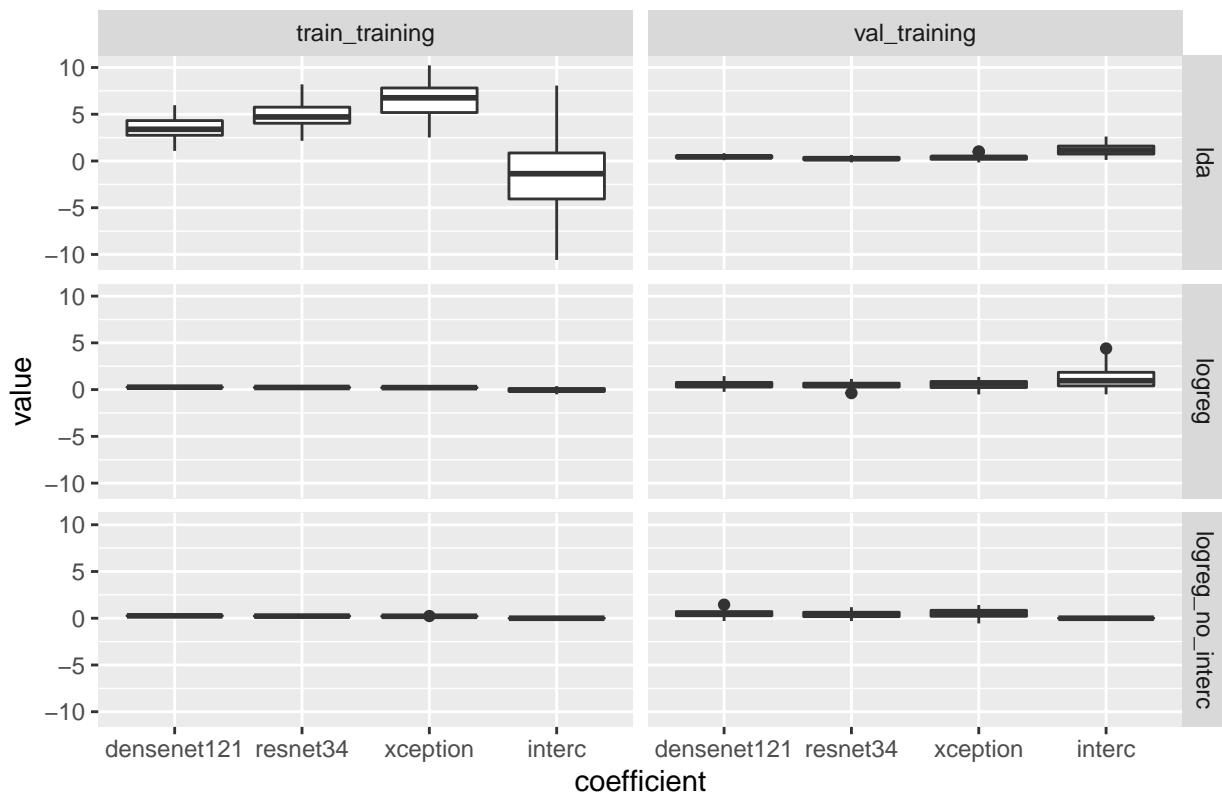
### Coefficients for class 4 vs 10



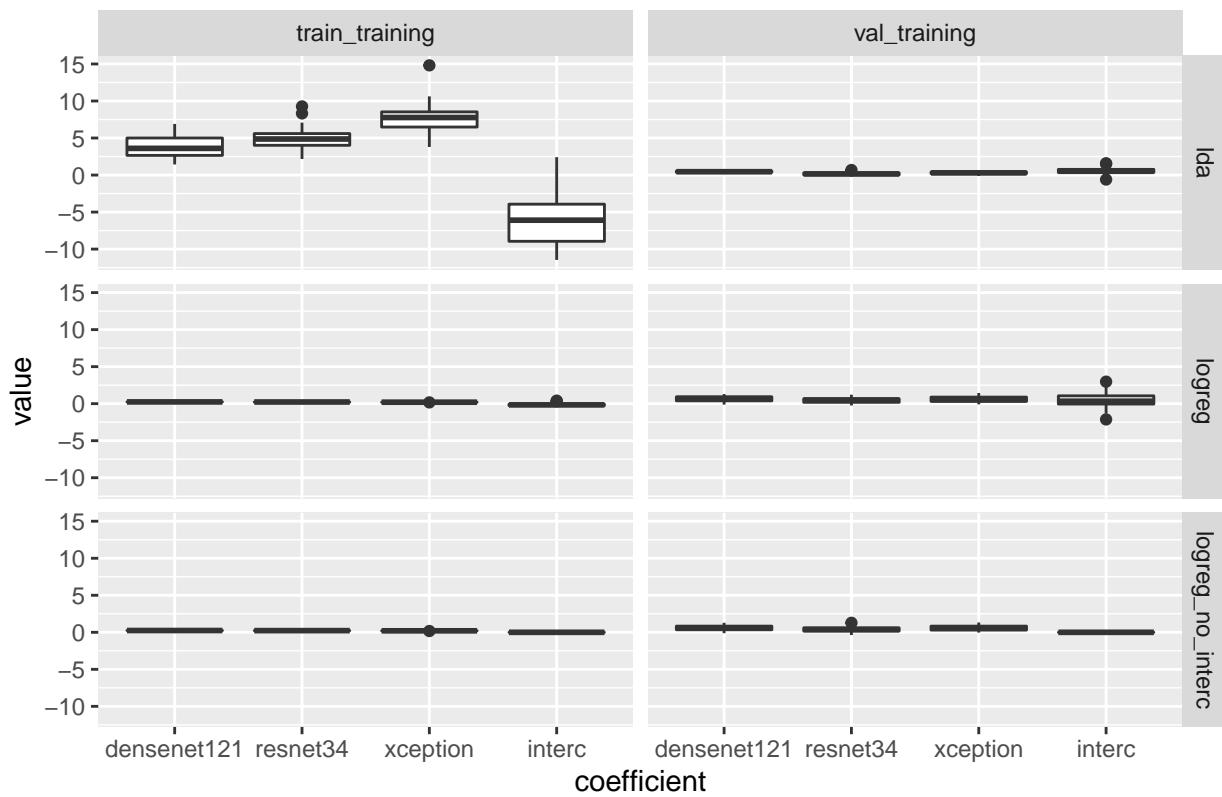
### Coefficients for class 5 vs 6



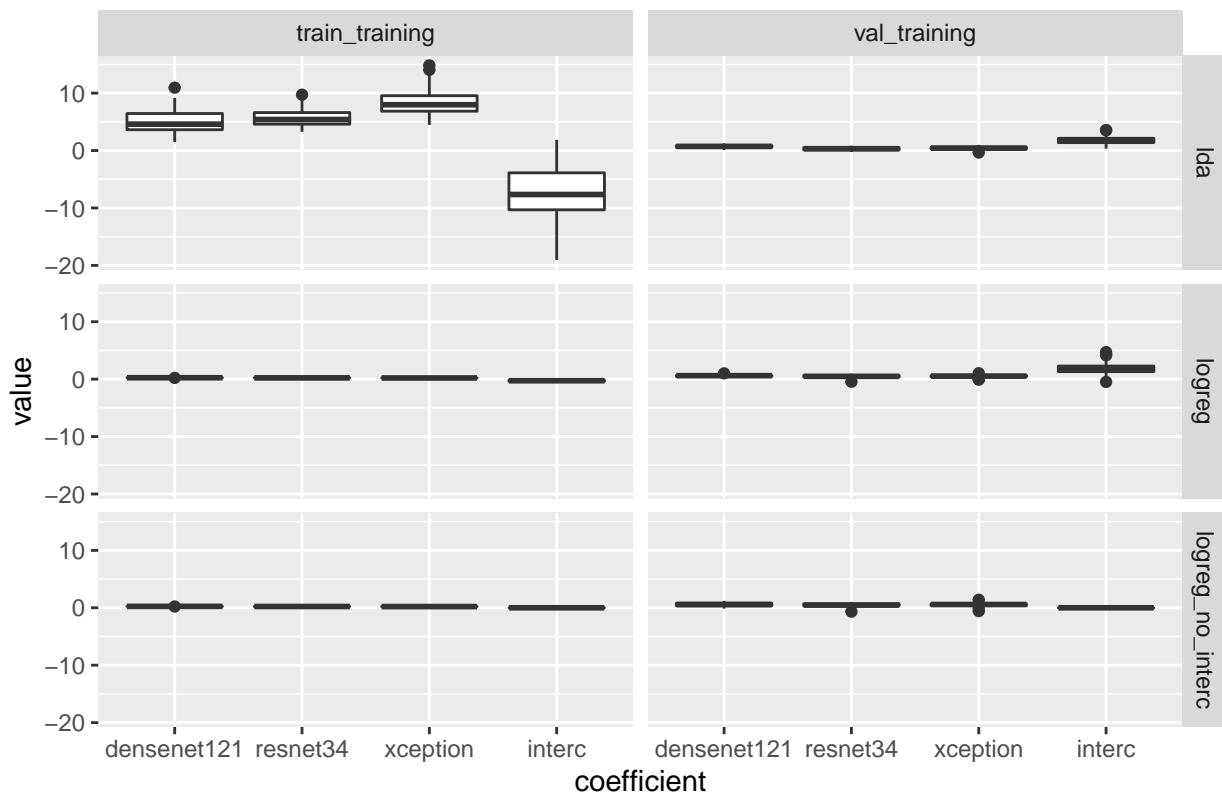
### Coefficients for class 5 vs 7



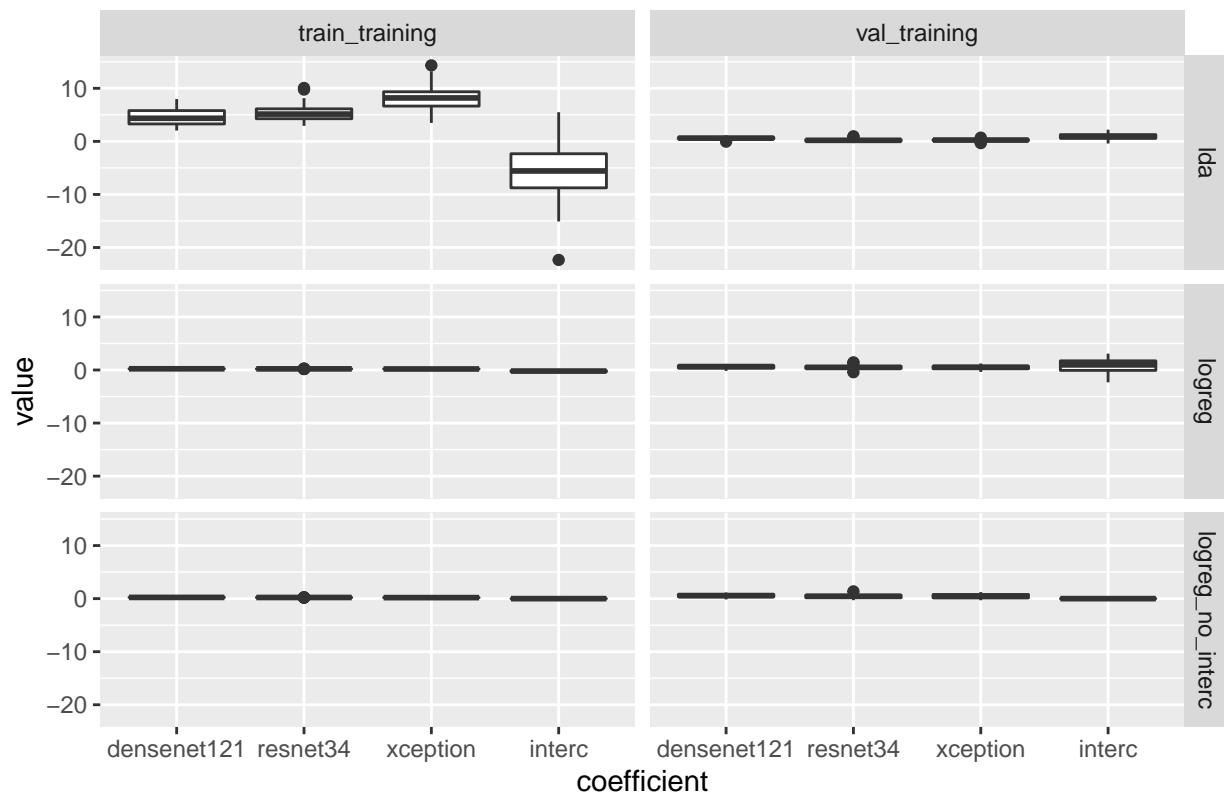
### Coefficients for class 5 vs 8



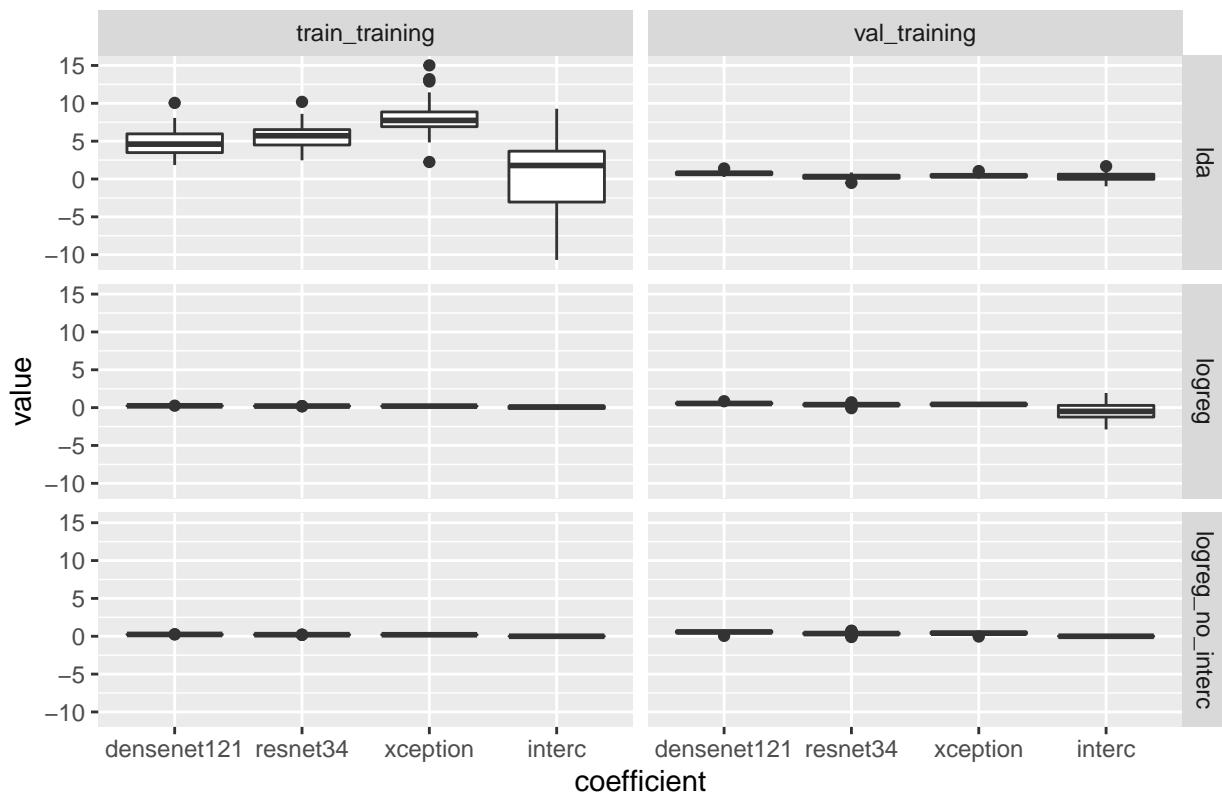
### Coefficients for class 5 vs 9



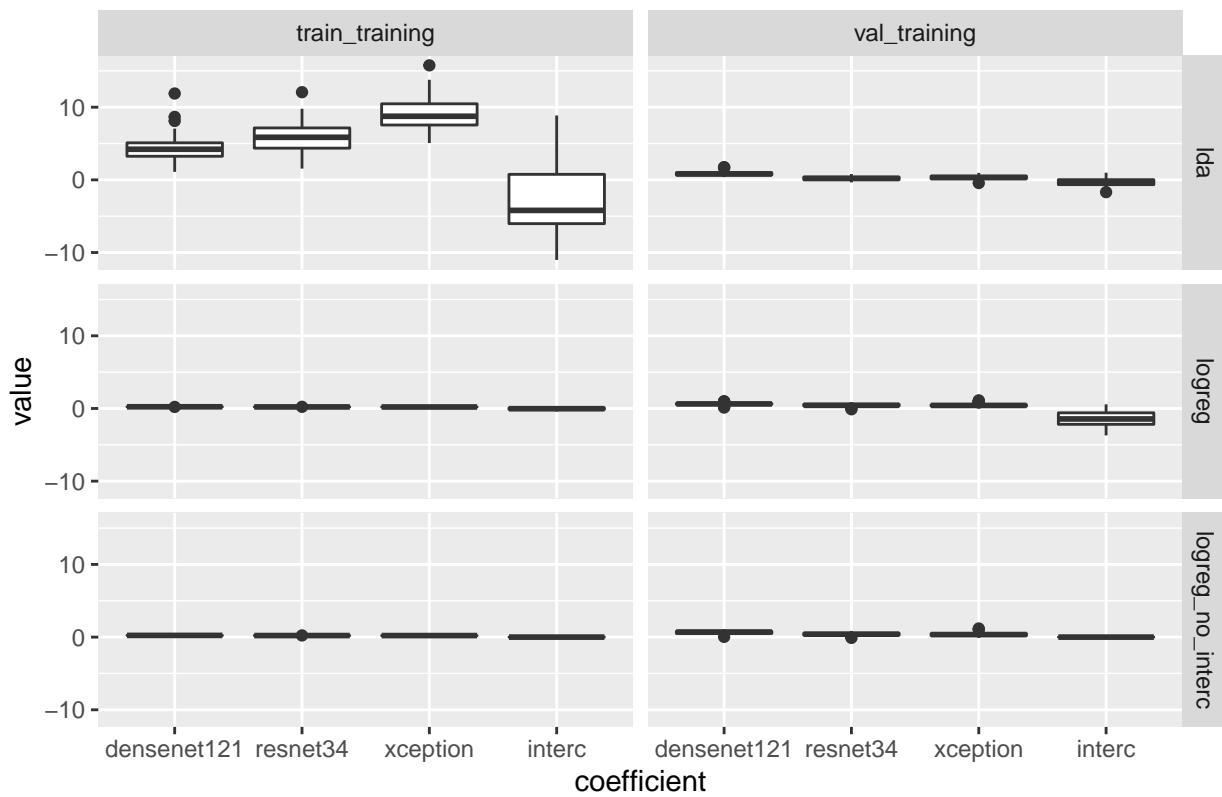
### Coefficients for class 5 vs 10



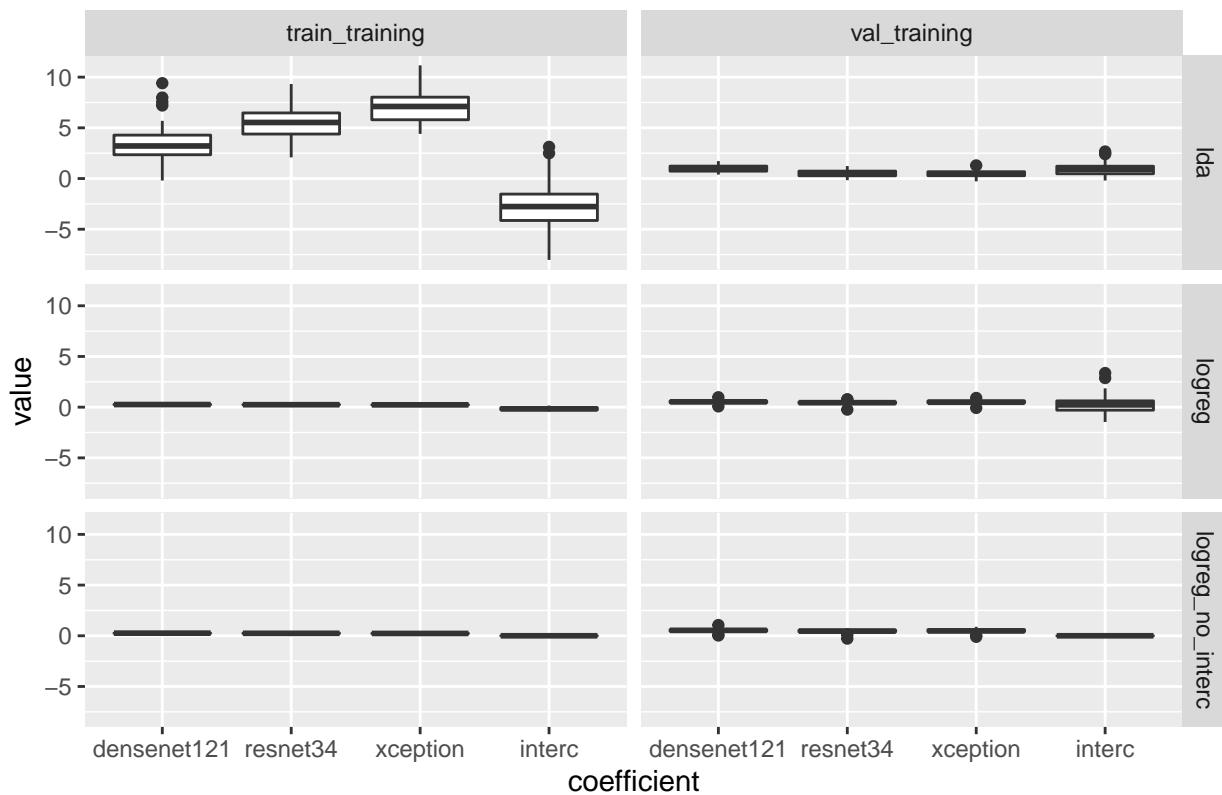
### Coefficients for class 6 vs 7



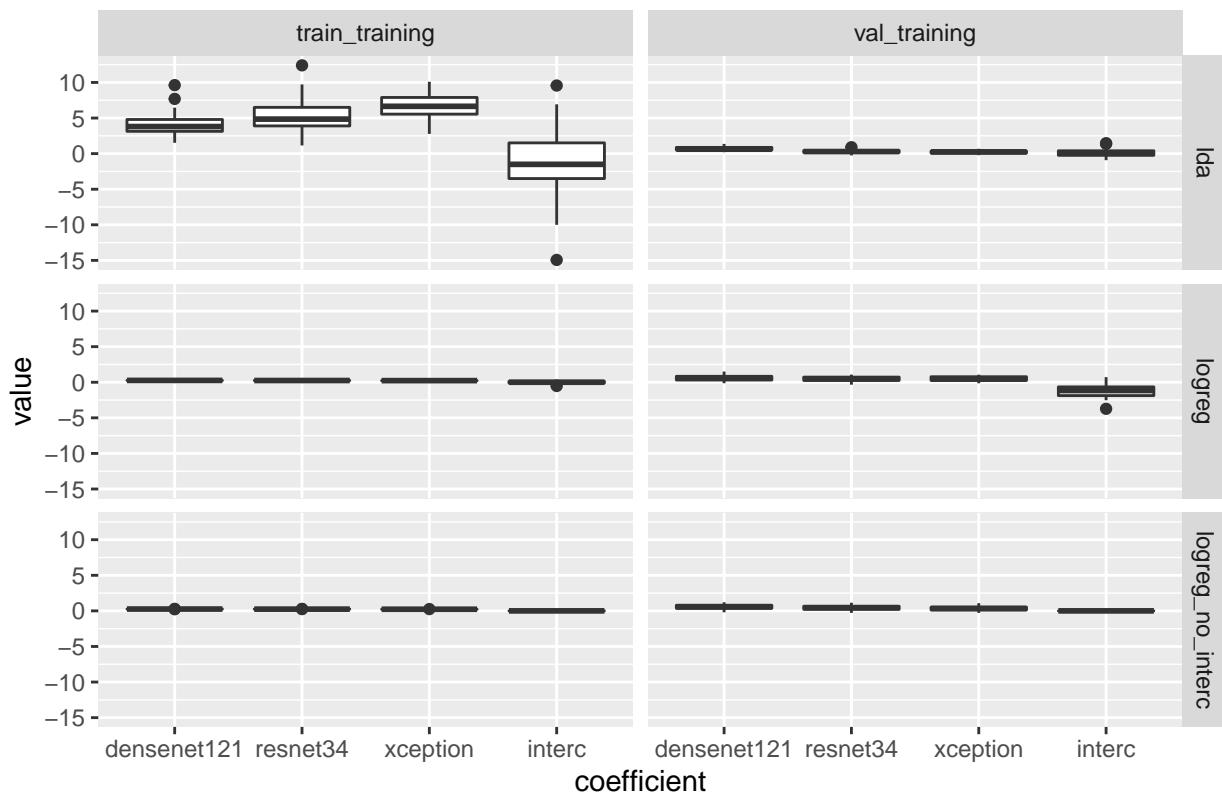
## Coefficients for class 6 vs 8



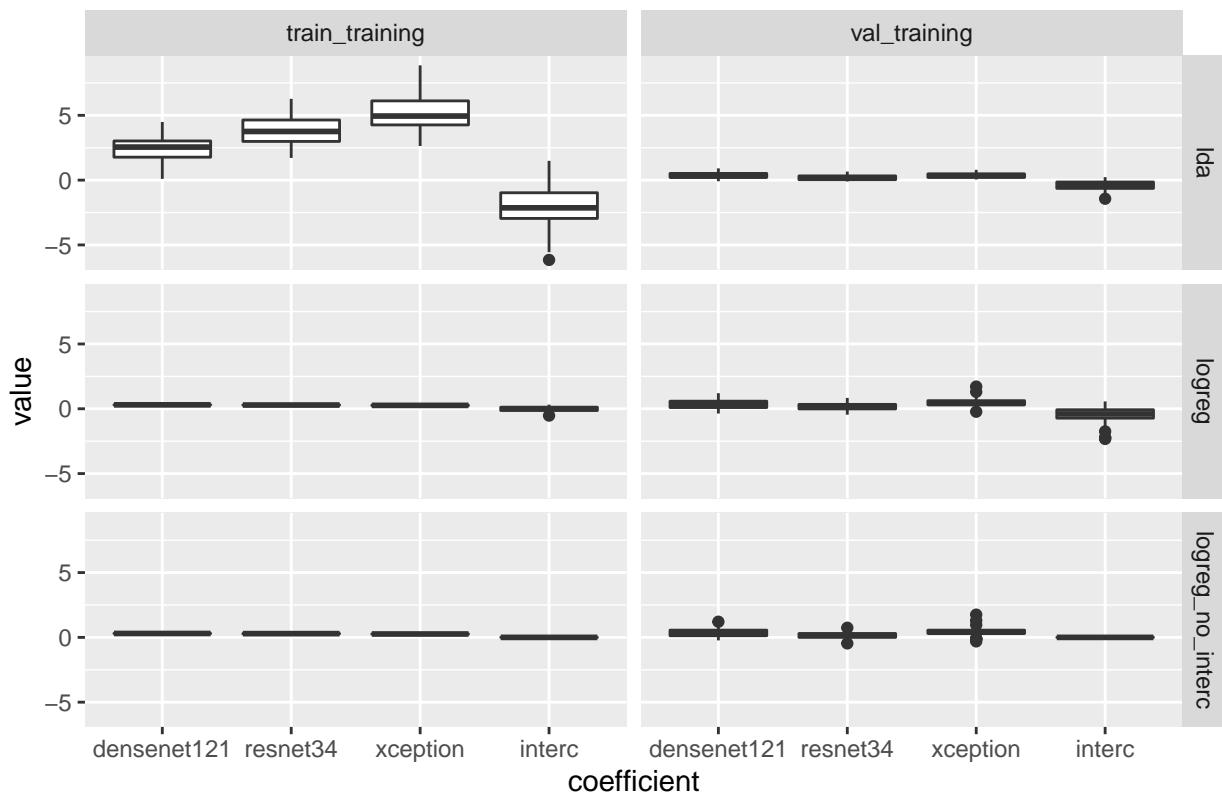
### Coefficients for class 6 vs 9



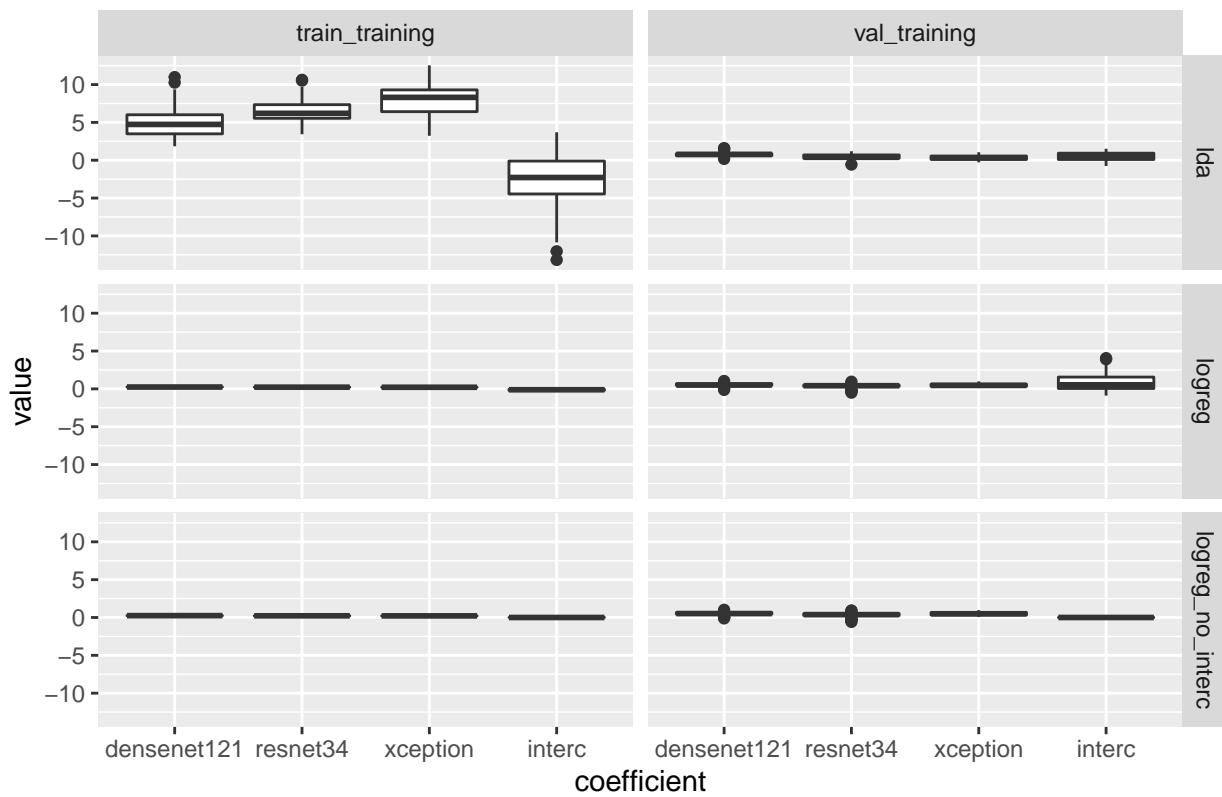
