

# CO2-NDVI-effect\_PETA\_GAM\_RLM

## Import data

```
library(tidyverse)
library(data.table); #setDTthreads(threads = 0)
library(lubridate);
library(dplyr);
library(RcppArmadillo)
library(sf); library(stars)
library(patchwork); library(colorspace)
library(zyp);
library(mgcv)
set.seed(333)
# IMPORT #####
oz_poly <- sf::read_sf("../data_general/GADM/gadm36_AUS.gpkg",
                       layer="gadm36_AUS_1")
oz_poly <- st_as_sf(oz_poly)
oz_poly <- st_simplify(oz_poly, dTolerance = 0.05)

# vegetation index record
vi <- arrow::read_parquet("../data_general/MCD43/MCD43_AVHRR_NDVI_hybrid_2020-10-12.parquet") %>%
  as.data.table()
vi <- vi %>% lazy_dt() %>%
  mutate(ndvi_hyb_e1 = coalesce(ndvi_mcd_nm_pred, NA_real_),
         ndvi_hyb_e2 = coalesce(ndvi_mcd, NA_real_)) %>%
  mutate(ndvi_hyb = coalesce(ndvi_hyb_e2, ndvi_hyb_e1)) %>%
  mutate(ndvi_hyb = ifelse(between(ndvi_hyb, 0, 1), ndvi_hyb, NA_real_)) %>%
  as.data.table()
norms_vti <- vi[, `:=` (month=month(date))] %>%
  .[,. (ndvi_u = mean(ndvi_hyb, na.rm=TRUE),
        ndvi_sd = sd(ndvi_hyb, na.rm=TRUE)), keyby=.(x,y,month)]
vi <- norms_vti[vi, on=.(x,y,month)] %>%
  .[, `:=` (ndvi_anom = ndvi_hyb - ndvi_u)] %>%
  .[, `:=` (ndvi_anom_sd = ndvi_anom/ndvi_sd)]

lvi <- lazy_dt(vi)

lvi %>% group_by(x,y) %>% summarize(ndvi_u =mean(ndvi_hyb,na.rm=TRUE)) %>% show_query()

## '_DT2'[, .(ndvi_u = mean(ndvi_hyb, na.rm = TRUE)), keyby = .(x,
##           y)]

norms_vti <- vi[, `:=` (month=month(date))] %>%
  .[,. (ndvi_u = mean(ndvi_hyb,na.rm=TRUE),
        ndvi_sd = sd(ndvi_hyb,na.rm=TRUE)), keyby=.(x,y,month)]
```

```

vi <- norms_vi[vi,on=.(x,y,month)] %>%
  .[, `:=` (ndvi_anom = ndvi_hyb - ndvi_u)] %>%
  .[, `:=` (ndvi_anom_sd = ndvi_anom/ndvi_sd)]

vi <- vi %>% select(-season)

dat <- arrow::read_parquet("/home/sami/scratch/ARD_ndvi_aclim_anoms.parquet",
                           col_select = c(
                               "date", "hydro_year", "id",
                               "precip", "precip_12mo", "precip_anom", "precip_anom_12mo", "map",
                               "vpd15", "vpd15_12mo", "vpd15_anom", "vpd15_anom_12mo", "mavpd15",
                               "pet", "pet_anom", "pet_anom_12mo", "pet_12mo", "mapet",
                               "pe", "pe_12mo", "pe_anom_12mo",
                               'vc', 'veg_class',
                               'month',
                               "x", "y", "year")) %>%
  as.data.table()

norms_mape <- dat %>% lazy_dt() %>%
  filter(date>=ymd("1982-01-01") & date<=ymd("2011-12-31")) %>%
  group_by(x,y,hydro_year) %>%
  summarize(ppet_12mo = mean(precip_12mo/pet_12mo,na.rm=TRUE)) %>%
  ungroup() %>%
  group_by(x,y) %>%
  summarize(mappet = mean(ppet_12mo,na.rm=TRUE)) %>%
  ungroup() %>%
  as.data.table()
dat <- merge(dat,norms_mape,by=c("x","y"))

dat <- dat[, `:=` (pe_12mo = precip_12mo/pet_12mo)]
dat <- merge(dat,
             vi,
             by=c("x","y","date","month","year"),
             all=TRUE,allow.cartesian=TRUE)
dat <- dat[order(x,y,date)][, ndvi_3mo := frollmean(ndvi_hyb,n = 3,fill = NA,align='center',na.rm=TRUE)]

rm(vi); gc(full=TRUE)

##           used     (Mb)  gc trigger     (Mb)   max used     (Mb)
## Ncells    2392989   127.8  3921038   209.5  3921038   209.5
## Vcells  1201452011  9166.4 3262010088 24887.2 3250881675 24802.3

# Attach CO2 from Mauna Loa
mlo <- readr::read_table("../data_general/CO2_growth_rate/co2_mm_mlo_20200405.txt",
                        skip = 72, col_names = F) %>%
  set_names(
    c("year","month","ddate","co2_avg","co2_int","co2_trend","ndays"))
) %>%
  mutate(date = ymd(paste(year,month,1))) %>%
  select(date,co2_int,co2_trend) %>%
  as.data.table()
dat <- merge(mlo,dat,by="date")
center_co2 <- mean(dat$co2_int)