### Coefficients for class 6 vs 10



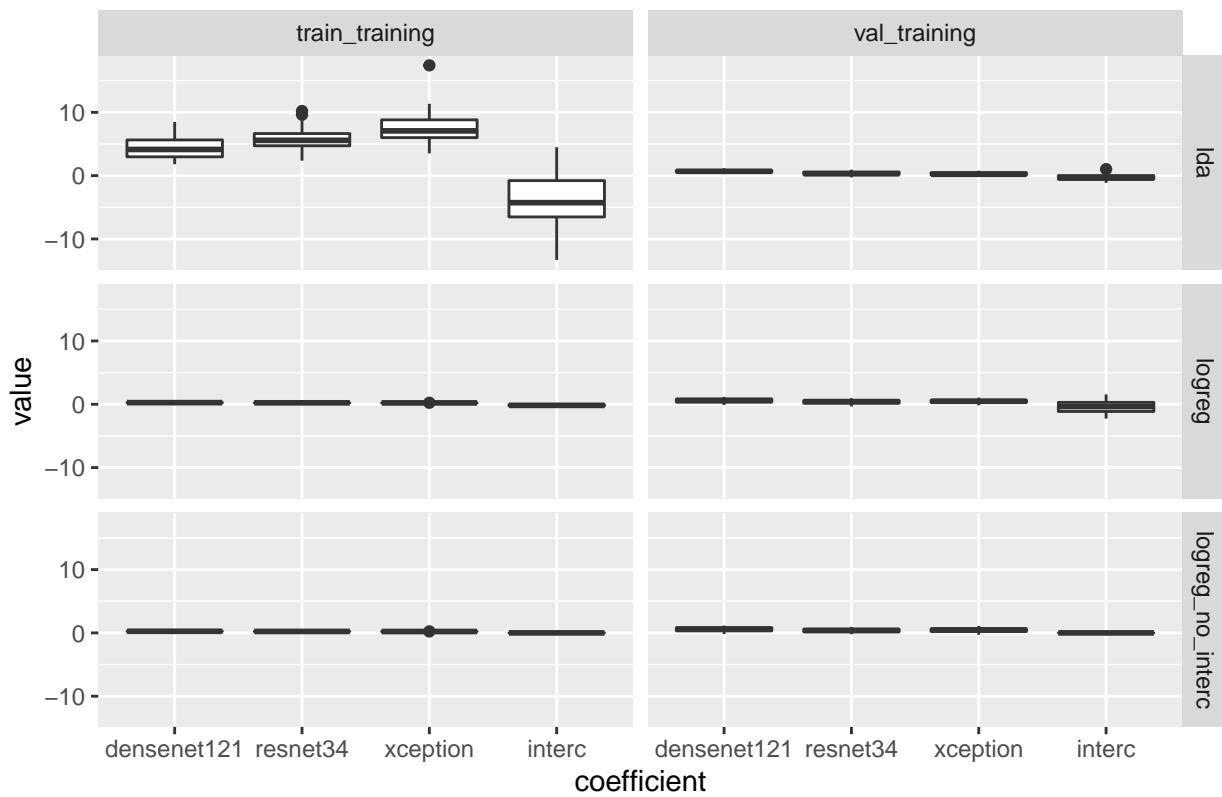
### Coefficients for class 7 vs 8



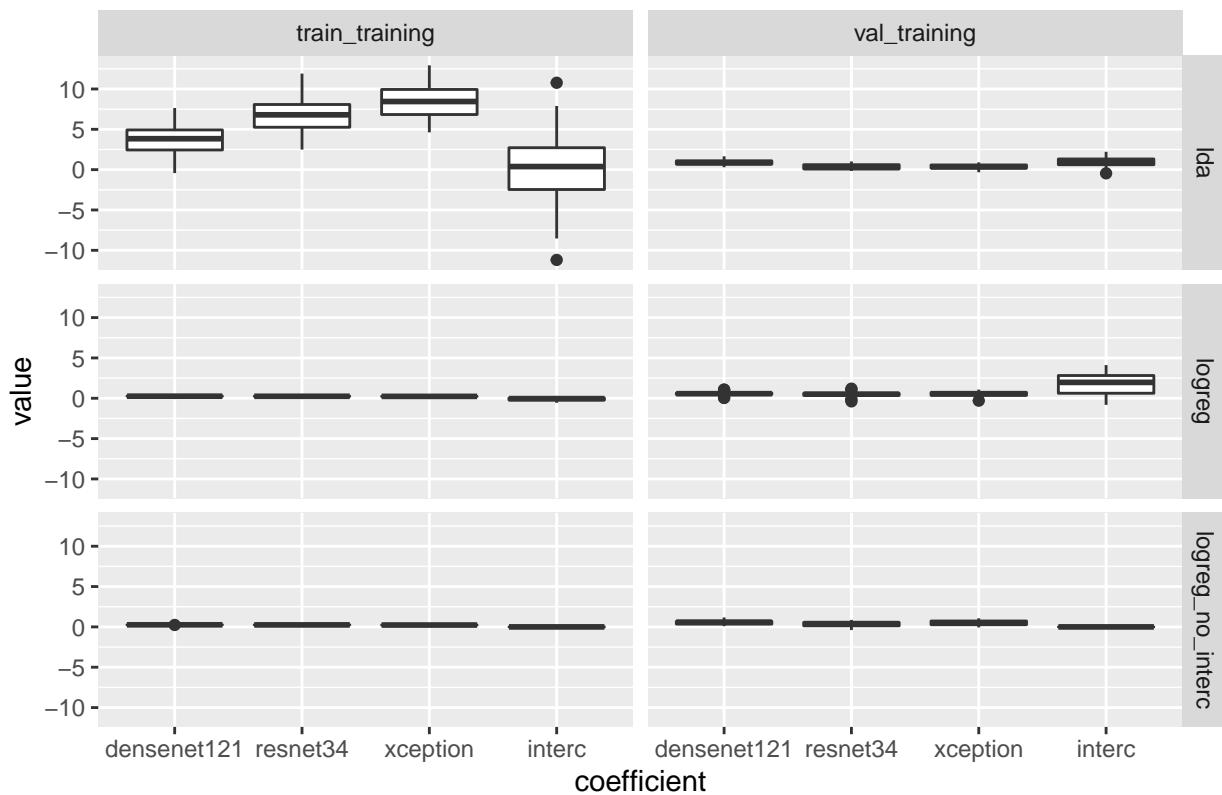
### Coefficients for class 7 vs 9



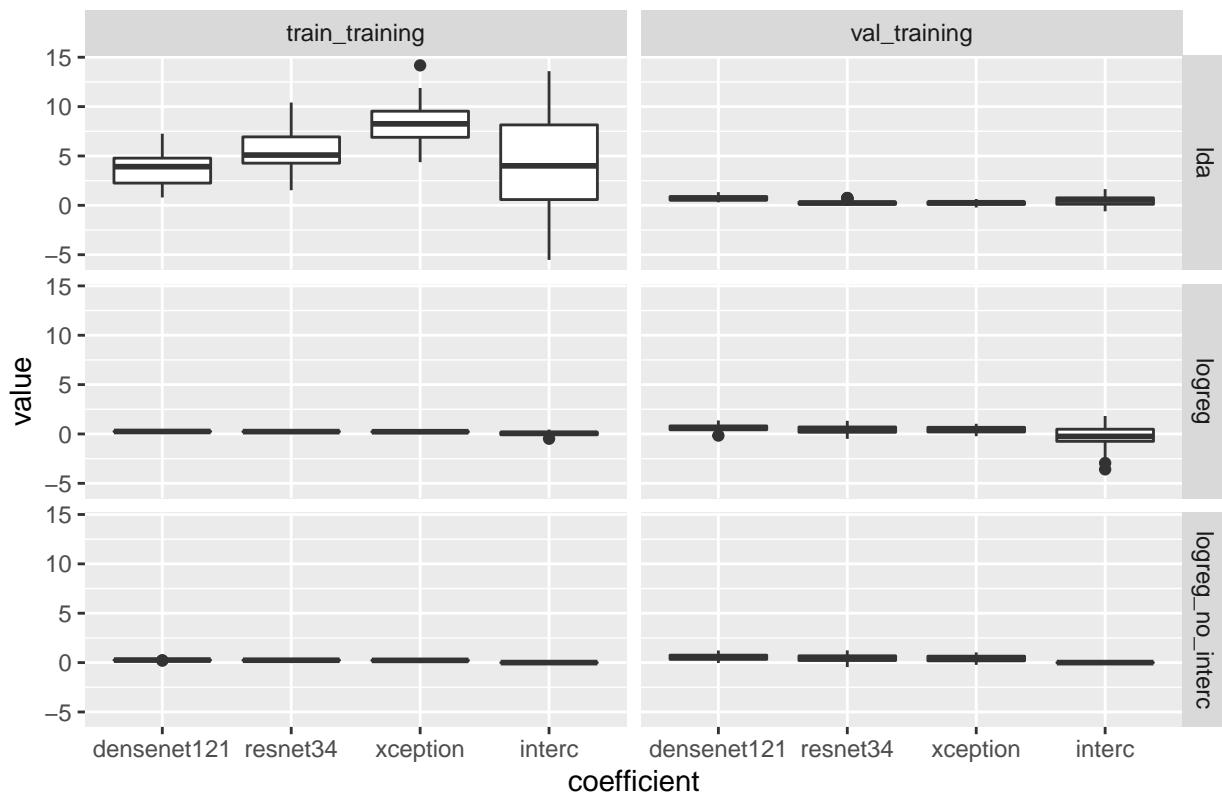
## Coefficients for class 7 vs 10



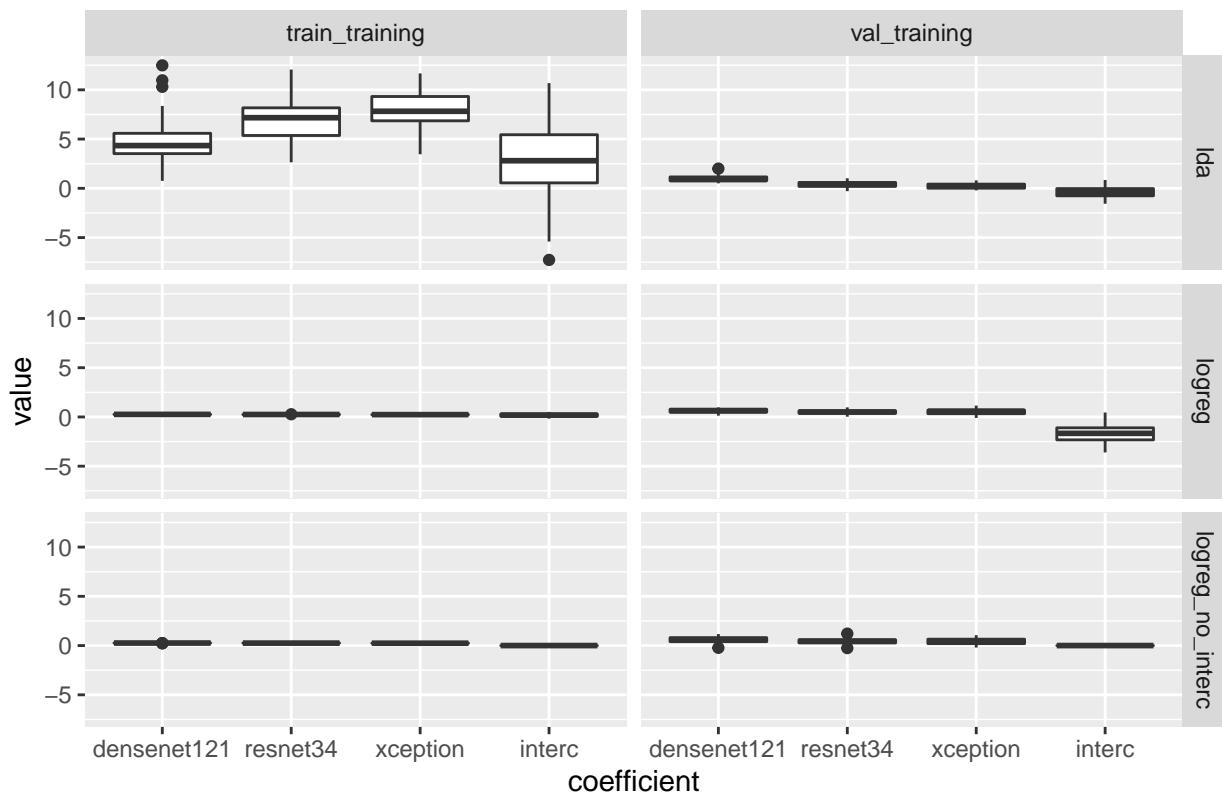
### Coefficients for class 8 vs 9



### Coefficients for class 8 vs 10



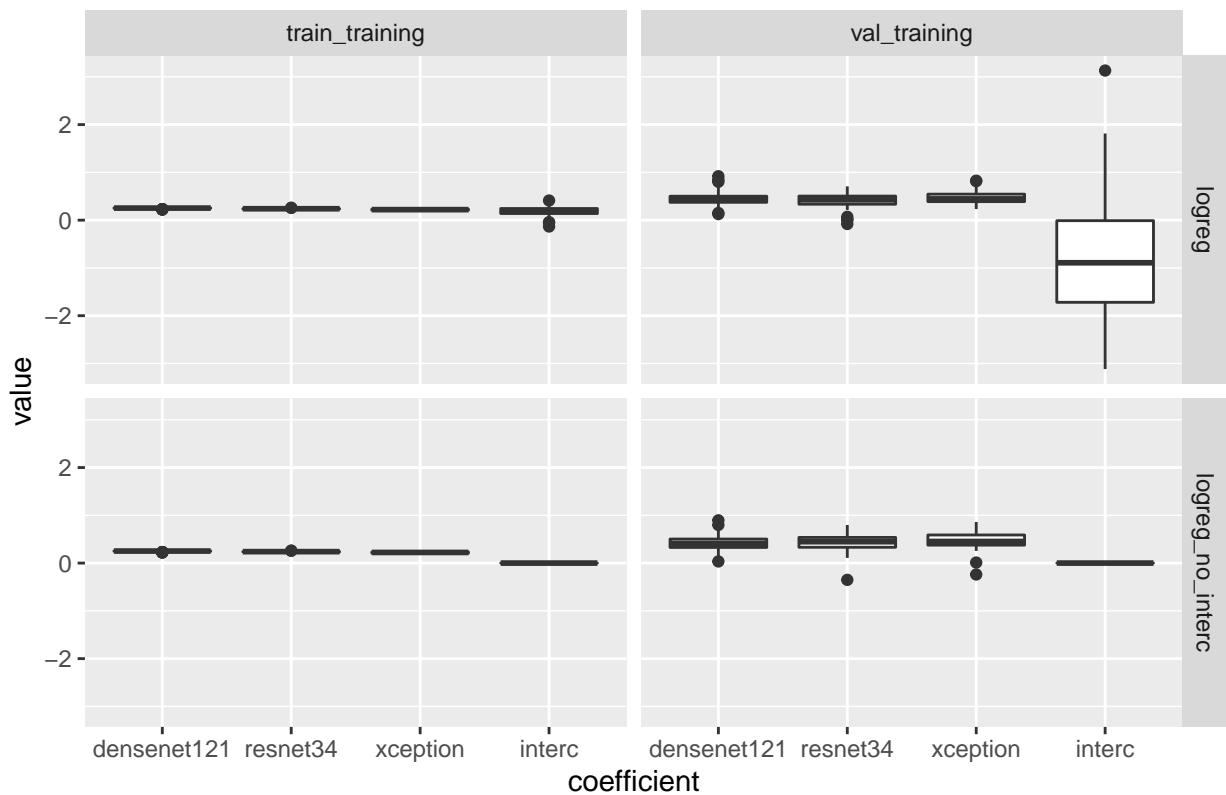
## Coefficients for class 9 vs 10



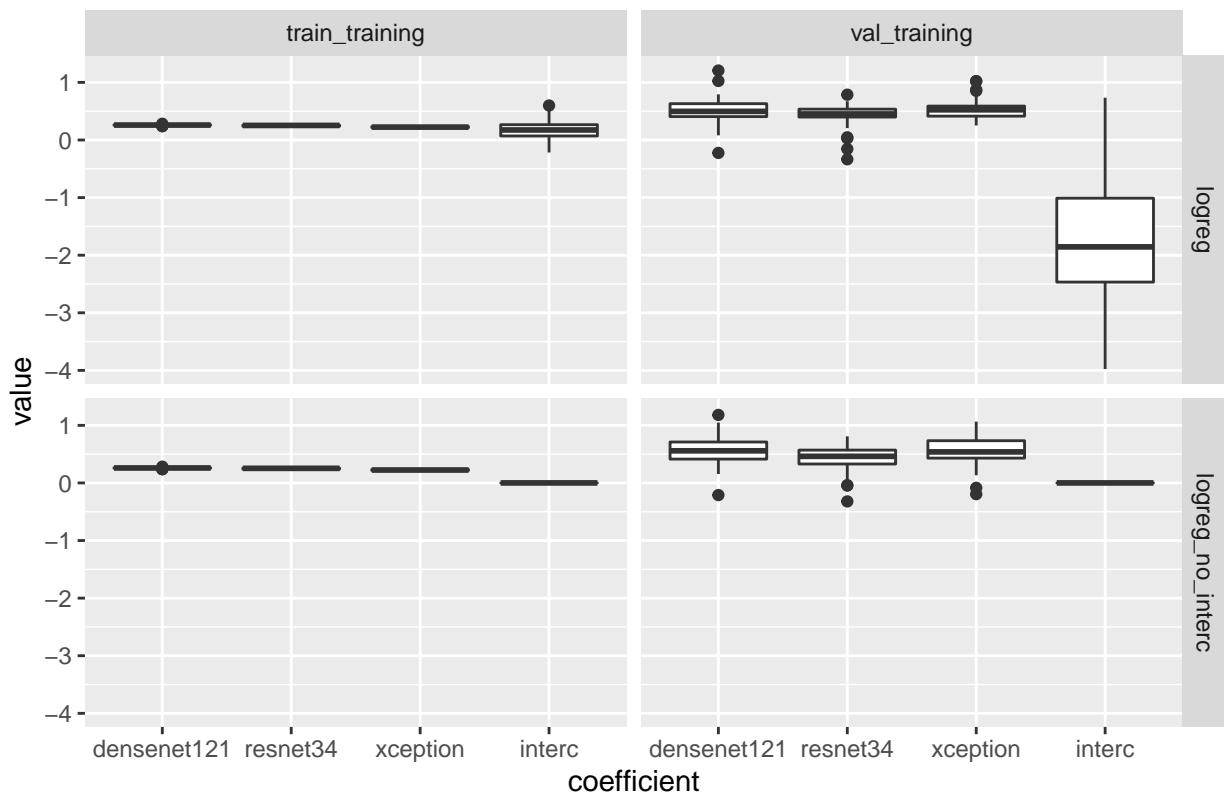
```

for (cl1 in 1:(10 - 1))
{
  for (cl2 in (cl1 + 1):10)
  {
    cur_plt <- combiner_coefs %>% filter(class1 == cl1 & class2 == cl2 & combining_method!="lda") %>% ggtitle(paste("Coefficients for class", cl1, "vs", cl2))
    facet_grid(cols=vars(train_type), rows=vars(combining_method)) + print(cur_plt)
  }
}
  
```

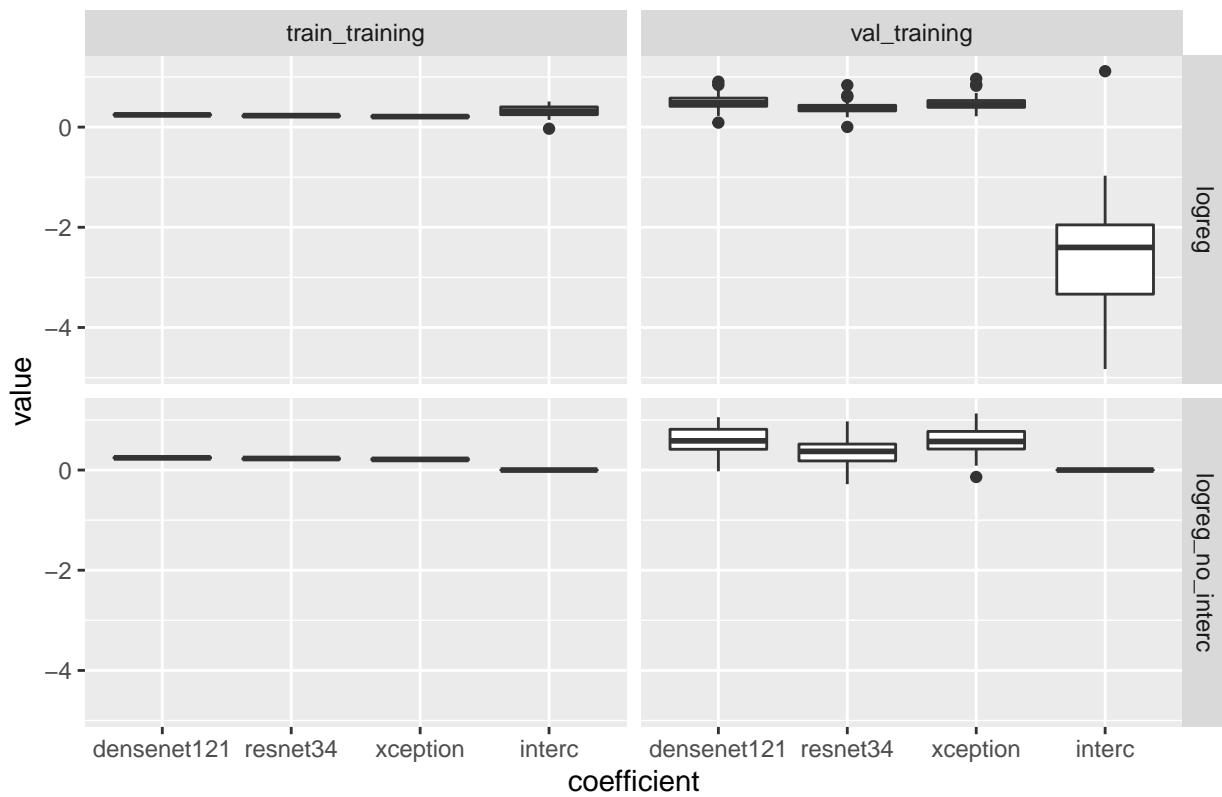
## Coefficients for class 1 vs 2



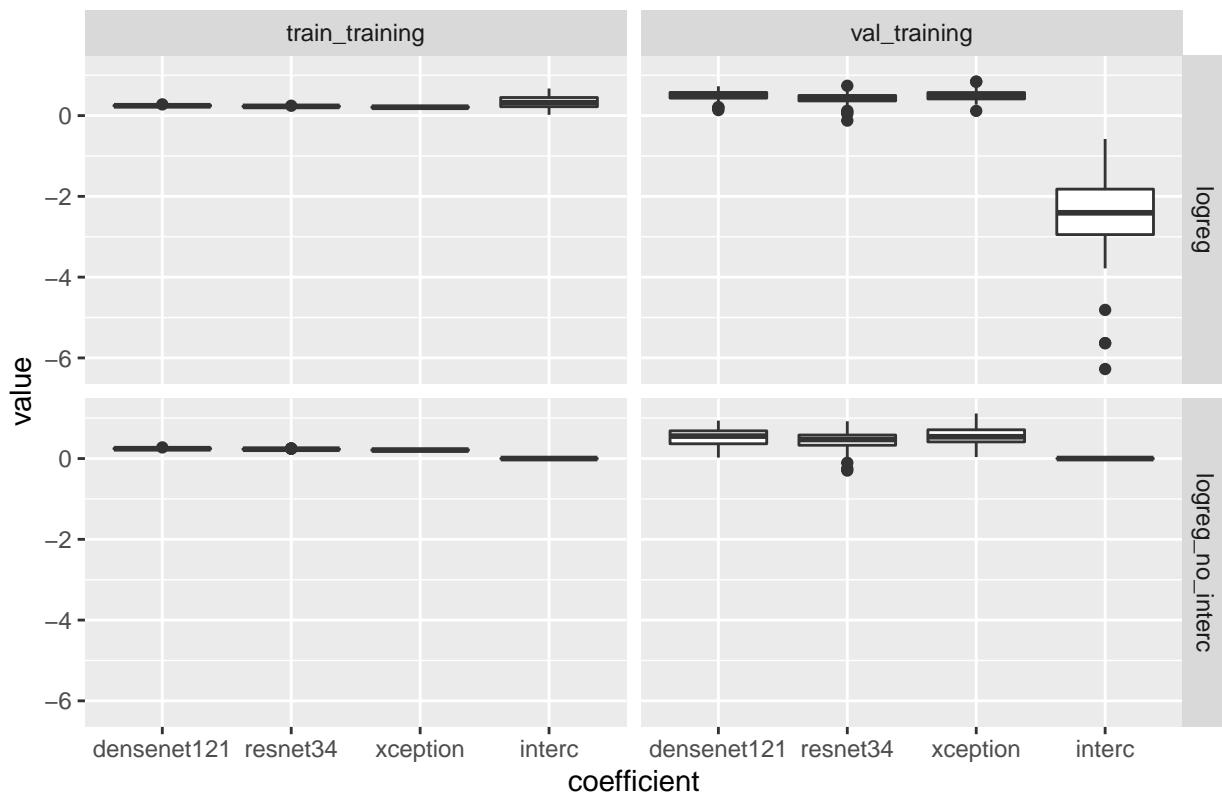
### Coefficients for class 1 vs 3



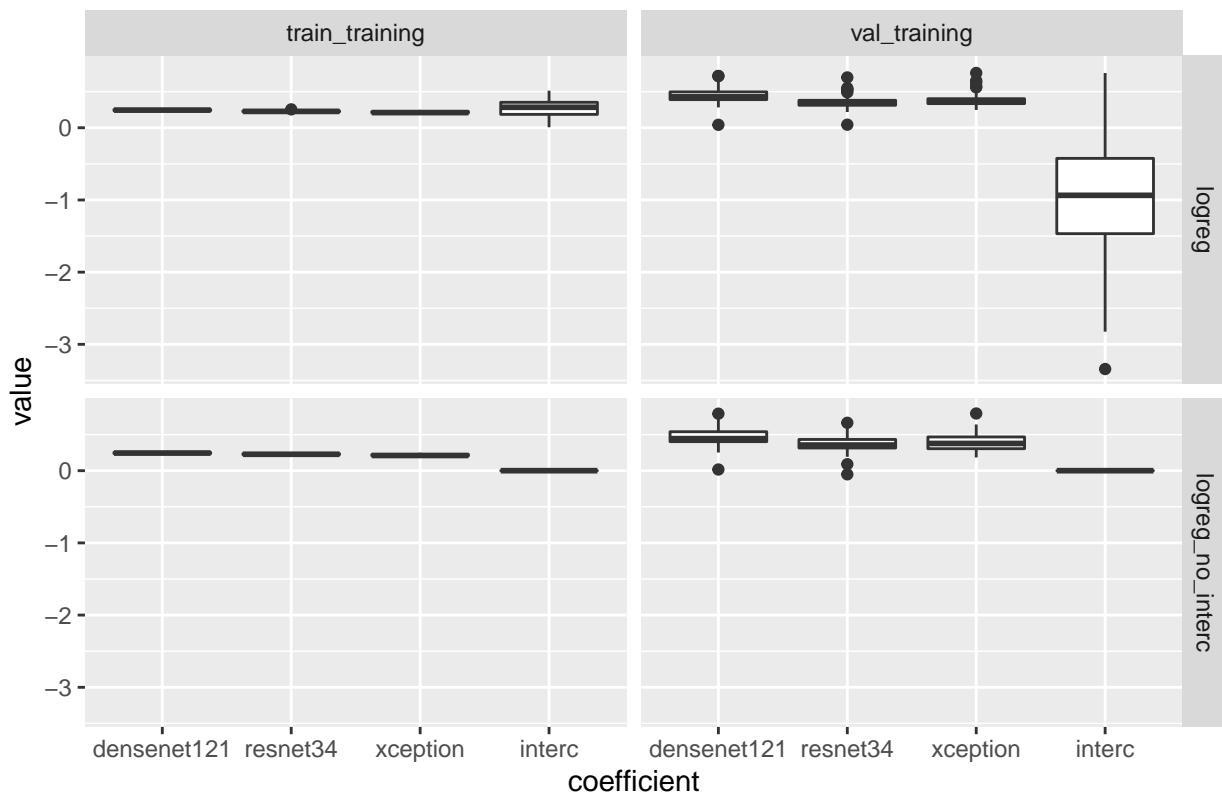
### Coefficients for class 1 vs 4



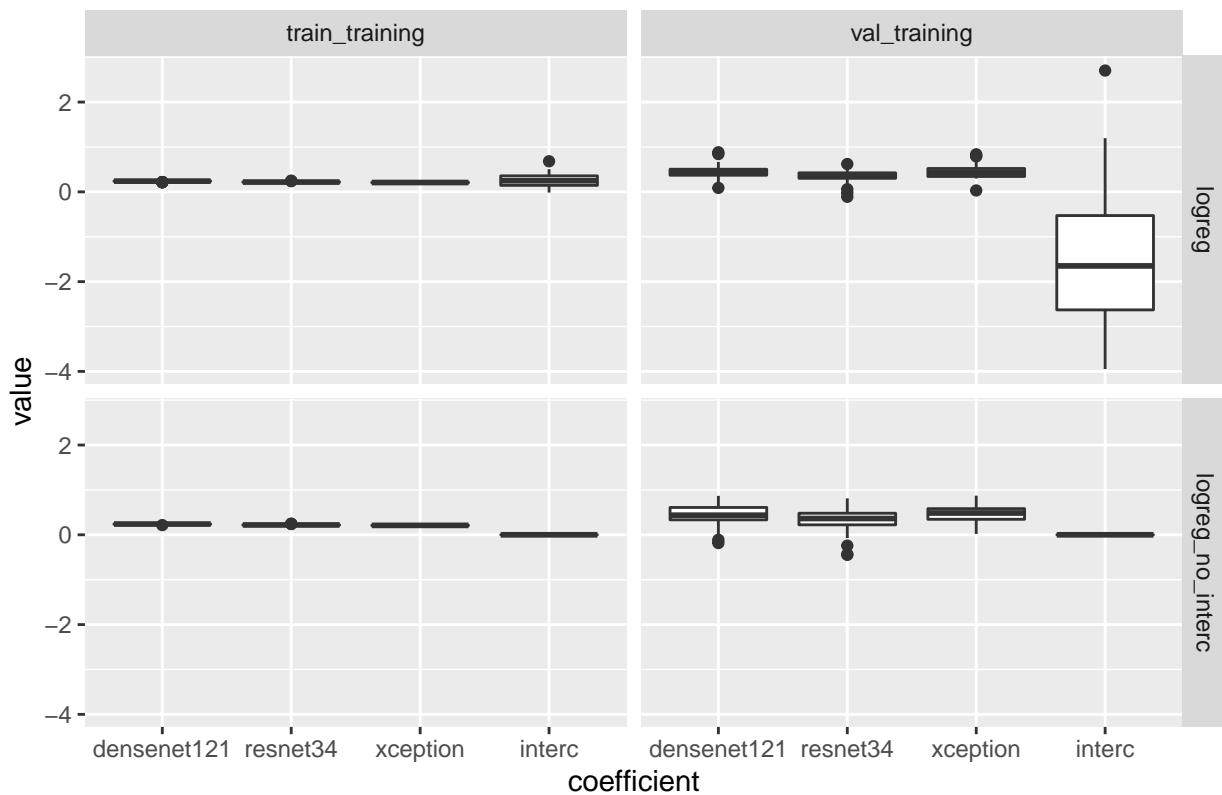
### Coefficients for class 1 vs 5



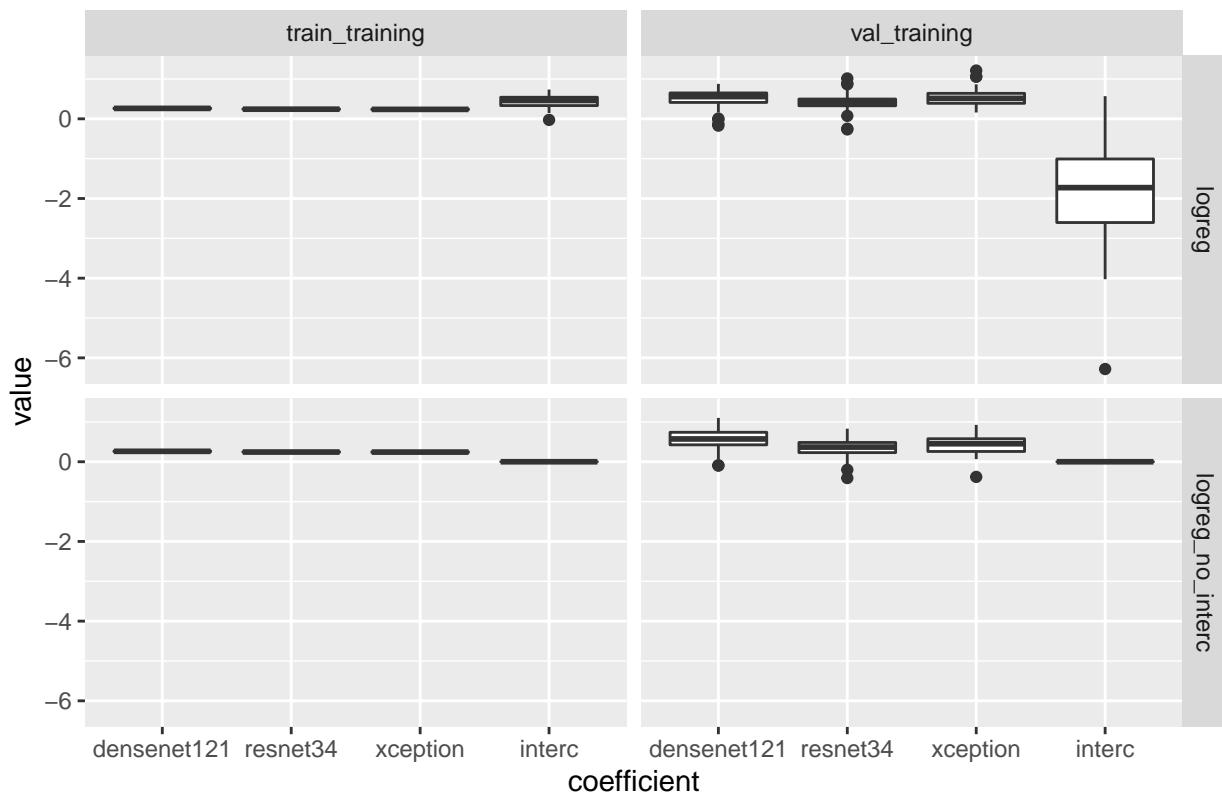
### Coefficients for class 1 vs 6



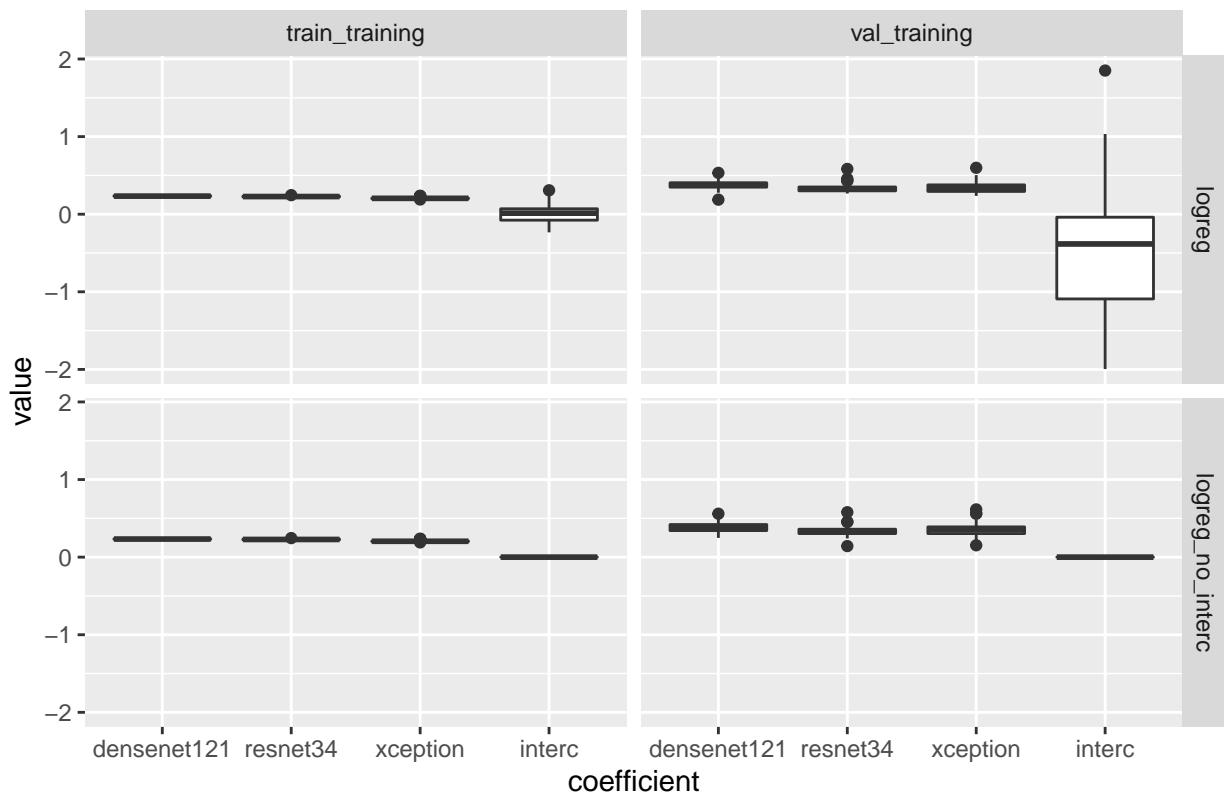
### Coefficients for class 1 vs 7



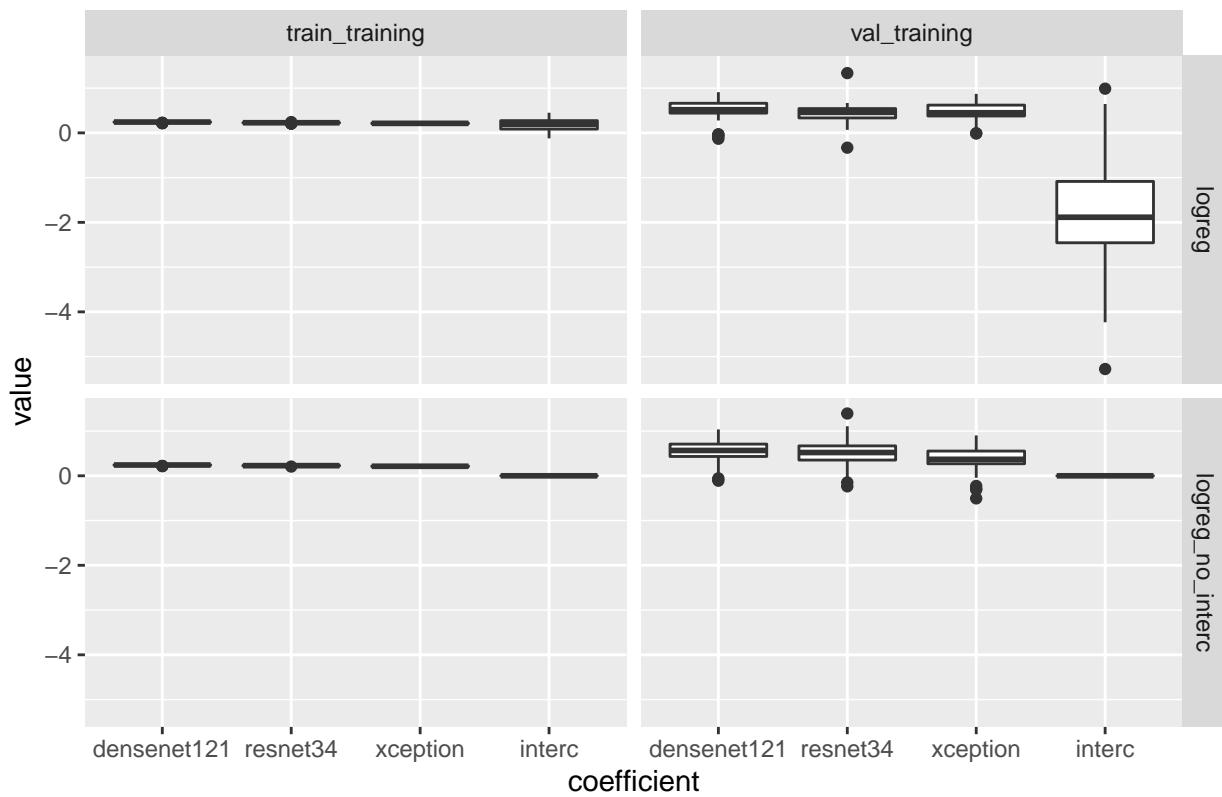
### Coefficients for class 1 vs 8



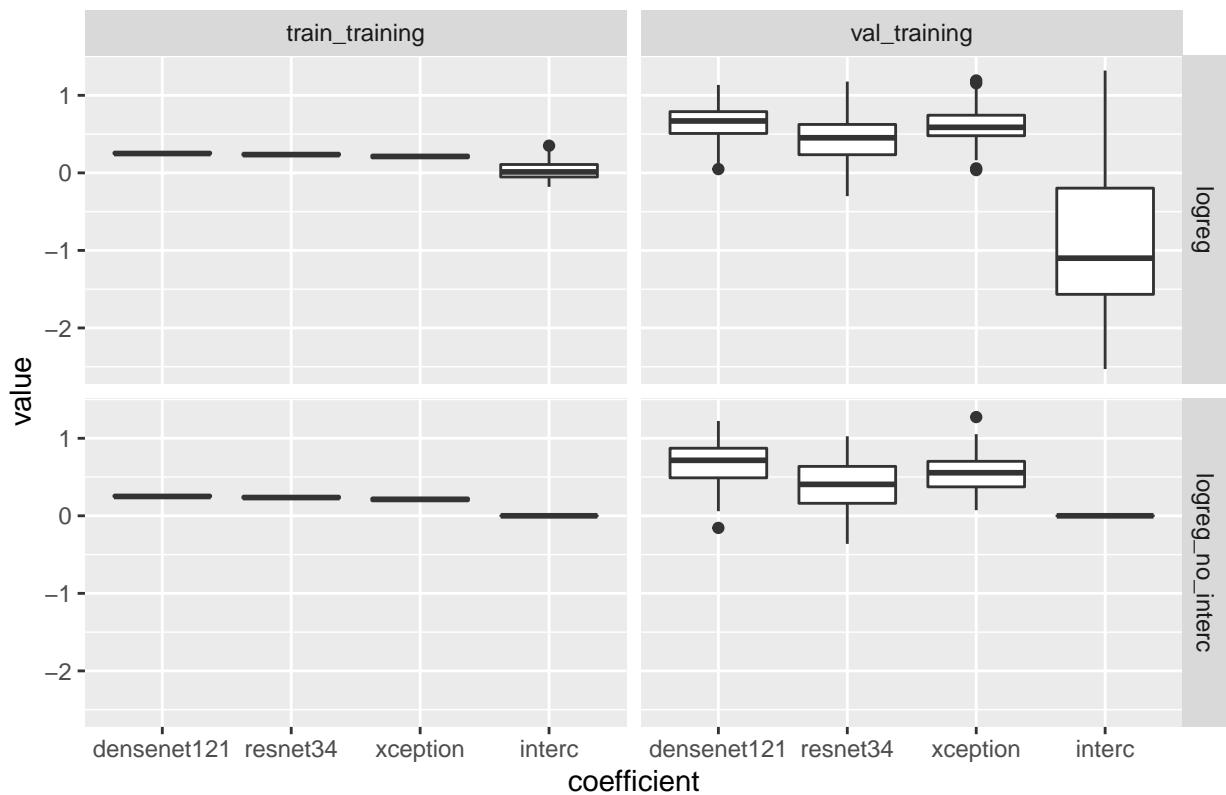
### Coefficients for class 1 vs 9



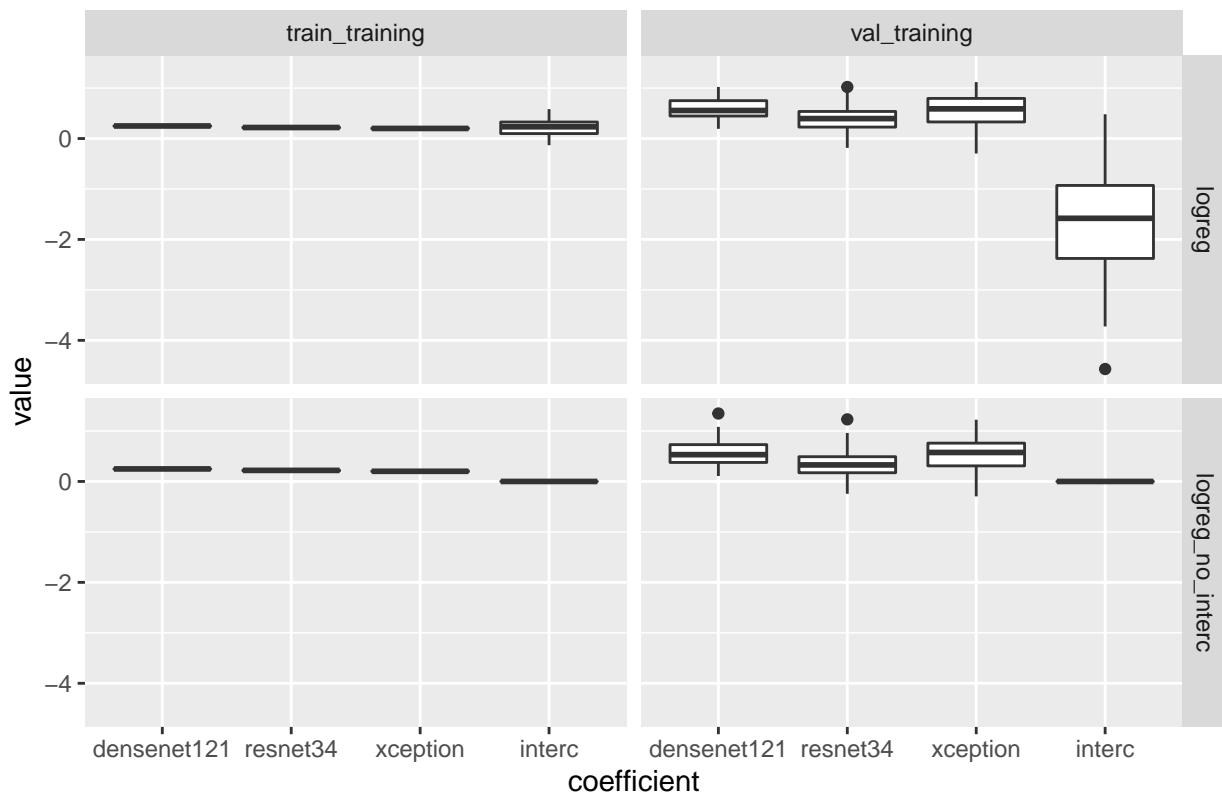
### Coefficients for class 1 vs 10



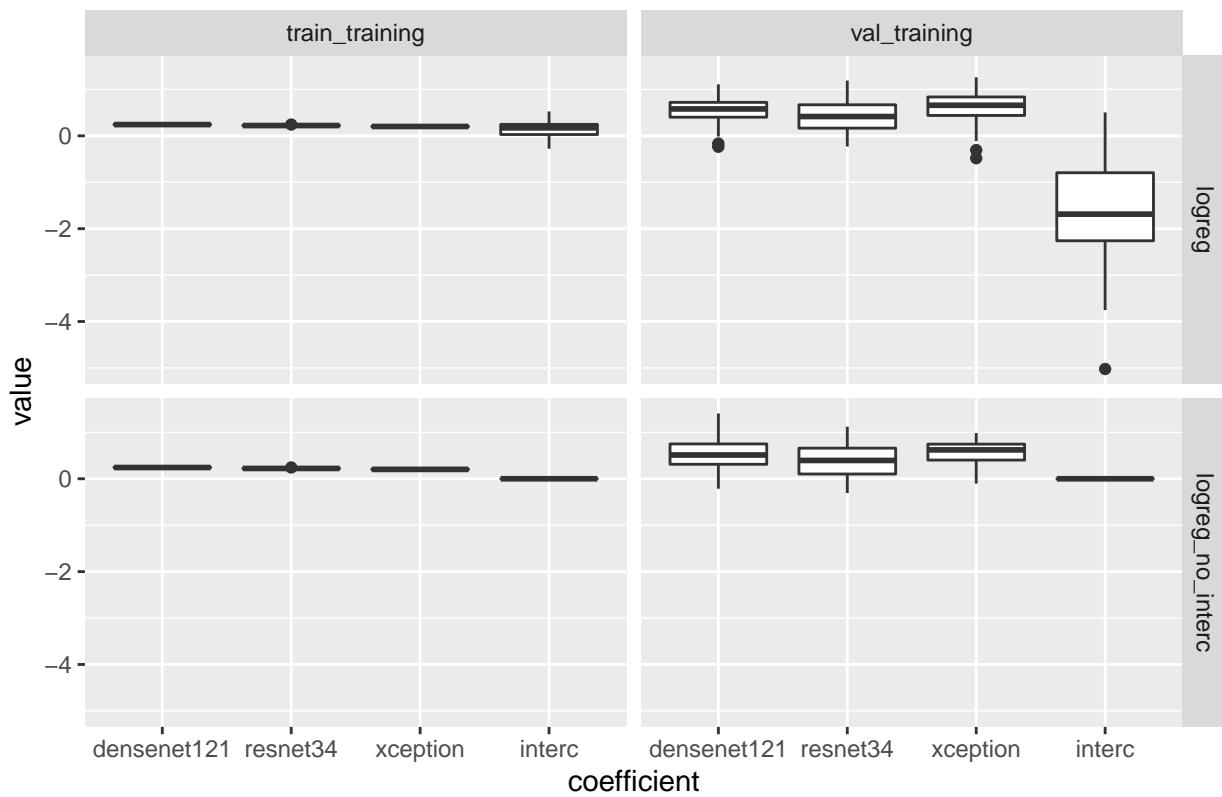
### Coefficients for class 2 vs 3



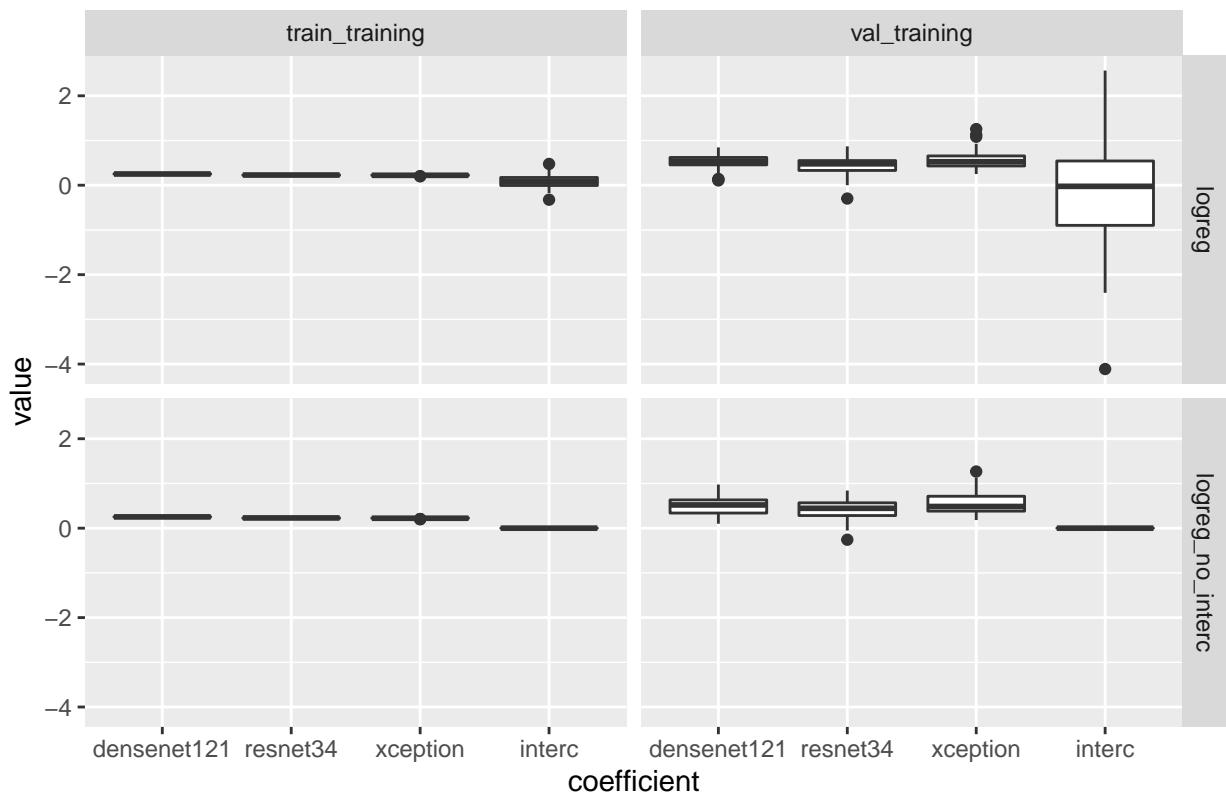