```

```

dat <- dat[, `:=` (cco2=co2_int-center_co2)]
gc()

##           used     (Mb) gc trigger     (Mb)   max used     (Mb)
## Ncells    2442619  130.5   3921038   209.5   3921038   209.5
## Vcells 1247957046 9521.2 3262010088 24887.2 3250881675 24802.3

dat <- dat[is.na(vc)==F]
dat <- dat[str_detect(vc, "Forests") |
            str_detect(vc, "Eucalypt") |
            str_detect(vc, "Rainforests")]
dat <- dat[x>= 140] # FILTER TO LON >= 140
dat[, `:=` (epoch = ifelse(date<ymd("2000-12-31"), 'avhrr', 'modis'))]
dat[, `:=` (year=year(date), month=month(date))] %>%
  .[, `:=` (season = case_when(month%in%c(3:5)~'MAM',
                                month%in%c(6:8)~'JJA',
                                month%in%c(9:11)~'SON',
                                month%in%c(12,1,2)~'DJF'))]

# Add seasons
dat <- dat %>% mutate(epoch = as_factor(epoch),
                        season = factor(season, levels=c("SON", "DJF", "MAM", "JJA")))

# Add Koppen climate zones
kop <- arrow::read_parquet("../data_general/Koppen_climate/BOM_Koppen_simplified7.parquet")
kop <- setDT(kop)
kop <- kop[,.(x,y,zone)]
dat <- merge(dat, kop, by=c("x", "y"), all = T)

# Load LAI climatology 2002-2019 (before BS fires)
lai <- stars::read_stars("../data_general/Oz_misc_data/MCD15A3H_meanAnnualLAI_20020801_20190801.tif") %>%
  set_names(c("lai")) %>%
  as.data.table()
*****
```

## Only pixel locations with more than 6 observations

In practice this is robust to the number of observations (6 - 18 as a minimum nob)

```

coords_keep <- dat %>%
  lazy_dt() %>%
  group_by(x,y,hydro_year) %>%
  summarize(nobs = sum(is.na(ndvi_hyb)==F)) %>%
  ungroup() %>%
  as.data.table()
coords_keep <- coords_keep %>%
  filter(nobs >= 6) %>%
  group_by(x,y) %>%
  summarize(nobs_annual = n()) %>%
  ungroup()
```