### Coefficients for class 2 vs 4



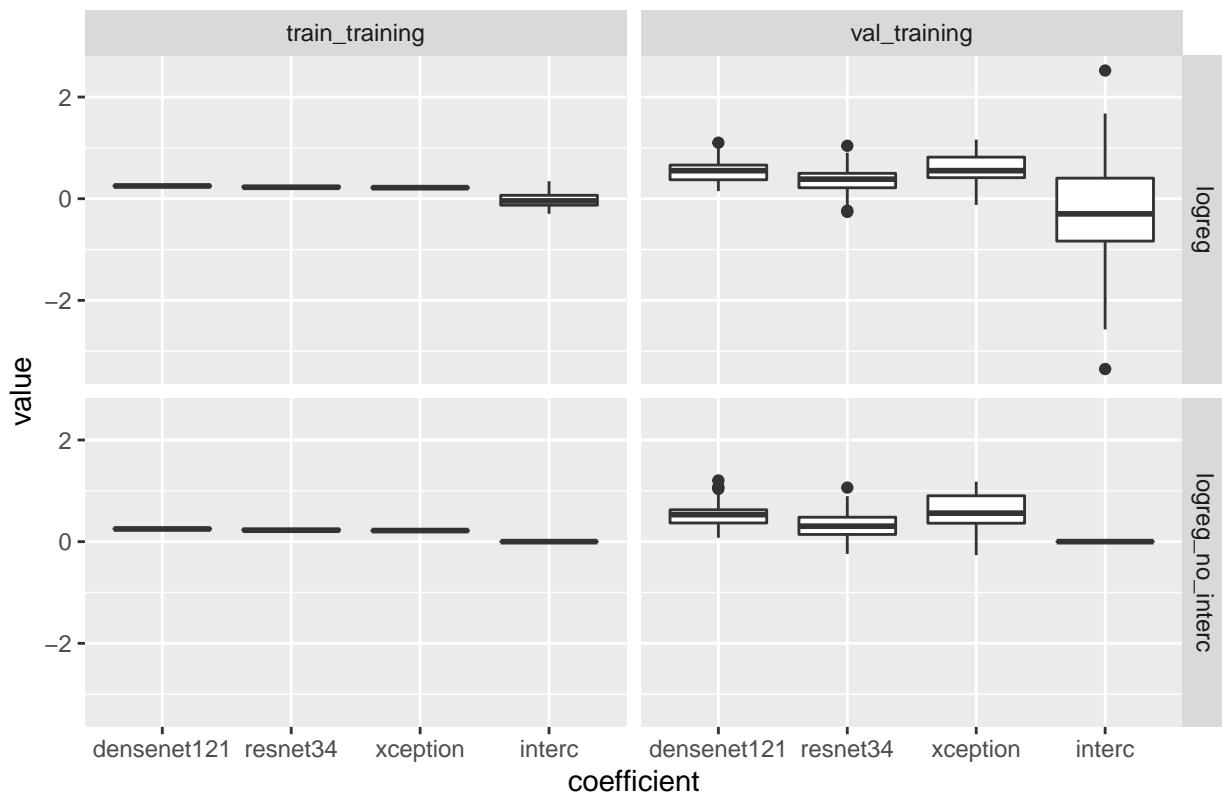
### Coefficients for class 2 vs 5



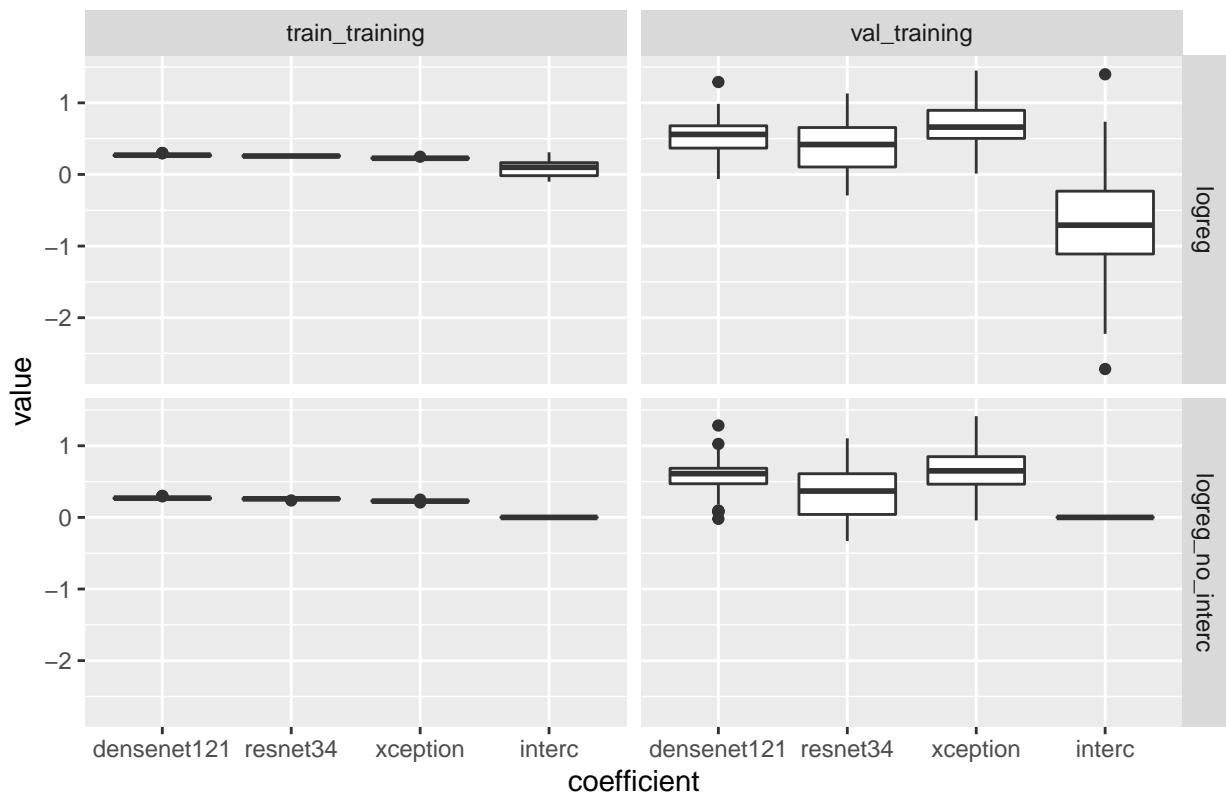
### Coefficients for class 2 vs 6



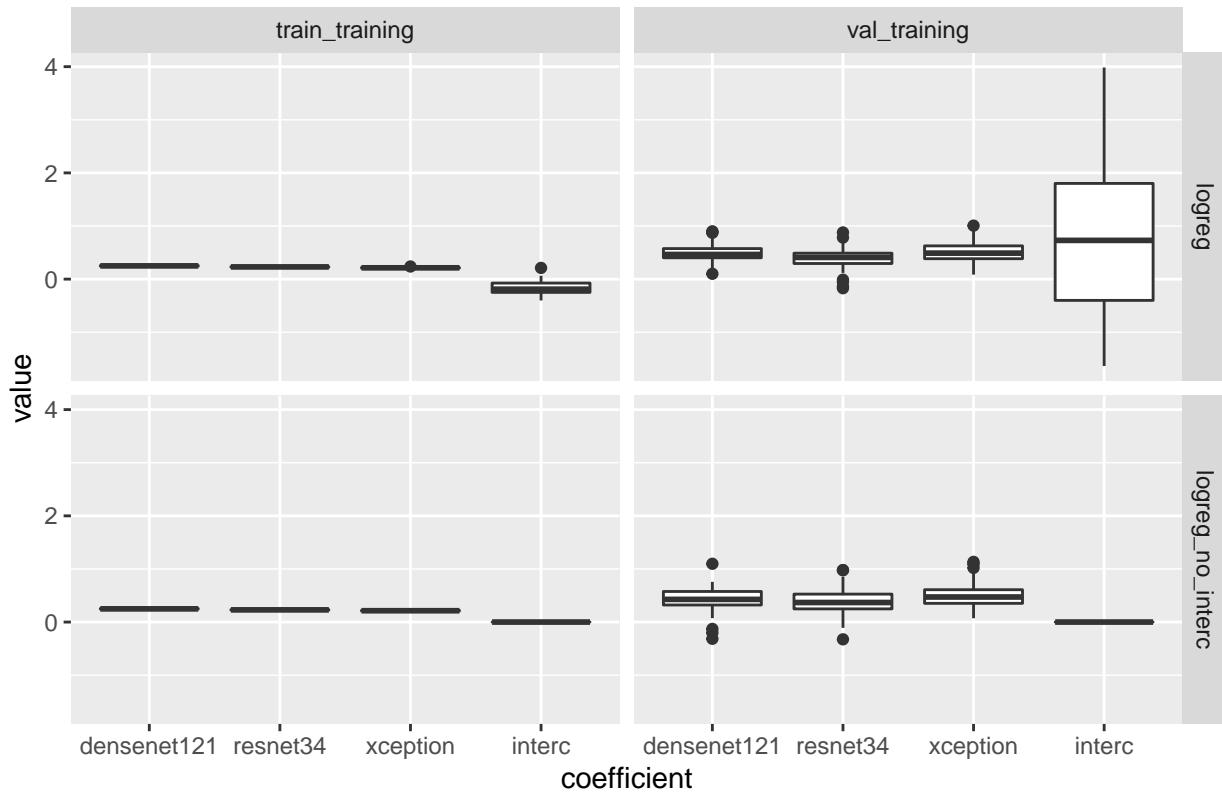
### Coefficients for class 2 vs 7



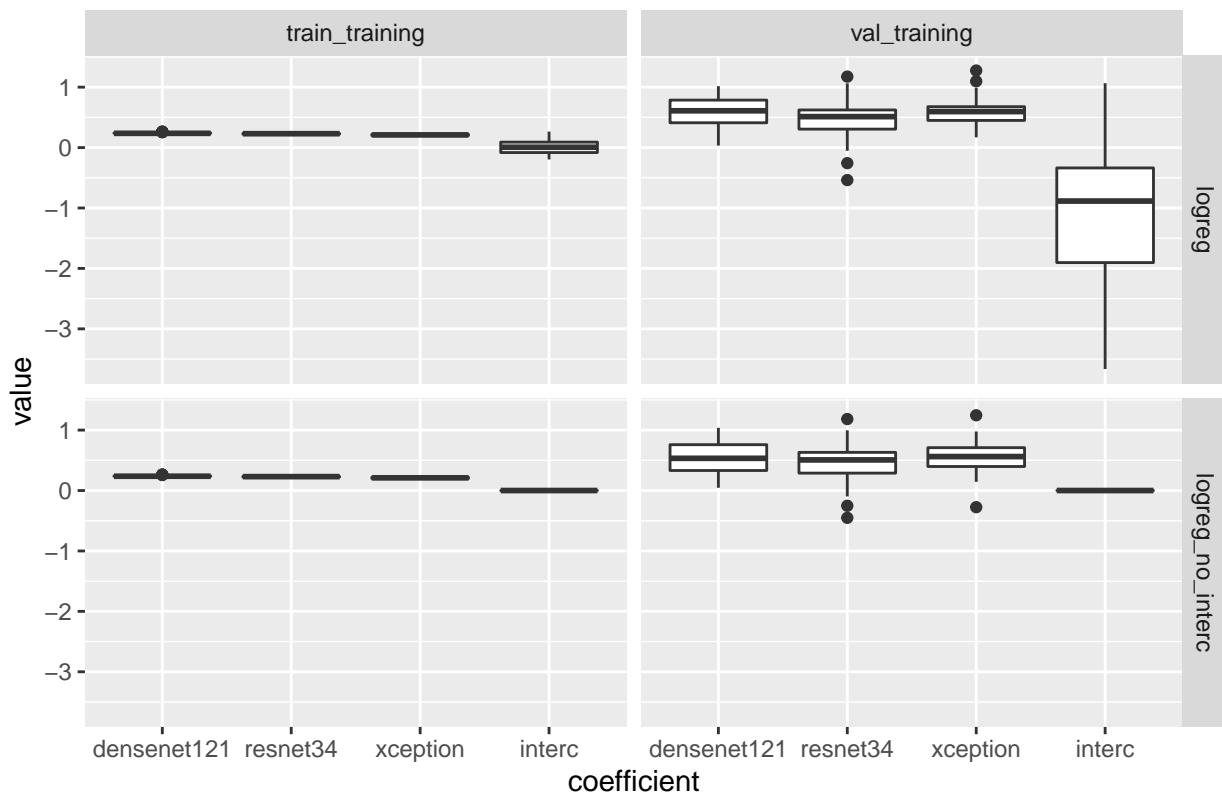
## Coefficients for class 2 vs 8



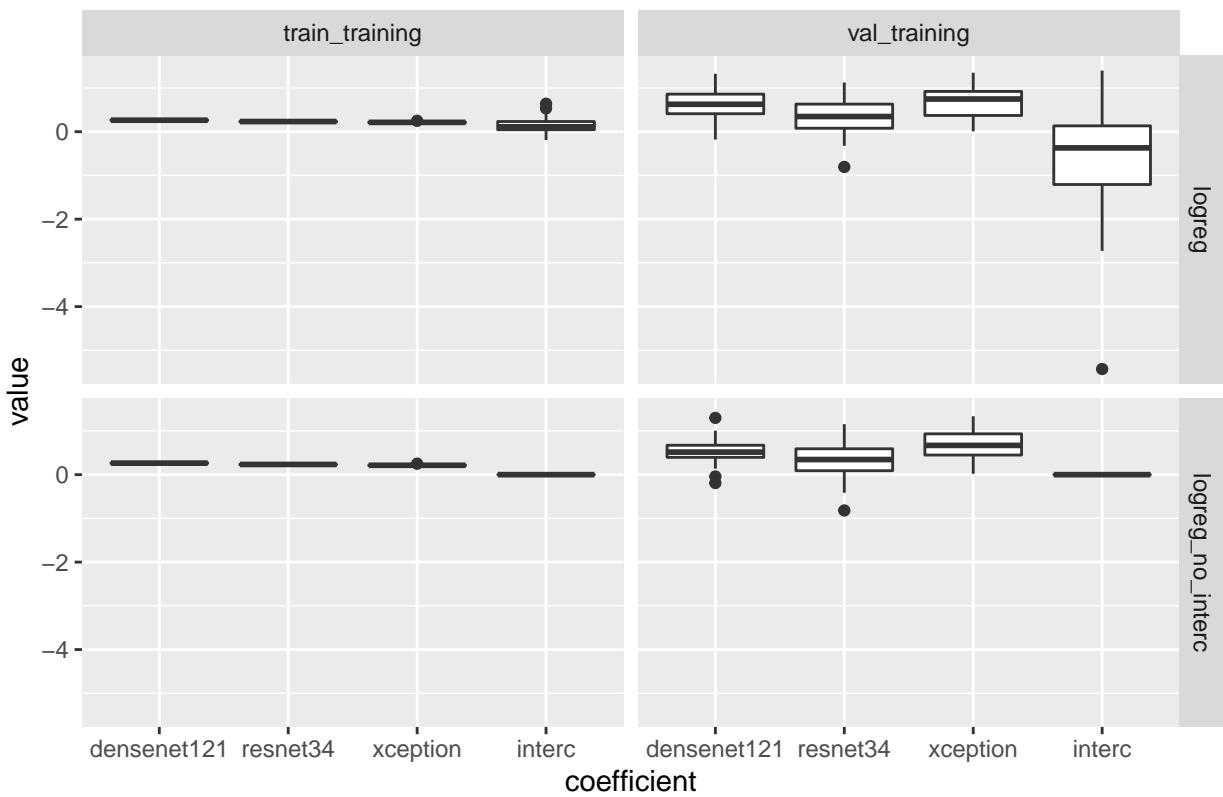
## Coefficients for class 2 vs 9



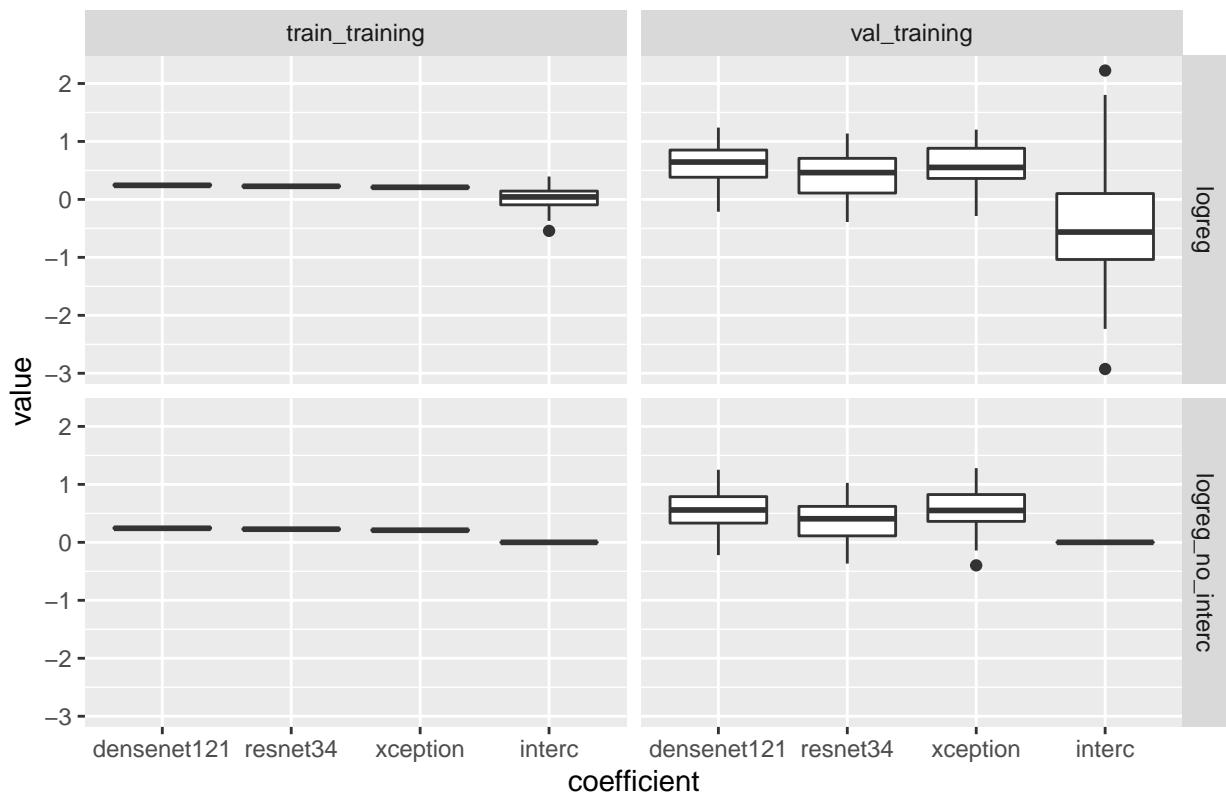
## Coefficients for class 2 vs 10



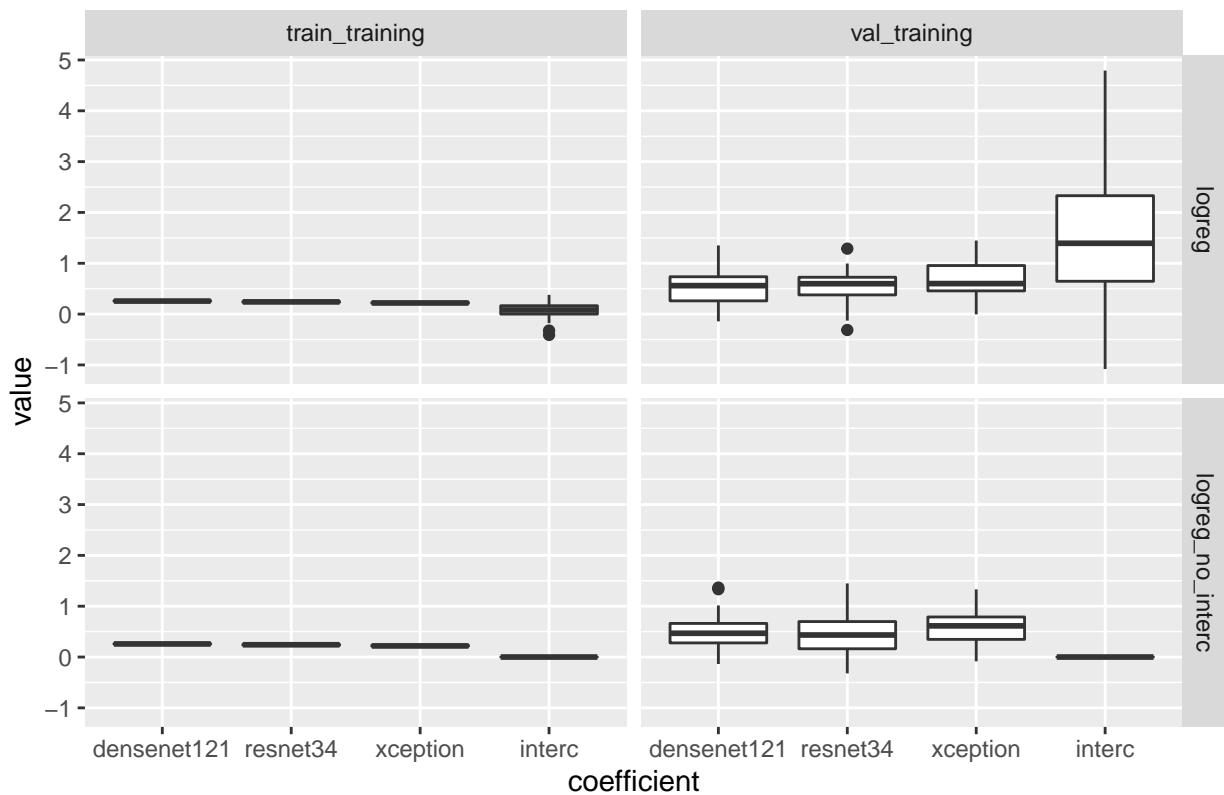
### Coefficients for class 3 vs 4



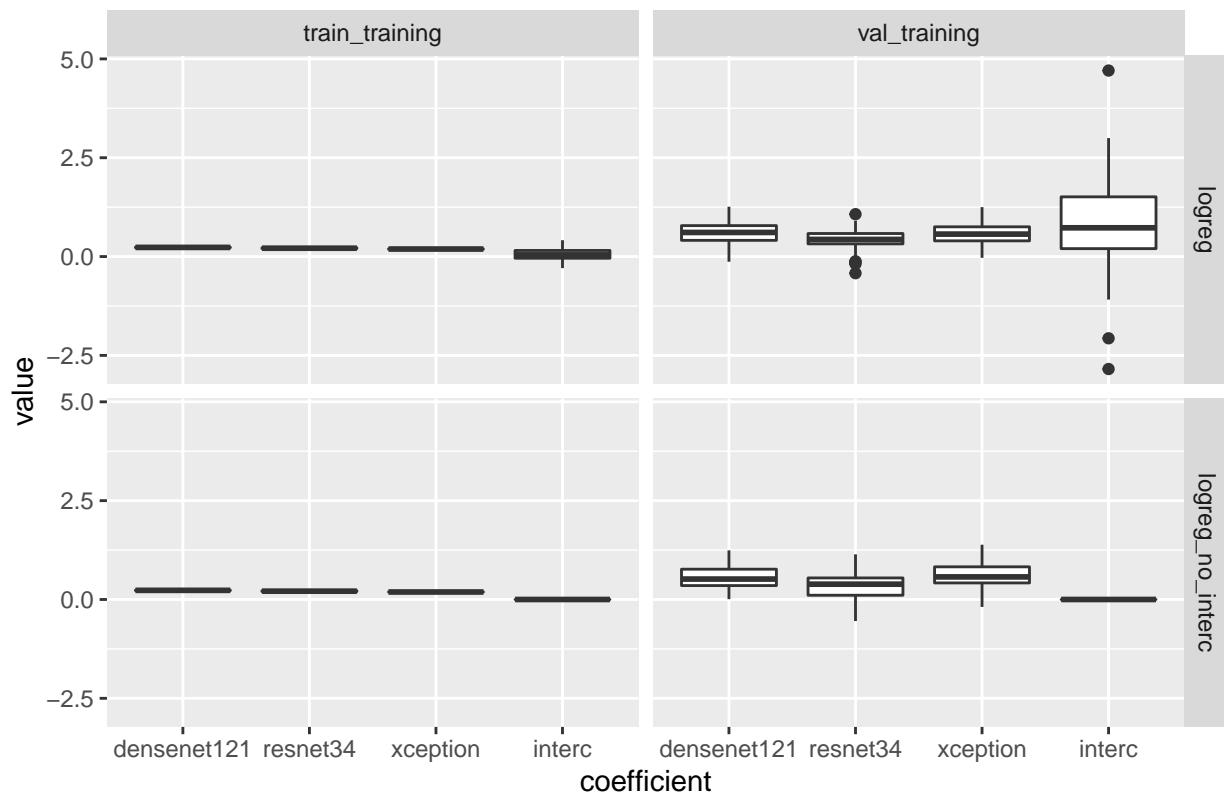
### Coefficients for class 3 vs 5



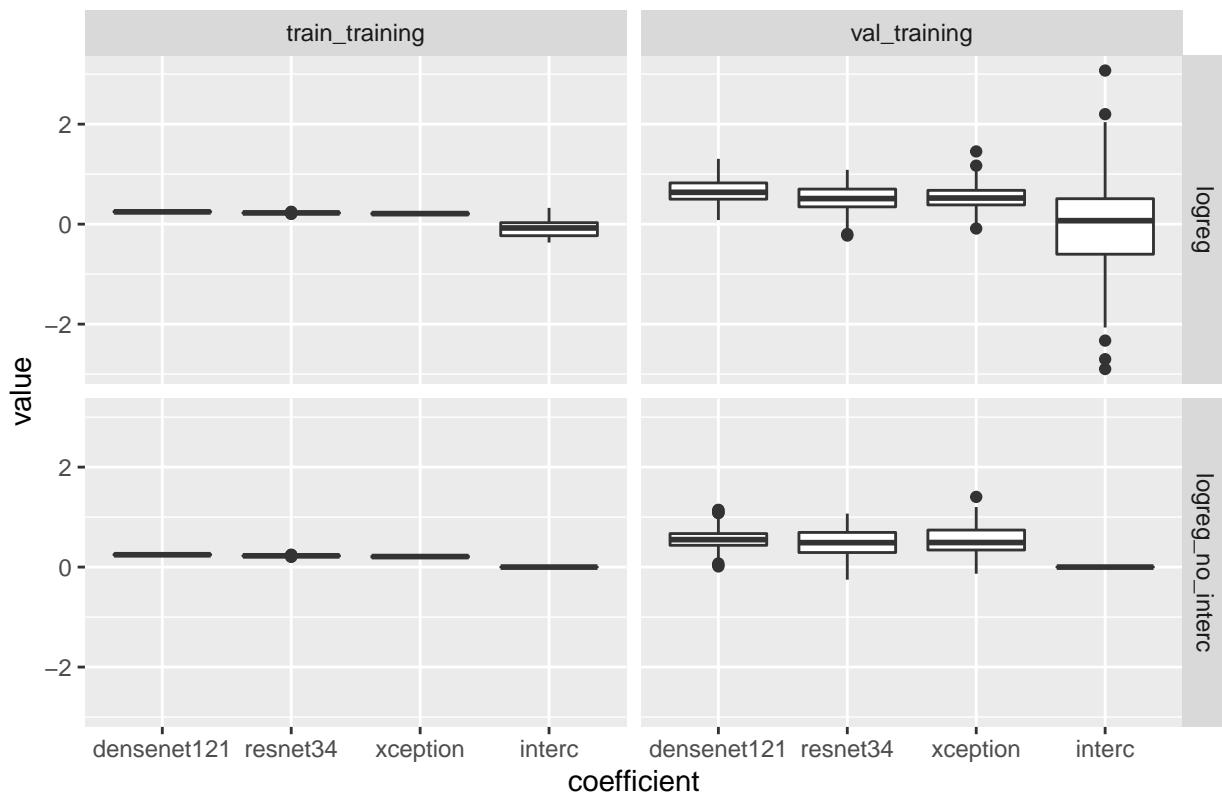
### Coefficients for class 3 vs 6



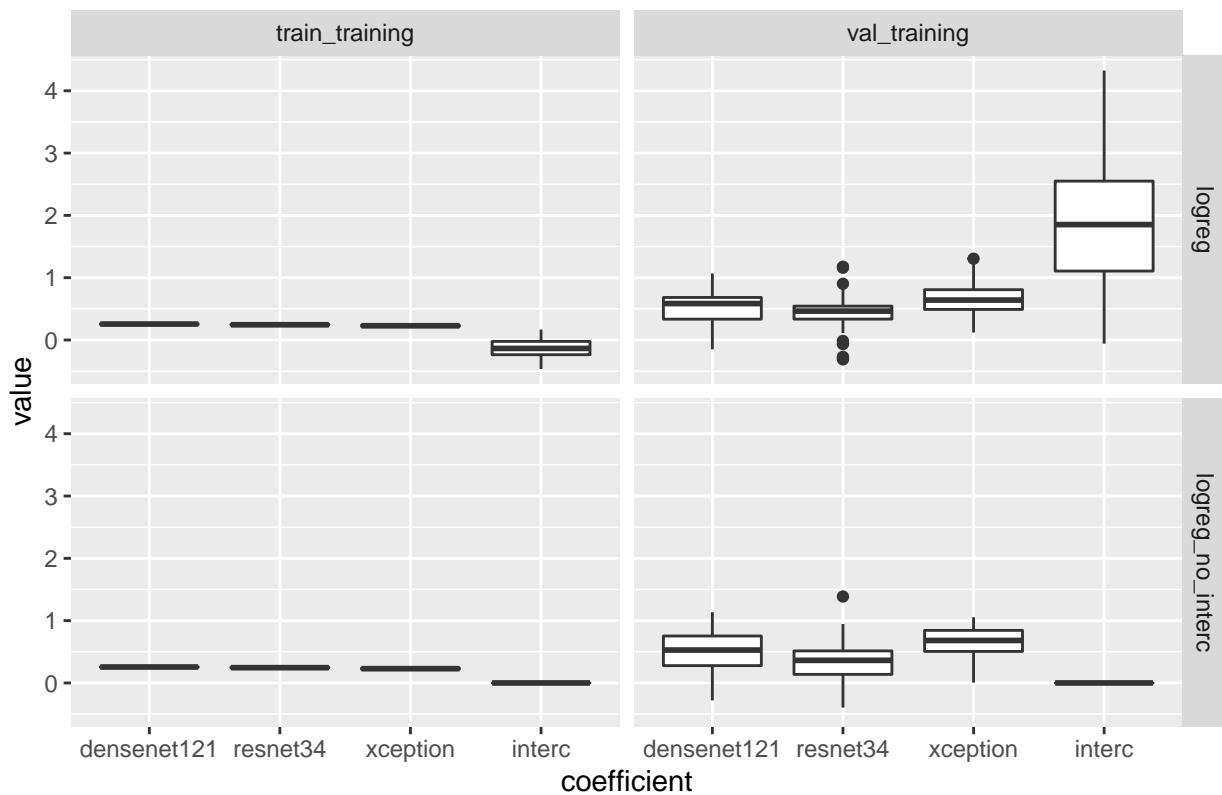
### Coefficients for class 3 vs 7



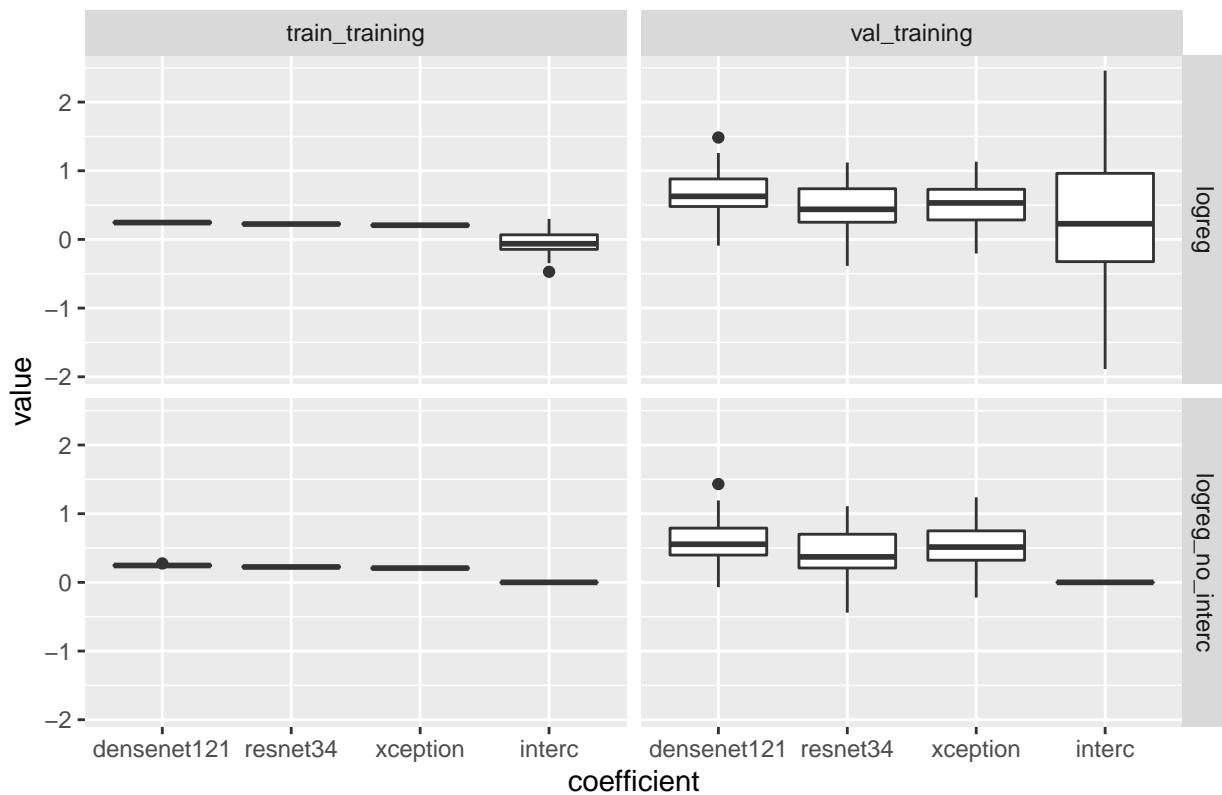
### Coefficients for class 3 vs 8



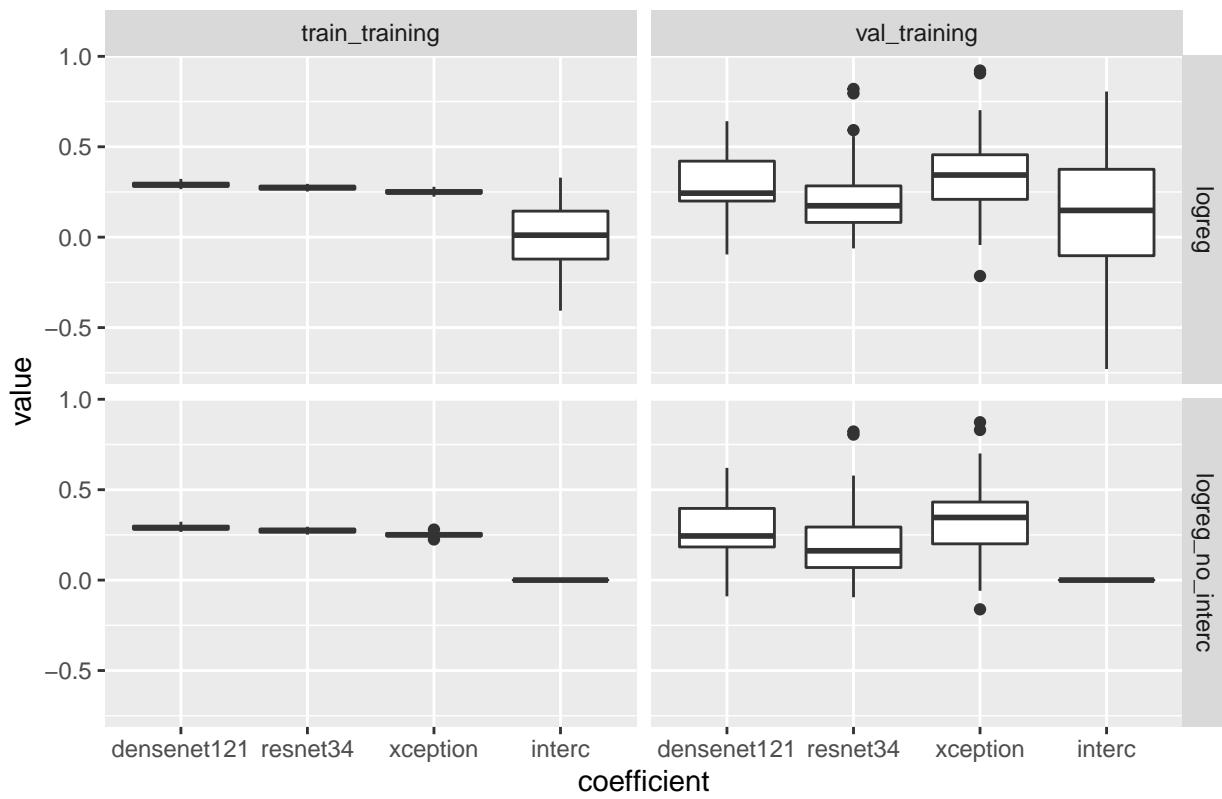
### Coefficients for class 3 vs 9



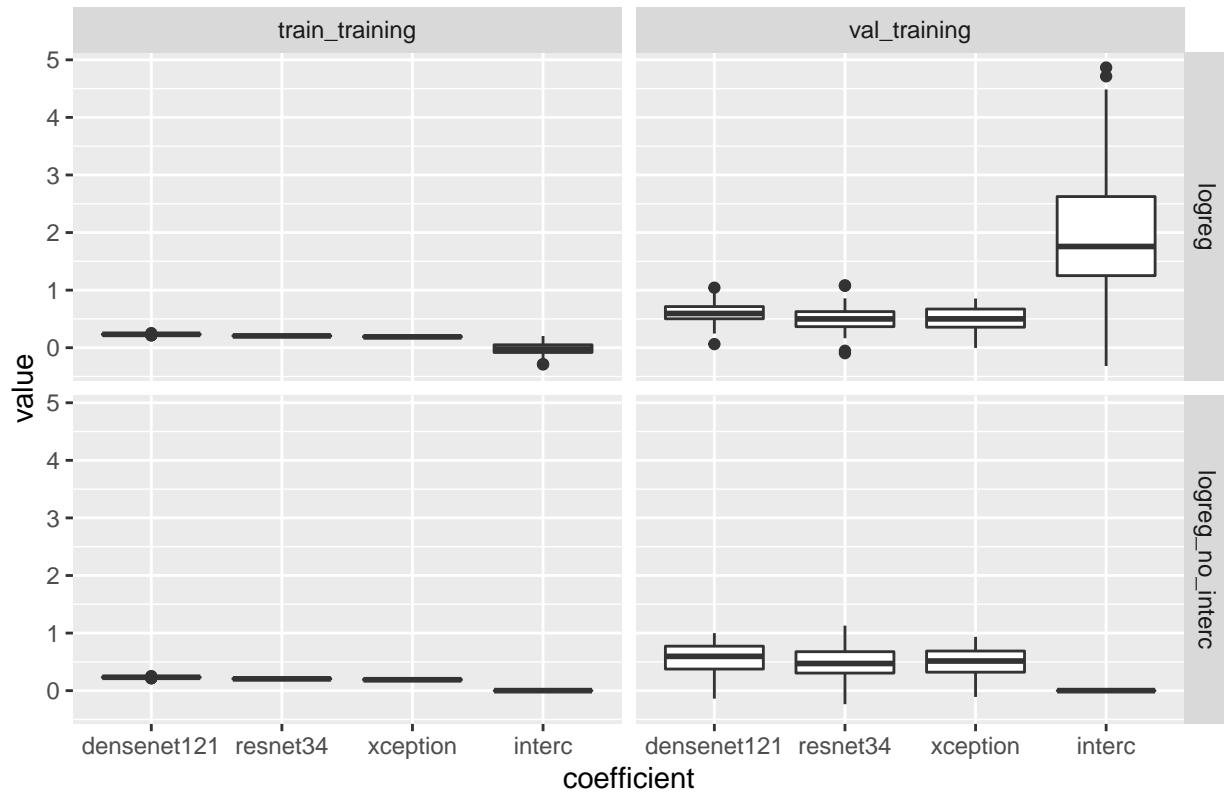
### Coefficients for class 3 vs 10



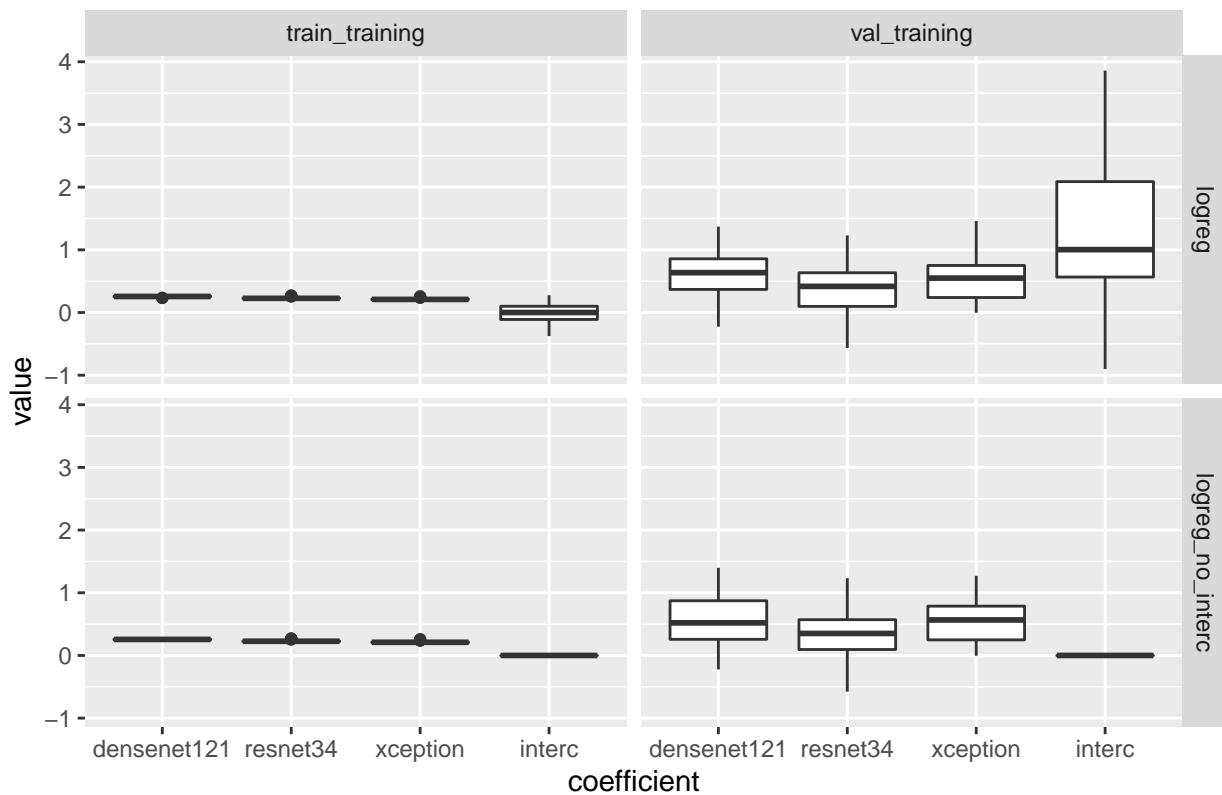
### Coefficients for class 4 vs 5



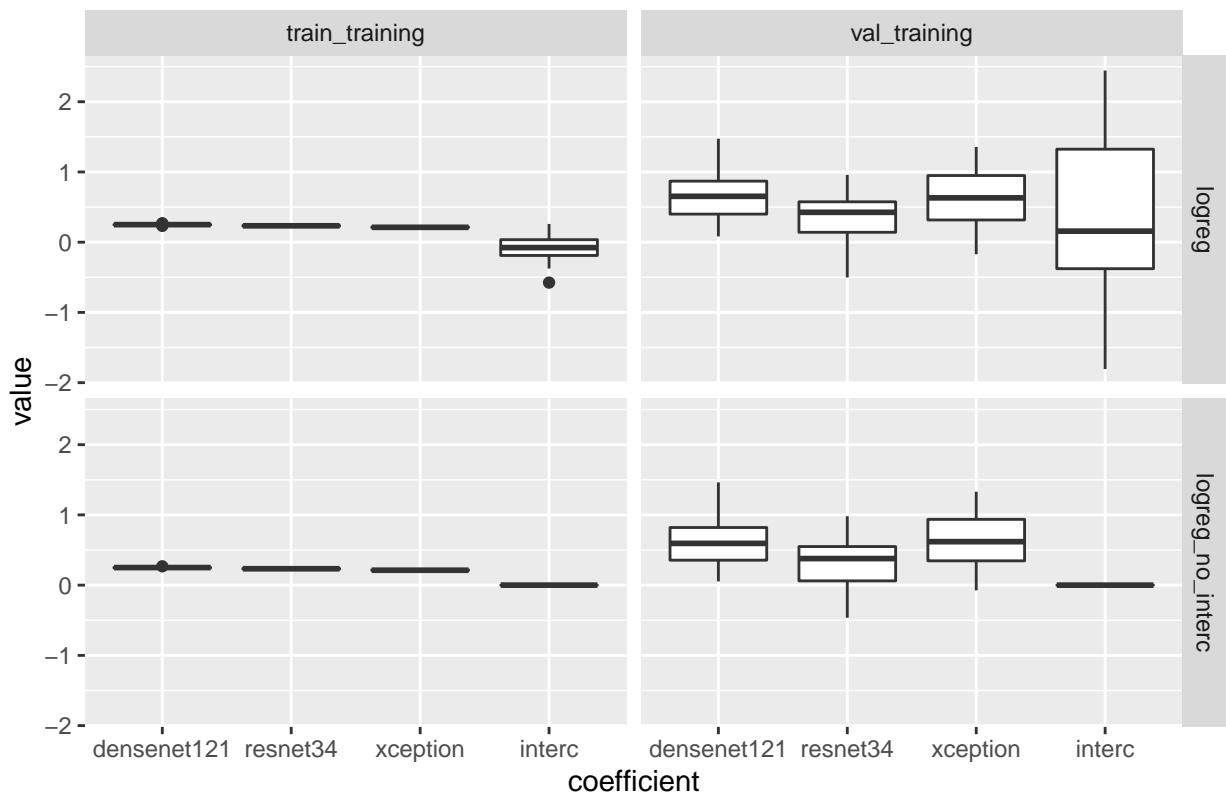
## Coefficients for class 4 vs 6



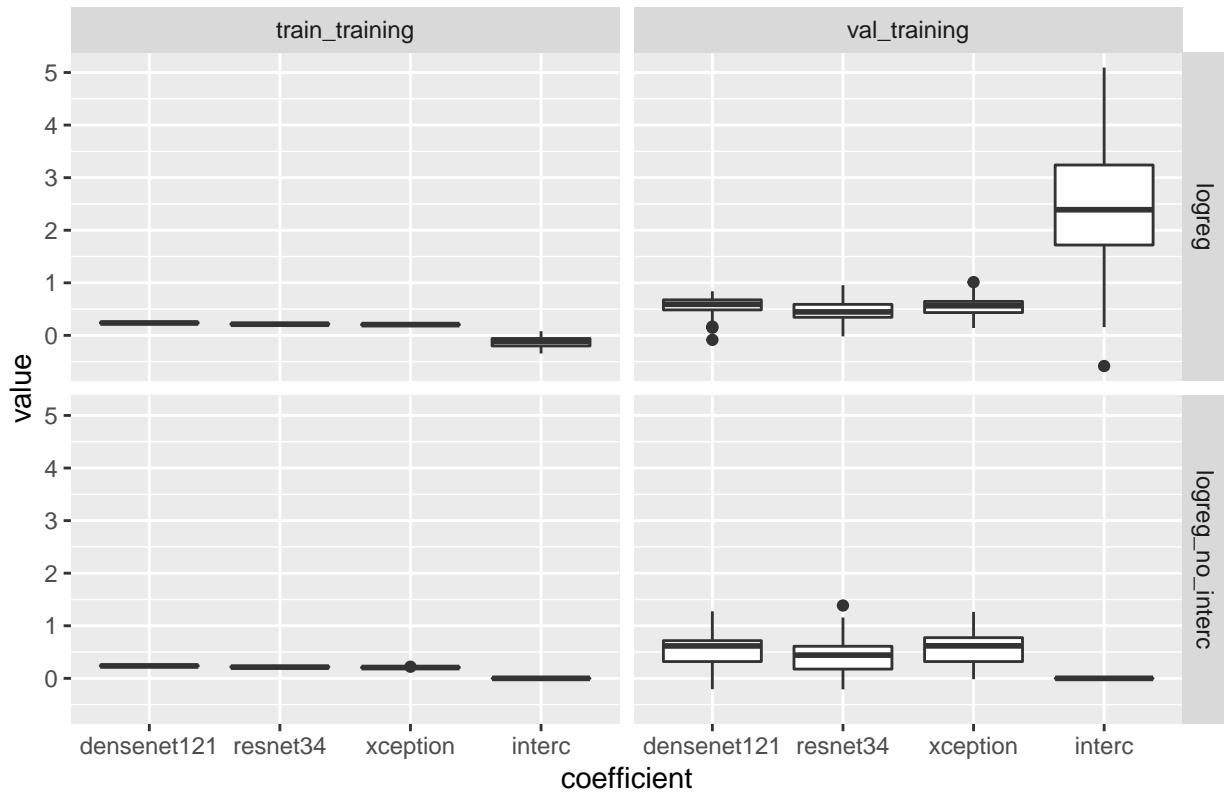
### Coefficients for class 4 vs 7



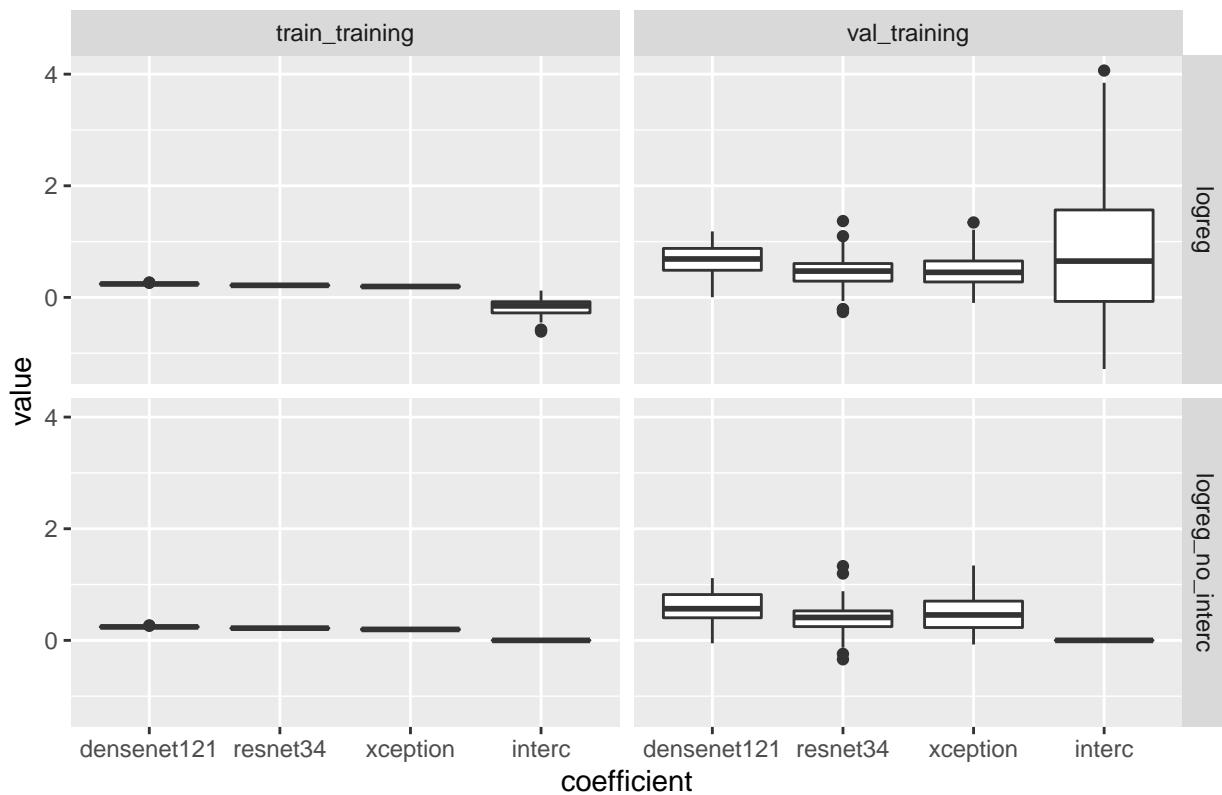
### Coefficients for class 4 vs 8



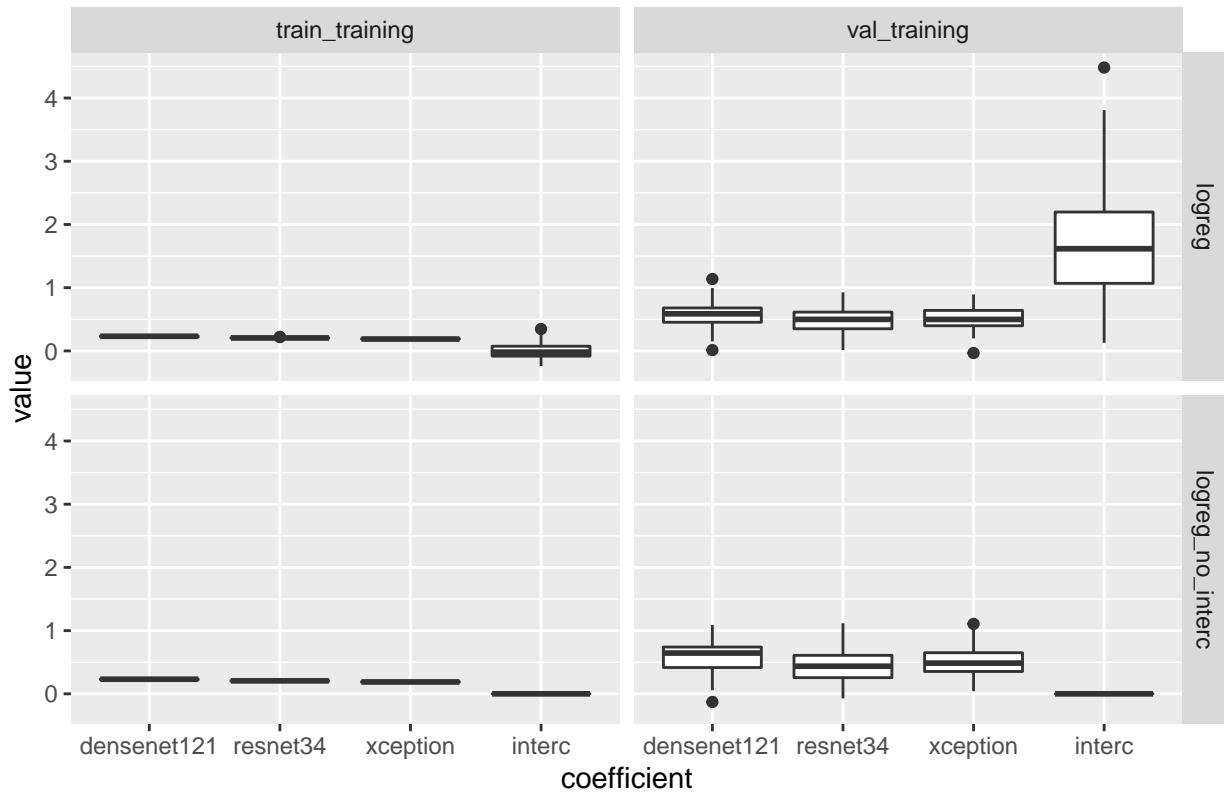
## Coefficients for class 4 vs 9



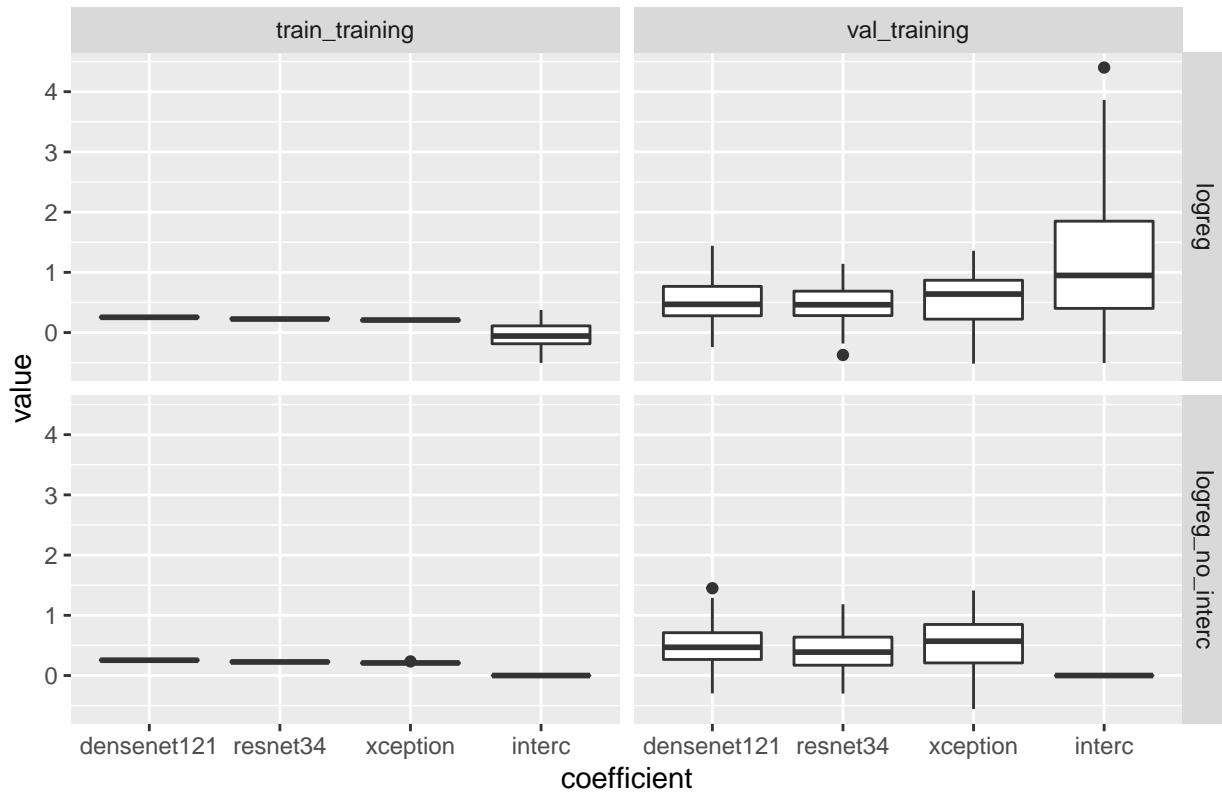
## Coefficients for class 4 vs 10



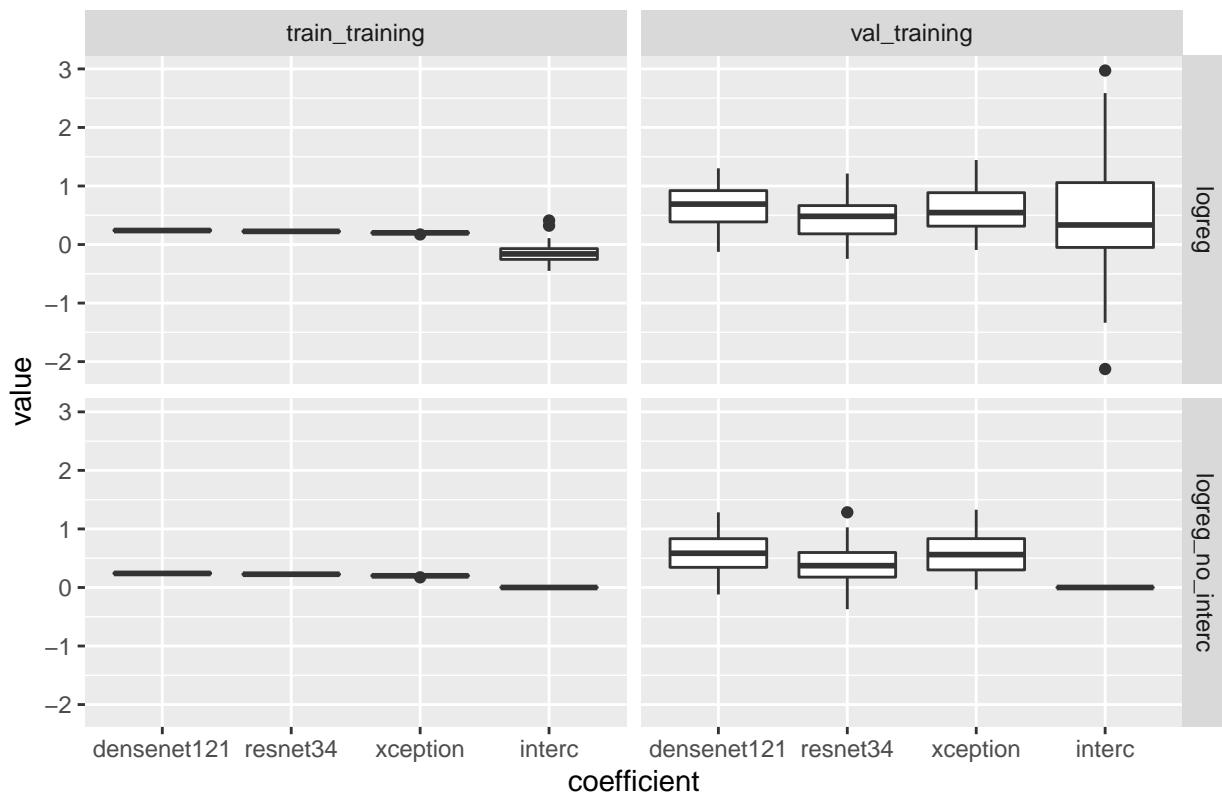