## Summarize NDVI and climate data to annual

```
dat_annual <- dat[ndvi_anom_sd >= -3.5 & ndvi_anom_sd <= 3.5] %>%
  .[date >= ymd("1981-11-01") & date <= ymd("2019-08-30")] %>%
  # . [,. (val = mean(ndvi_3mo, na.rm=TRUE)), by=.(x,y,season,hydro_year)] %>%
  .[, `:=` (epoch=ifelse(hydro_year < 2001, 0, 1))] %>%
  .[is.na(ndvi_hyb)==F] %>%
  .[is.na(pe_anom_12mo)==F] %>%
  merge(., coords_keep, by=c('x', 'y'))
dat_annual <- dat_annual[, `:=` (hydro_year_c = hydro_year-1982,
                                frac_p_anom = precip_anom_12mo/map,
                                frac_ppet_anom = pe_anom_12mo/mappet,
                                frac_pet_anom = pet_anom_12mo/mappet,
                                frac_vpd_anom = vpd15_anom_12mo/mavpd15)]
dat_annual <- dat_annual %>%
  select(x,y,date,ndvi_hyb,co2_trend,
         hydro_year,hydro_year_c,
         frac_p_anom,frac_ppet_anom,frac_pet_anom,frac_vpd_anom,
         map,mappet,mapet,mavpd15,
         epoch,nobs_annual)
dat_annual <- dat_annual[nobs_annual >= 10]
dat_annual <- dat_annual[,. (ndvi_hyb = mean(ndvi_hyb,na.rm=TRUE),
                            co2 = mean(co2_trend,na.rm=TRUE),
                            frac_p_anom = mean(frac_p_anom,na.rm=TRUE),
                            frac_pet_anom = mean(frac_pet_anom,na.rm=TRUE),
                            frac_ppet_anom = mean(frac_ppet_anom,na.rm=TRUE),
                            frac_vpd_anom = mean(frac_vpd_anom,na.rm=TRUE),
                            epoch = mean(epoch, na.rm=TRUE),
                            map = mean(map,na.rm=TRUE),
                            mapet = mean(mapet, na.rm=TRUE),
                            mappet = mean(mappet, na.rm=TRUE),
                            mavpd15 = mean(mavpd15,na.rm=TRUE)),
                           by=.(x,y,hydro_year_c)]
dat_annual <- dat_annual[is.infinite(frac_ppet_anom)==F]
```

## Regression: Long-term rate of NDVI increase per year

fraction of area experiencing greening:

```
sum(lt_ndvi_year$b1>0)/nrow(lt_ndvi_year)
```

```
## [1] 0.9046721
```

## Fit GAM of CO2 effect and climate

```
dat_train <- dat_annual[sample(.N, floor(nrow(dat_annual)/2))]
dat_test <- fsetdiff(dat_annual,dat_train)
```

```

g3_eval <- bam(ndvi_hyb~
                te(mappet,co2,k=5)+                      # MAP:MAPET x CO2, with ~5 knots
                te(mavpd15,frac_vpd_anom,k=5,bs='cs')+   # VPD x VPD_anom ...
                te(map,frac_p_anom,k=5,bs='cs')+          # MAP x P_anom ...
                te(mapet,frac_pet_anom,k=5,bs='cs')+      # MAPET x PET_anom ...
                epoch,
                data=dat_train,
                discrete = T, #
                select=TRUE, # turns on penalization term for splines
                method='fREML') # restricted maximum likelihood

```

## Out of sample R2 of GAM

```

yardstick::rsq_trad_vec(truth=dat_test$ndvi_hyb, estimate=predict(g3_eval,newdata=dat_test))

## [1] 0.9029181

```

## RMSE of GAM

```

yardstick::rmse_vec(truth=dat_test$ndvi_hyb, estimate=predict(g3_eval,newdata=dat_test))

## [1] 0.05093744

```

## Re-fit GAM with full data

```

g3_full <- bam(ndvi_hyb~
                  te(mappet,co2,k=5)+
                  te(mavpd15,frac_vpd_anom,k=5,bs='cs')+
                  te(map,frac_p_anom,k=5,bs='cs')+
                  te(mapet,frac_pet_anom,k=5,bs='cs')+
                  epoch,
                  data=dat_annual,
                  discrete = T,select=TRUE, method='fREML')

```

## Regression: Robust linear model (RLM) CO2 + VPD + P + PET

## Regression Thiel Sen: VPD

```

system.time(
  lt_v_sen <- dat[date>=ymd("1981-12-01")][date<=ymd("2019-11-30")] %>%
    .[,`:=` (hydro_year_c = hydro_year-1982)] %>%
    .[is.na(vpd15)==F] %>%
    .[,.(vpd15 = mean(vpd15,na.rm=TRUE)), by=.(x,y,hydro_year_c)] %>%
    .[,.(beta = list(coef(zyp.sen(vpd15~hydro_year_c)))),by=.(x,y)] %>%

```

```

. [, `:=` (b0=unlist(beta) [1],
           b1=unlist(beta) [2]), by=.(x,y)]
)

```

```

##      user  system elapsed
##  28.300 15.309 12.131

```

## Regression Thiel Sen: sqrt(VPD)

```

system.time(
  lt_vsq_sen <- dat[date>=ymd("1981-12-01")][date<=ymd("2019-11-30")] %>%
  .[, `:=` (hydro_year_c = hydro_year-1982)] %>%
  .[is.na(vpd15)==F] %>%
  .[,.(vpd15 = mean(sqrt(vpd15),na.rm=TRUE)), by=.(x,y,hydro_year_c)] %>%
  .[,.(beta = list(coef(zyp.sen(vpd15~hydro_year_c)))),by=.(x,y)] %>%
  .[, `:=` (b0=unlist(beta) [1],
            b1=unlist(beta) [2]), by=.(x,y)]
)

##      user  system elapsed
##  30.678 14.863 14.757

```

## Construct relative increase of VPD in 2019

```

# mavpd15 is annual mean from 1982:2011
pred_vpd_e1 <- merge(unique(dat_annual[,.(x,y,mavpd15)]), lt_v_sen[,.(x,y,b0)], by=c("x","y")) %>%
  .[, `:=` (e='e1',
            frac_vpd_anom = (b0-mavpd15)/mavpd15)]

pred_vpd_e2 <- lt_v_sen[, `:=` (vpd15_0 = b0,
                                 vpd15_1 = b0+16*b1,
                                 vpd15_2 = b0+37*b1)] %>%
  merge(unique(dat_annual[,.(x,y,mavpd15)]), ., by=c("x","y")) %>%
  .[, `:=` (frac_vpd_anom = (vpd15_2-mavpd15)/mavpd15,
            e='e2')]

```

## Percent increase of VPD, NDVI, Ca

```

lt_v_sen <- lt_v_sen[b0 > 0]
dVPD_VPD_sen <- mean(37*lt_v_sen$b1,na.rm=TRUE)/mean(lt_v_sen$b0,na.rm=TRUE)

# Actual percent increase in NDVI
lt_ndvi_year <- lt_ndvi_year[b0 > 0]
dNDVI_NDVI_rlm <- mean(37*lt_ndvi_year$b1,na.rm=TRUE)/mean(lt_ndvi_year$b0,na.rm=TRUE)

# Percent increase in Ca
dCa_Ca <-

```

```

diff(range(mlo[date>=ymd("1981-12-01") & date <= ymd("2019-08-30")]$co2_trend))/
  mean(mlo[date>=ymd("1981-12-01") & date <= ymd("1982-11-01")]$co2_trend)
dCa_Ca_e1 <-
  diff(range(mlo[date>=ymd("1981-12-01") & date <= ymd("2000-11-30")]$co2_trend))/
  mean(mlo[date>=ymd("1981-12-01") & date <= ymd("1982-11-01")]$co2_trend)
dCa_Ca_e2 <-
  diff(range(mlo[date>=ymd("2000-12-01") & date <= ymd("2019-08-30")]$co2_trend))/
  mean(mlo[date>=ymd("2000-12-01") & date <= ymd("2001-11-01")]$co2_trend)

# Expected WUE related increase using even split of WUE benefit (Donohue 2013)
0.5*(dCa_Ca - 0.5*dVPD_VPD_sen)