## Coefficients for class 5 vs 6



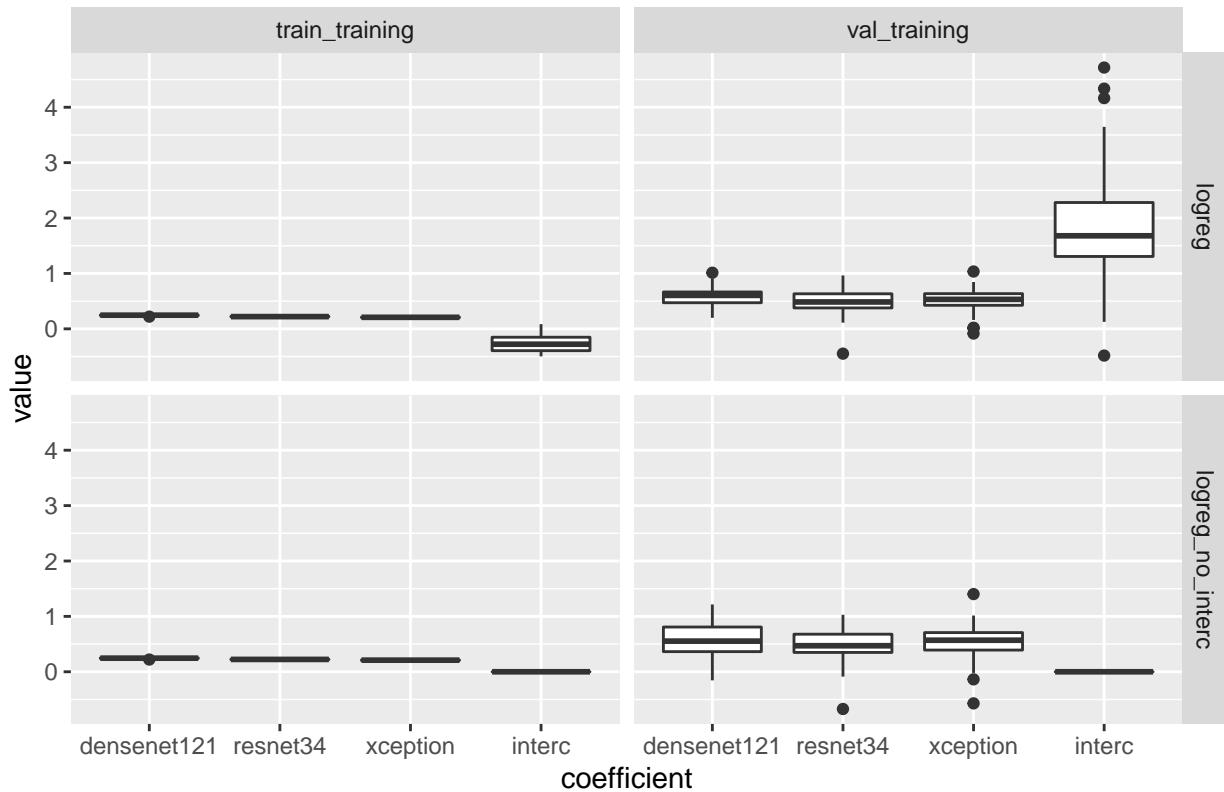
## Coefficients for class 5 vs 7



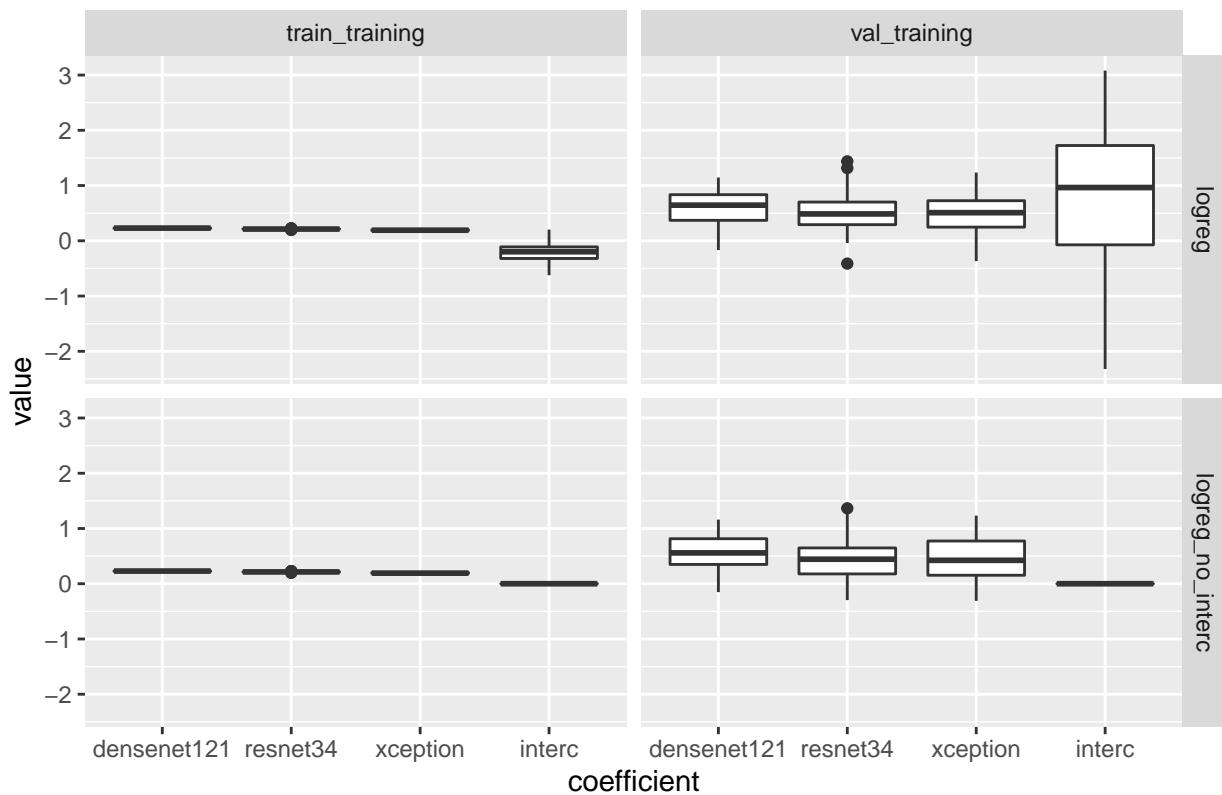
### Coefficients for class 5 vs 8



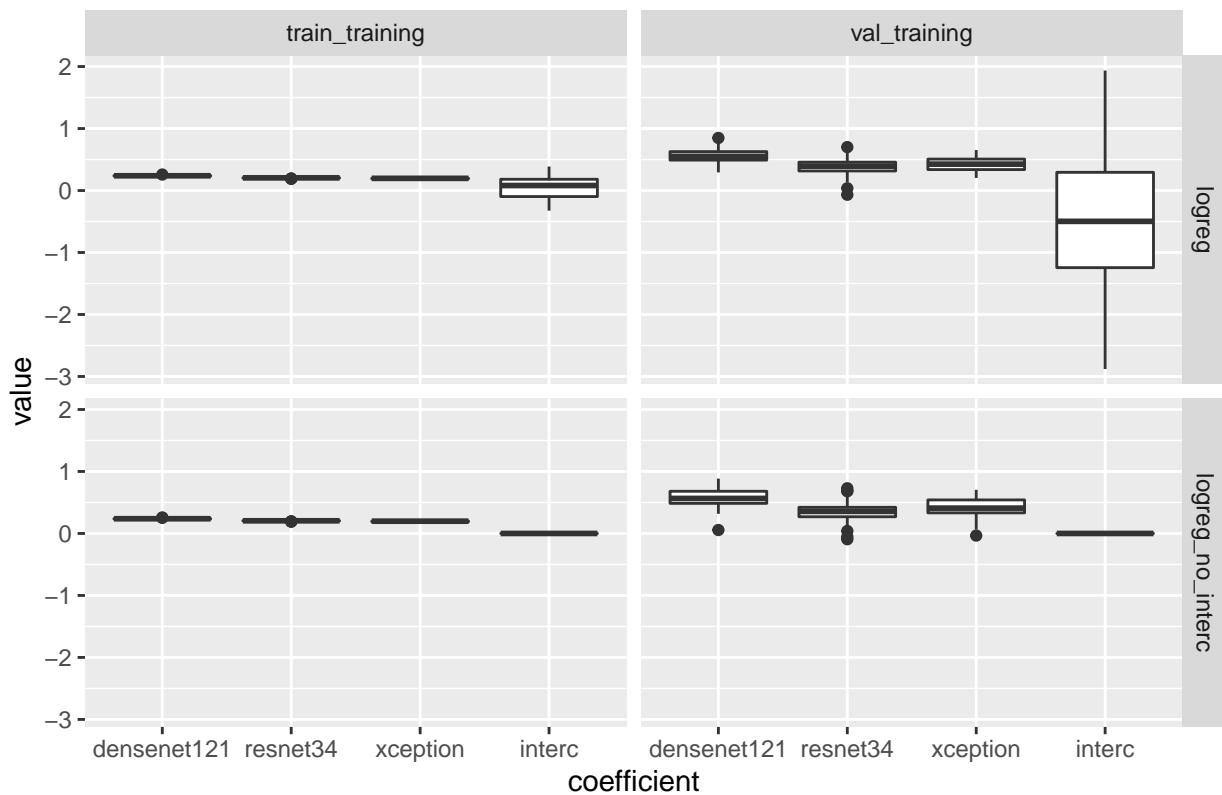
## Coefficients for class 5 vs 9



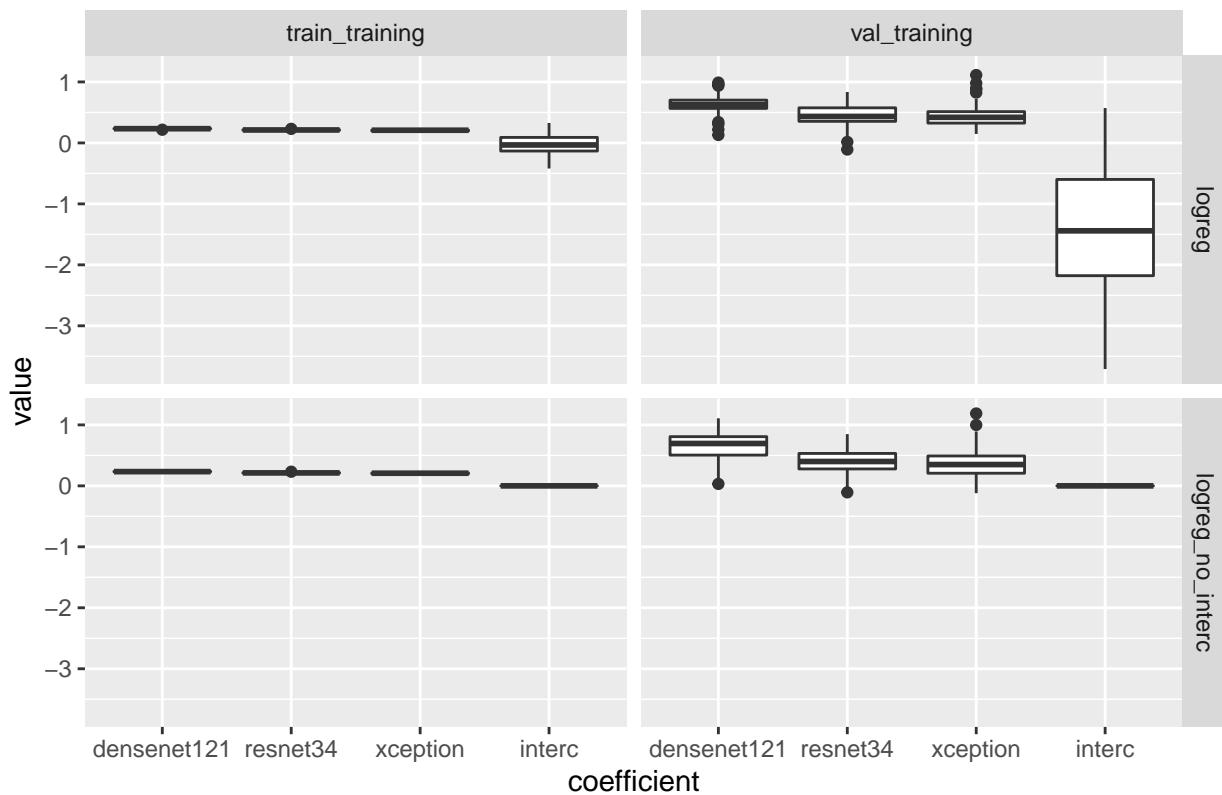
### Coefficients for class 5 vs 10



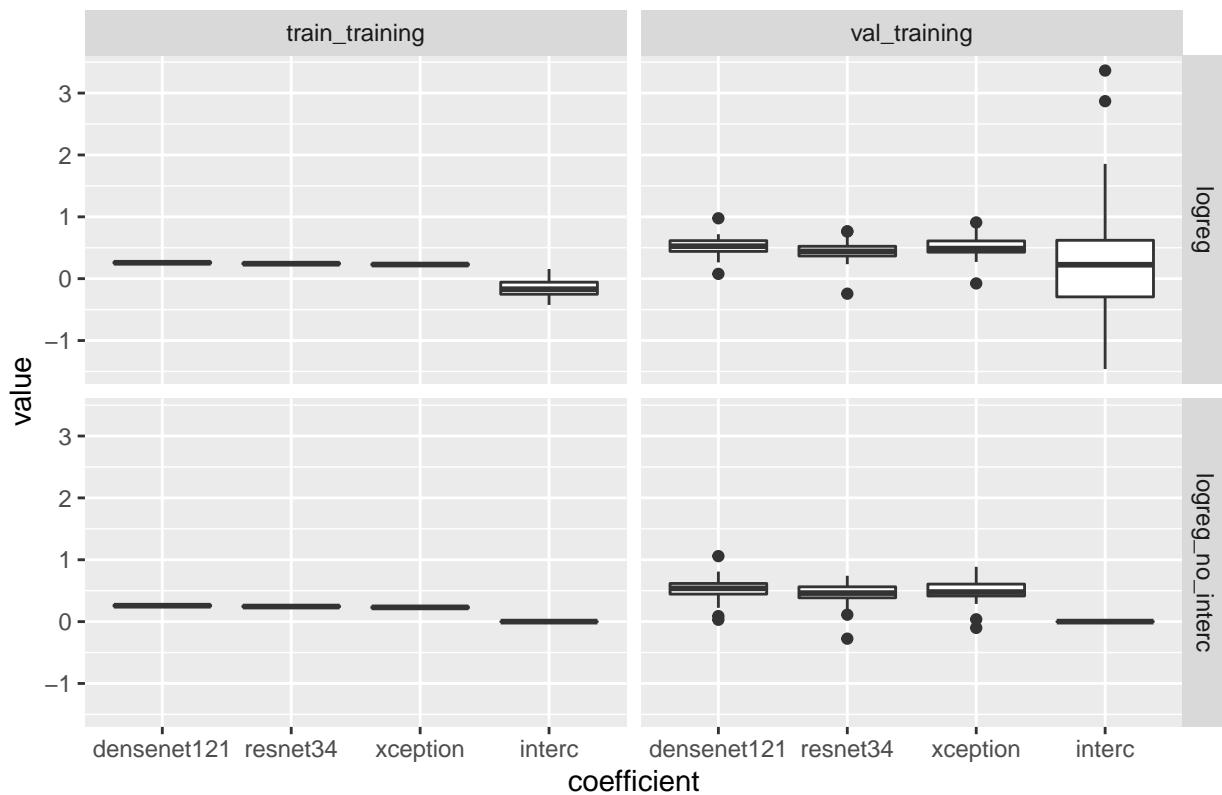
### Coefficients for class 6 vs 7



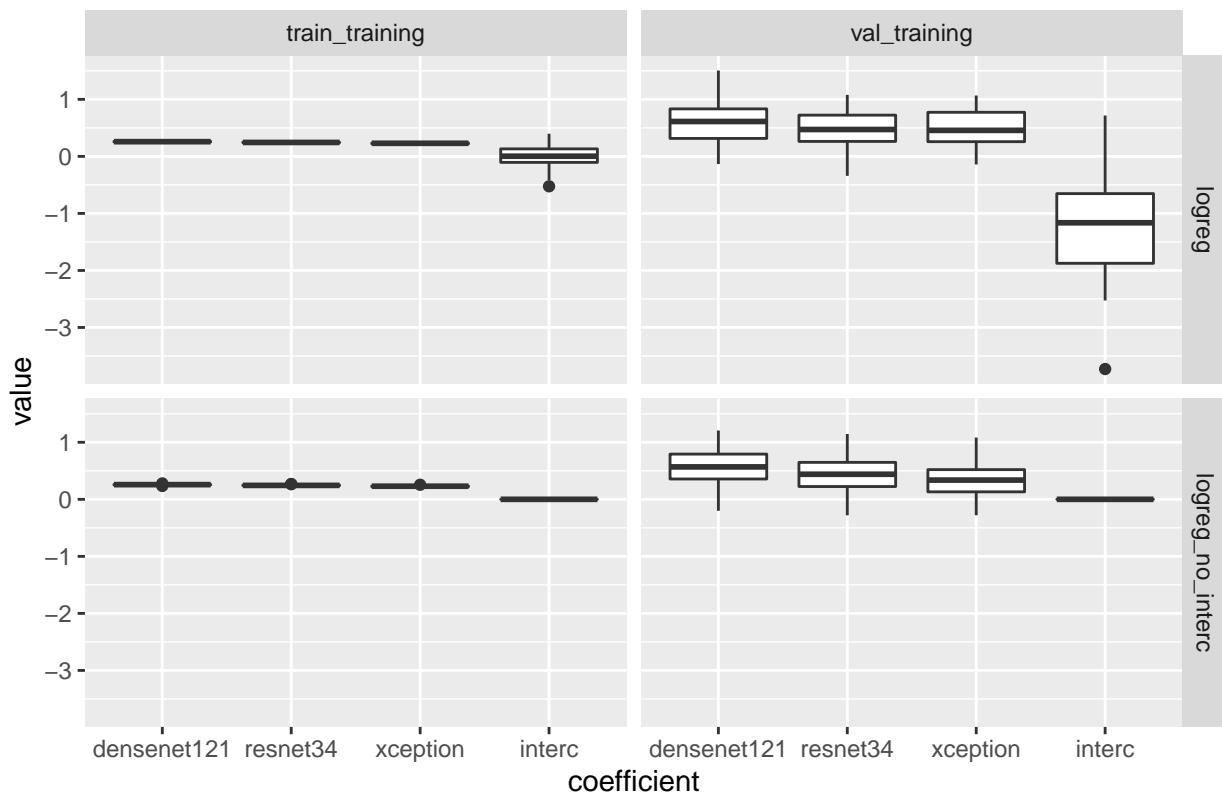
## Coefficients for class 6 vs 8



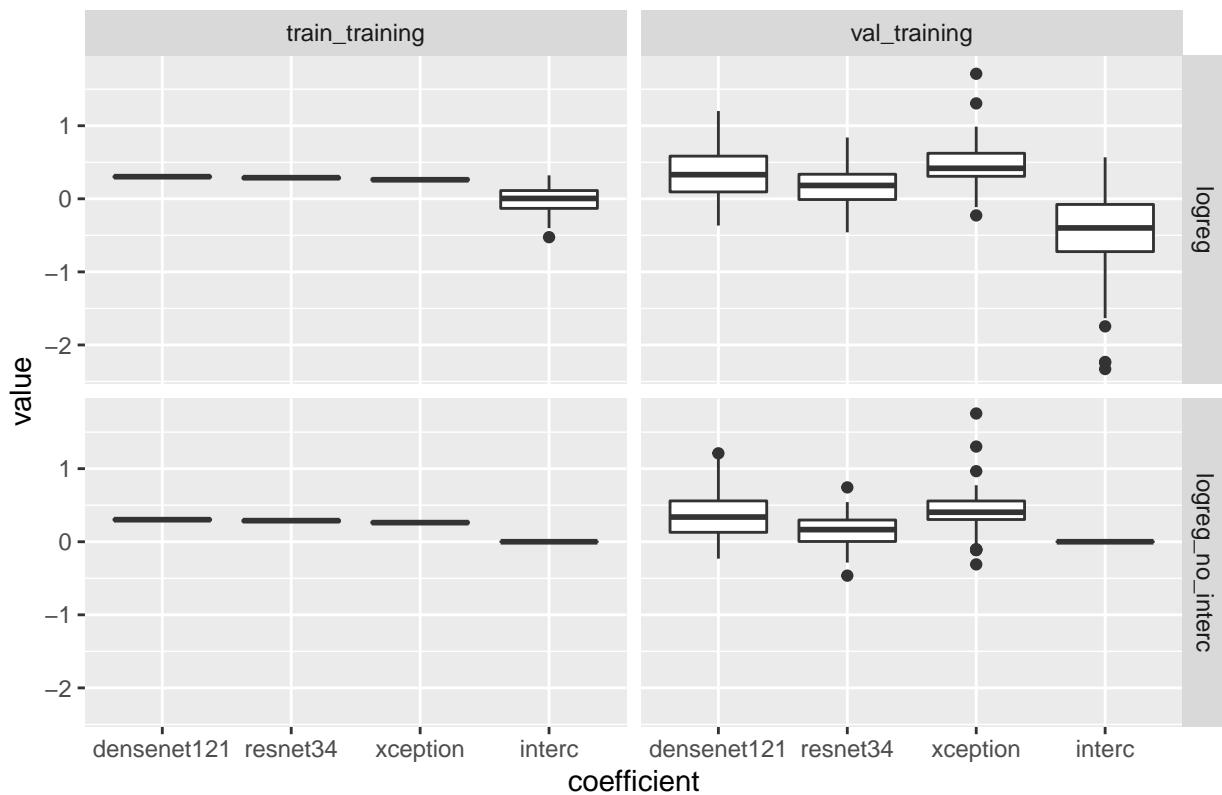
### Coefficients for class 6 vs 9



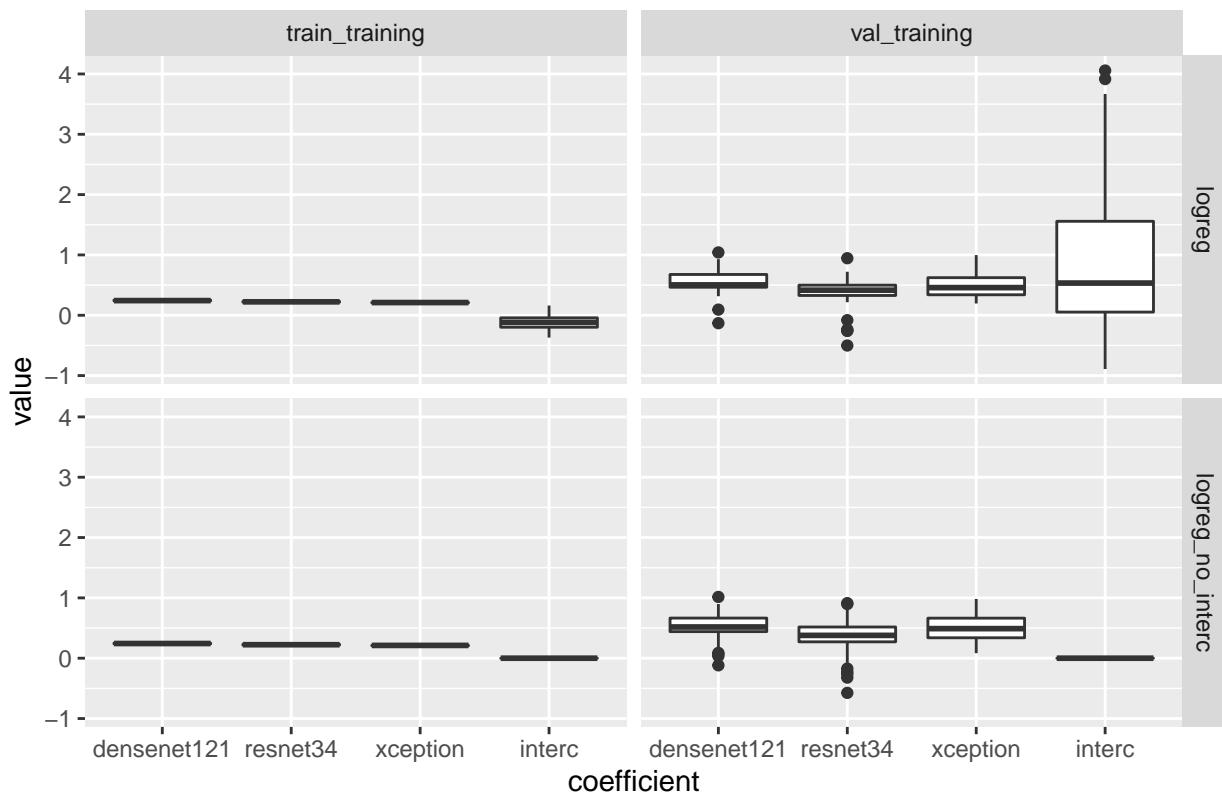
### Coefficients for class 6 vs 10



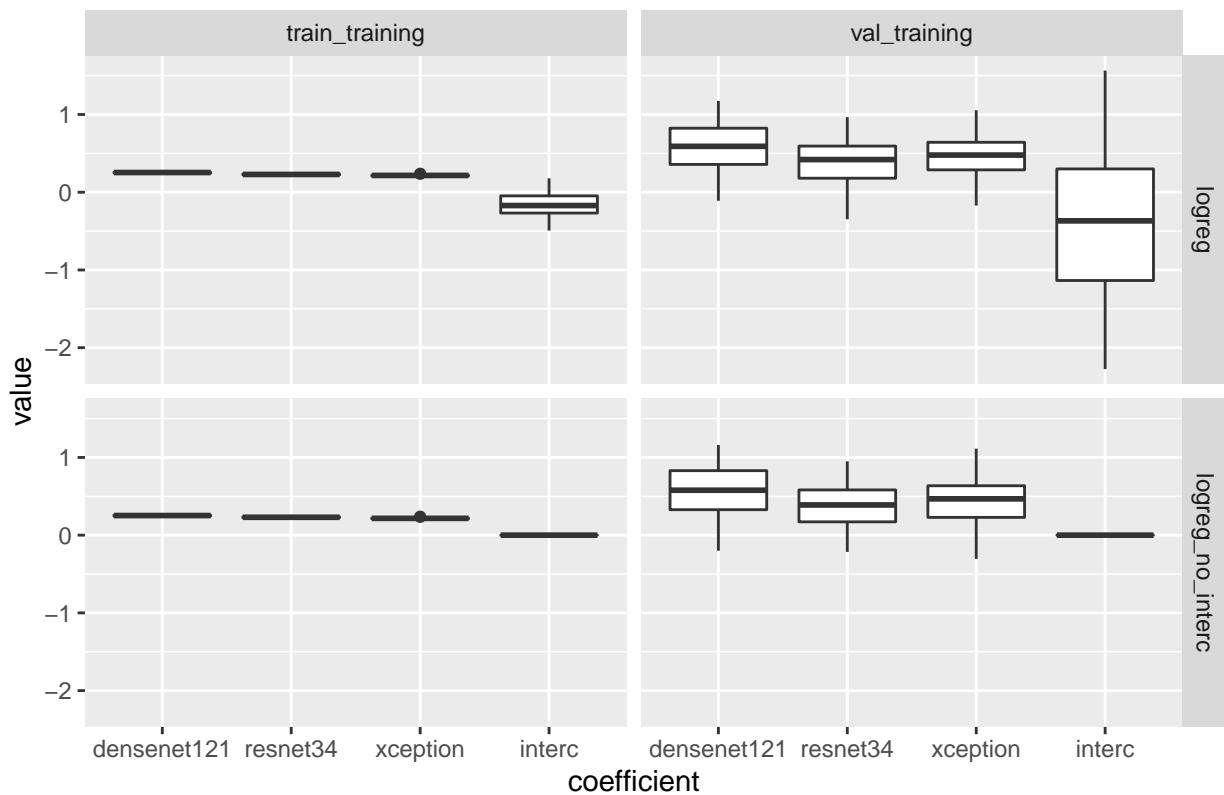
### Coefficients for class 7 vs 8



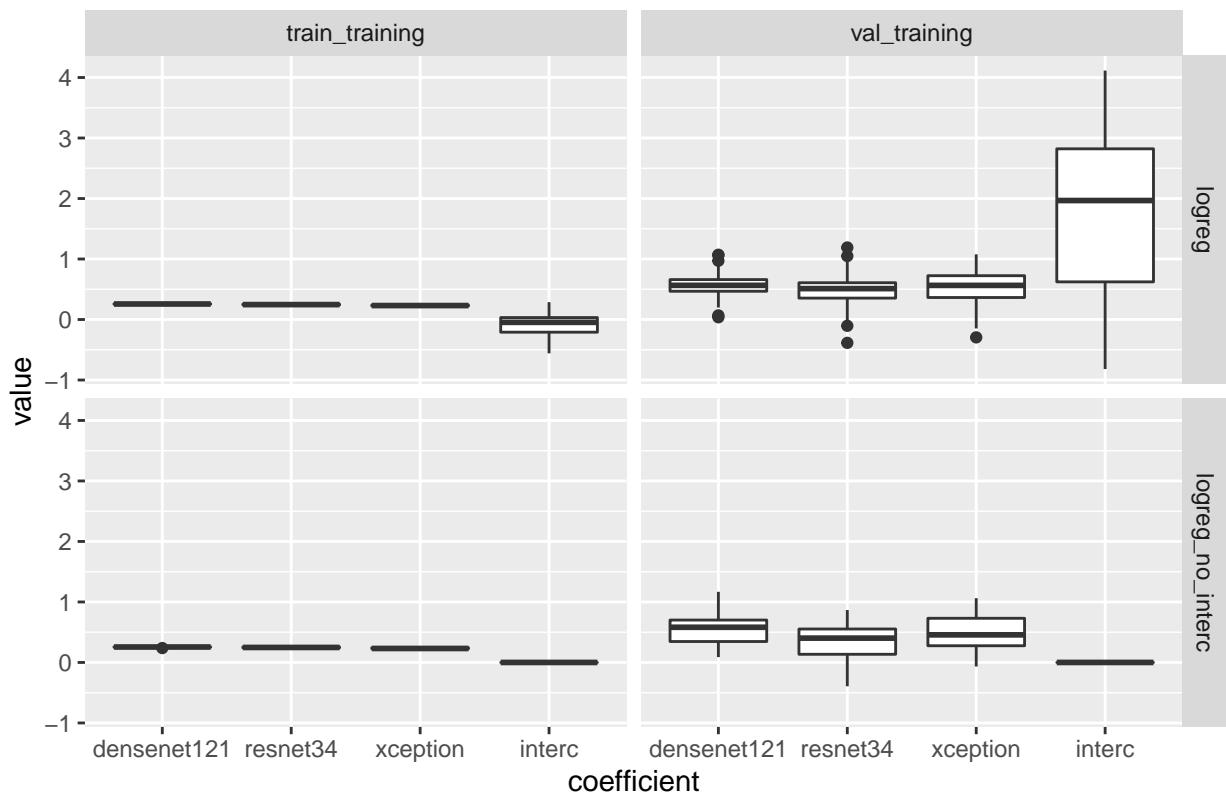