```

## [1] 0.08850517

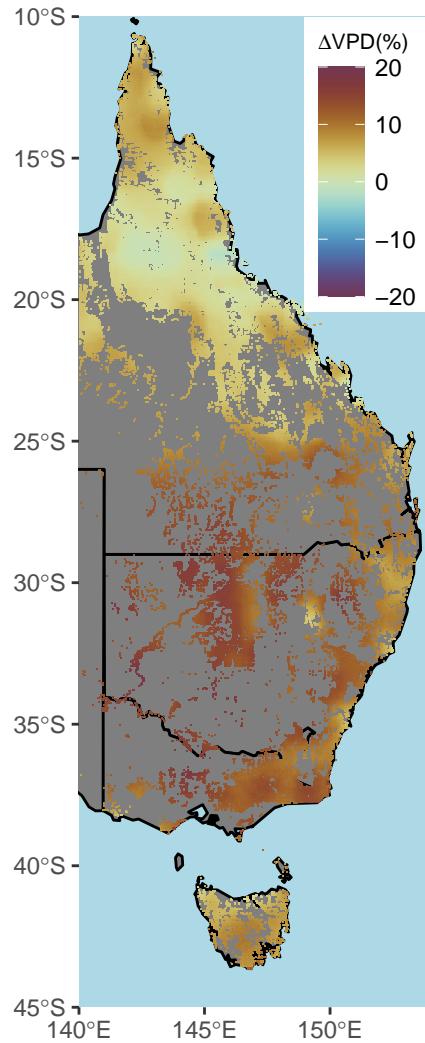
## Plotting —

### Relative change in VPD from 1982-2019

```

p_vpd_sen <- lt_v_sen %>%
  as_tibble() %>%
  filter(between(b1,-0.1,0.1)) %>%
  filter(b0 > 0) %>%
  ggplot(data=.,aes(x,y,fill=100*38*b1/b0))+ 
  geom_sf(data=oz_poly, inherit.aes = F, fill='gray50',color='black')+ 
  geom_tile()+
  scale_x_continuous(breaks=seq(140,154,by=5))+ 
  coord_sf(xlim = c(140,154),
            ylim = c(-45,-10), expand = FALSE)+ 
  labs(x=NULL,y=NULL)+ 
  scico::scale_fill_scico(expression(paste(Delta, "VPD(%))), 
                           palette ='roma0', direction=-1,
                           limits=c(-20,20),
                           oob=scales::squish)+ 
  # scale_fill_viridis_c(expression(paste(Delta, "VPD(%))), 
  #   option='A', limits=c(0,20), oob=scales::squish)+ 
  theme(panel.background = element_rect(fill='lightblue'),
        panel.grid = element_blank(),
        legend.title = element_text(size=8),
        legend.position = c(1,1),
        legend.justification = c(1,1)); p_vpd_sen

```



## WUE pred using 0.5 allocation to foliar

```

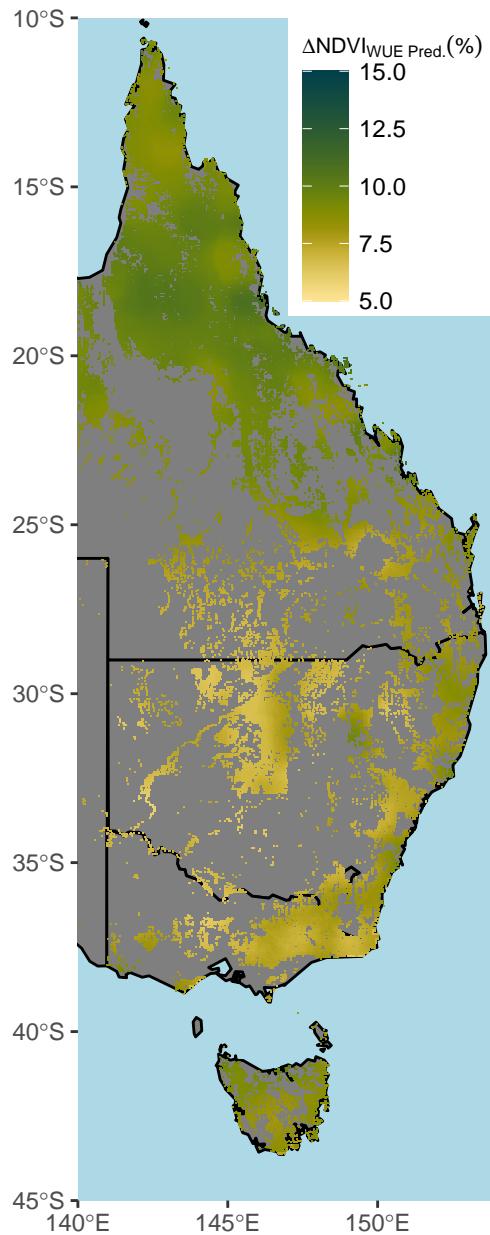
p_wue_sen <- lt_v_sen %>%
  as_tibble() %>%
  filter(b0 > 0) %>% # remove bad fits (don't really exist)
  filter(between(b1,-0.05,0.05)) %>% # remove bad fits (don't really exist)
  mutate(dVPD_VPD = b1*37/b0) %>% # relative fractional change in VPD
  mutate(expectation = 0.5*(dCa_Ca - 0.5*dVPD_VPD)) %>% # 0.5 alloc WUE pred
  # pull(expectation) %>% quantile(., c(0.01,0.99))
  ggplot(data=.,aes(x,y,fill=expectation*100))+ 
  geom_sf(data=oz_poly, inherit.aes = F, fill='gray50',color='black')+ 
  geom_tile()+
  scale_x_continuous(breaks=seq(140,154,by=5))+ 
  coord_sf(xlim = c(140,154),
            ylim = c(-45,-10), expand = FALSE)+ 
  labs(x=NULL,y=NULL)+ 
  scico::scale_fill_scico(expression(paste(Delta*NDVI[WUE~Pred.](%""))),
                           palette = 'bamako',

```

```

        direction = -1,
        limits=c(5,15), #na.value = 'red',
        oob=scales::squish
)+>
theme(panel.background = element_rect(fill='lightblue'),
      panel.grid = element_blank(),
      legend.title = element_text(size=8),
      legend.position = c(1,1),
      legend.justification = c(1,1)); p_wue_sen

```



GAM estimated CO[2] contribution to NDVI where the effect of changes in P and PET are removed and the effect of increasing VPD is retained.

```

p_gam_pred <- bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),
                         tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%
inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2'))),
            by=c('e')) %>%
mutate(frac_p_anom=0,
       frac_ppet_anom=0,
       frac_pet_anom=0,
       # frac_vpd_anom=0,
       epoch=0) %>%
inner_join(., tibble(unique(dat_annual[,(x,y,map,mapet,mappet,mavpd15)]))), by=c('x','y')) %>%
inner_join(., kop,by=c('x','y')) %>%
mutate(pred = predict(g3_full,newdata=.,type='response')) %>%
filter(is.na(pred)==F) %>%
select(x,y,e,frac_vpd_anom,pred,co2) %>%
pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
mutate(dNDVI = 100*((pred_e2-pred_e1)/pred_e1)) %>%
mutate(d_vpd = frac_vpd_anom_e2 - frac_vpd_anom_e1) %>%
ggplot(data=.,aes(x,y,fill=dNDVI))+
geom_sf(data=oz_poly, inherit.aes = F, fill='gray50',color='black')+  

geom_tile()+
scale_x_continuous(breaks=seq(140,154,by=5))+  

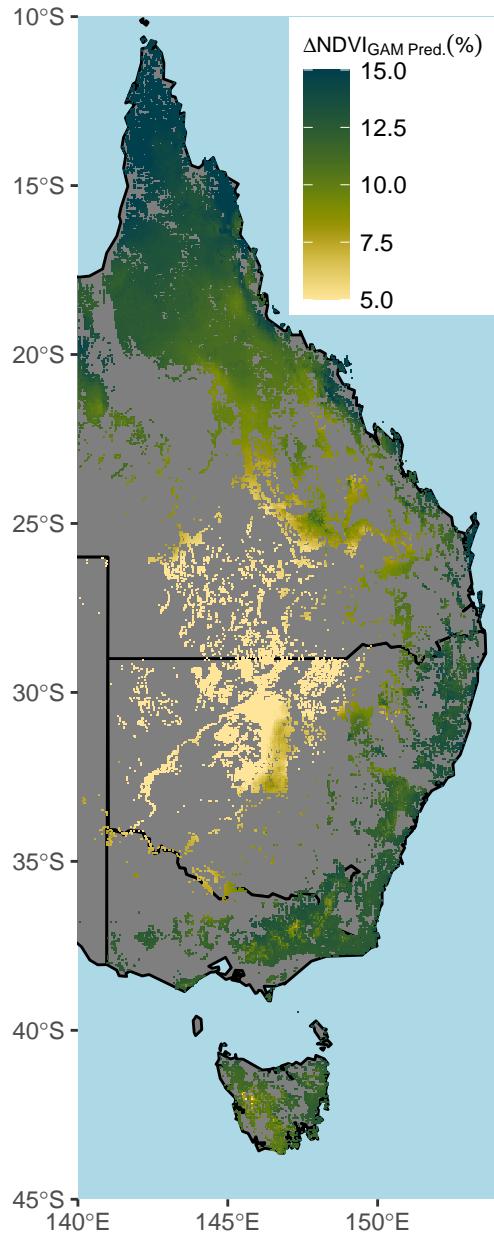
coord_sf(xlim = c(140,154),
          ylim = c(-45,-10), expand = FALSE)+  

labs(x=NULL,y=NULL)+  

scico::scale_fill_scico(expression(paste(Delta*NDVI [GAM~Pred.] ("%"))),
                        palette = 'bamako',
                        direction = -1,
                        limits=c(5,15), #na.value = 'red',
                        oob=scales::squish
)+  

theme(panel.background = element_rect(fill='lightblue'),
      panel.grid = element_blank(),
      legend.title = element_text(size=8),
      legend.position = c(1,1),
      legend.justification = c(1,1)); p_gam_pred