### Coefficients for class 7 vs 9



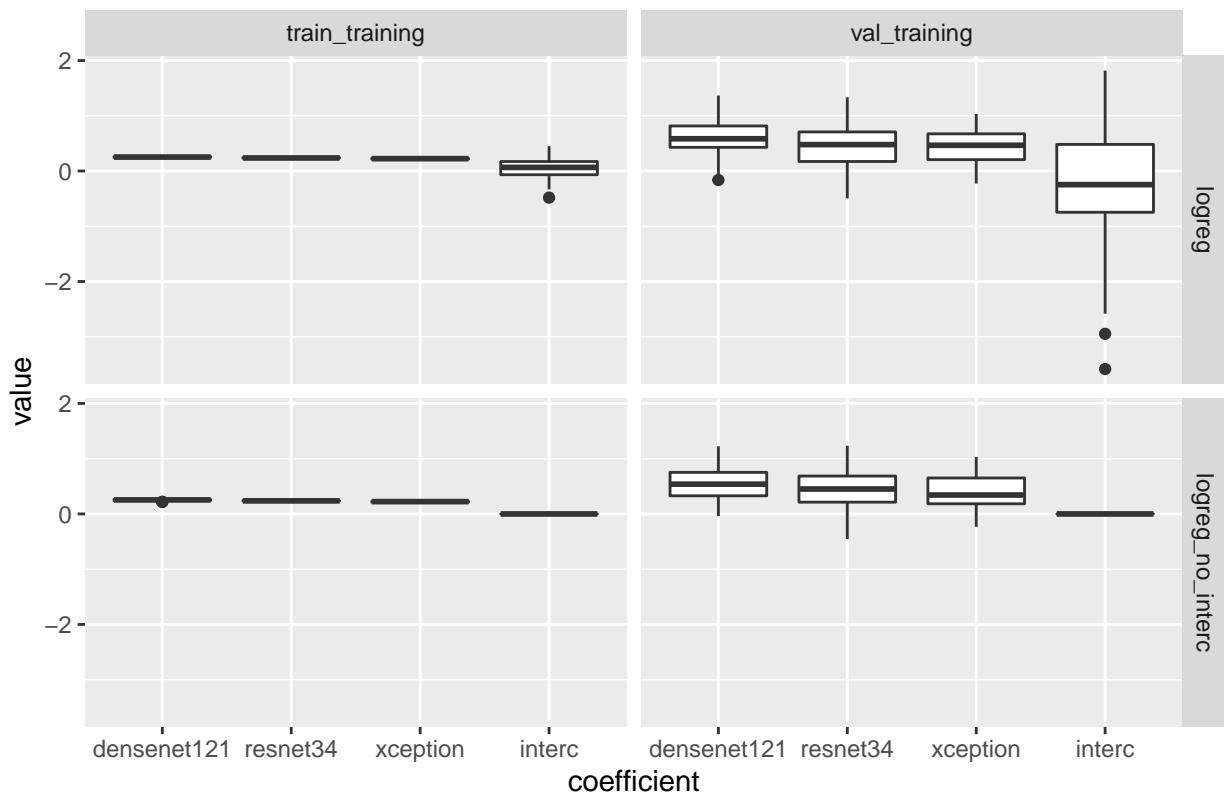
### Coefficients for class 7 vs 10



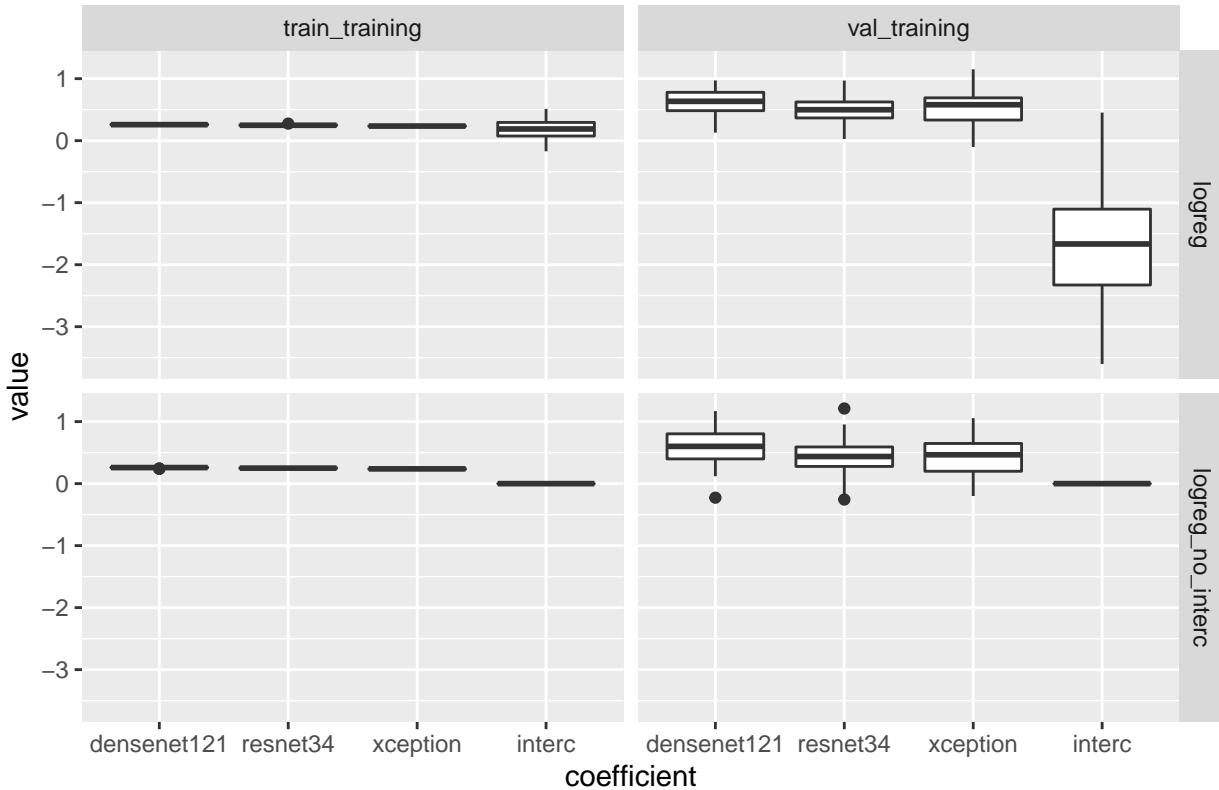
### Coefficients for class 8 vs 9



### Coefficients for class 8 vs 10



## Coefficients for class 9 vs 10



In case of lda train\_training coefficients tend to have higher variance, in case of logreg it seem to be the oposite, val\_training coefficients have higher variance.

```
avg_combiner_coefs <- combiner_coefs %>% filter(coefficient != "interc") %>% group_by(class1, class2, p)

## `summarise()` has grouped output by 'class1', 'class2', 'precision', 'train_type', 'coefficient'. You can use the .groups argument to change this.

avg_combiner_c_w <- pivot_wider(avg_combiner_coefs, names_from = coefficient, values_from = value)
avg_combiner_c_w[, c("class1", "class2")] <- lapply(avg_combiner_c_w[, c("class1", "class2")], as.factor)
avg_combiner_c_w$top_net <- factor(c("densenet121", "resnet34", "xception"))[max.col(as.matrix(avg_combiner_c_w))]

coefs_grid <- ggplot(avg_combiner_c_w, aes(x=class2, y=class1, fill=top_net)) +
  geom_raster() +
  scale_fill_brewer(type="qual") +
  facet_grid(cols=vars(train_type), rows=vars(combining_method)) +
  scale_y_discrete(limits=rev, breaks=seq(0, 100, 10)) +
  scale_x_discrete(breaks=seq(0, 100, 10)) +
  guides(fill=guide_legend(title="Network")) +
  xlab("Class") +
  ylab("Class") +
  ggtitle("Network with highest lda weight for class pairs") +
  theme(plot.title = element_text(hjust = 0.5),
        axis.ticks = element_blank(),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank())
```

coefs\_grid