```



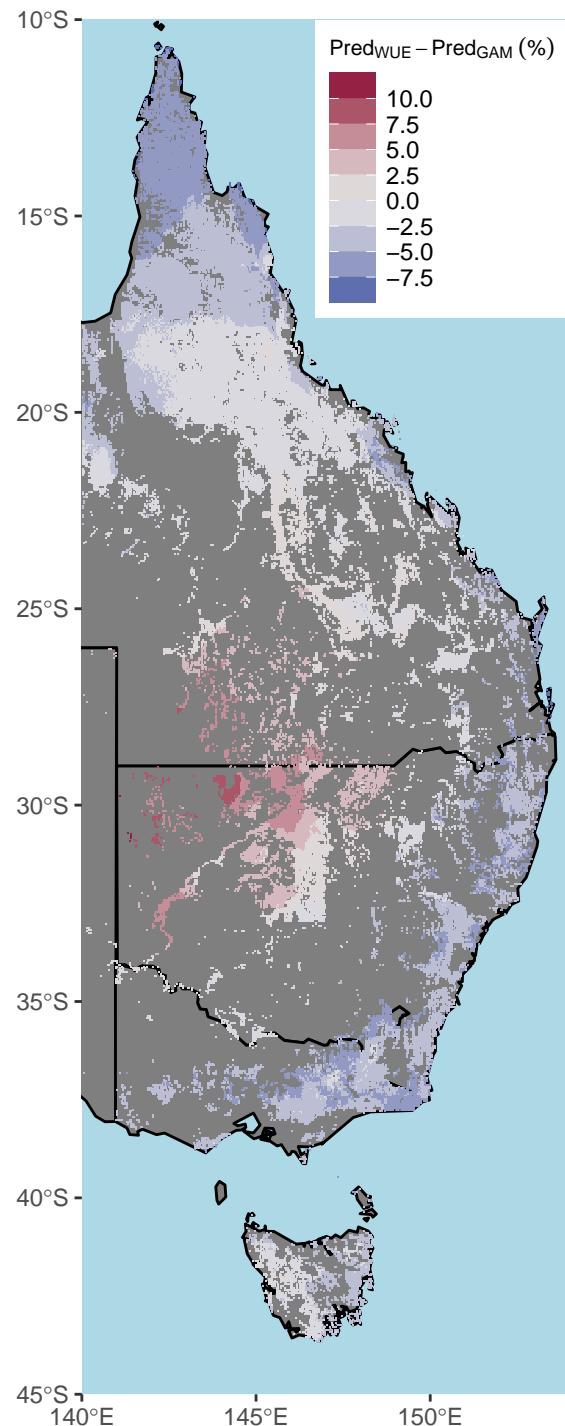
### Plot residuals between 0.5\*WUE and GAM

```
p_res <- inner_join({lt_v_sen %>%
  as_tibble() %>%
  filter(b0 > 0) %>%
  filter(between(b1,-0.05,0.05)) %>%
  mutate(dVPD_VPD = b1*37/b0) %>%
  mutate(expectation = 100*0.5*(dCa_Ca - 0.5*dVPD_VPD)) %>%
  select(x,y,expectation)},
  {bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),
    tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%
  inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2'))),
```

```

        by=c('e')) %>%
mutate(frac_p_anom=0,
       frac_ppet_anom=0,
       frac_pet_anom=0,
       # frac_vpd_anom=0,
       epoch=0) %>%
inner_join(., tibble(unique(dat_annual[,.(x,y,map,mapet,mappet,mavpd15)]))), by=c('x','y')) %>%
inner_join(., kop,by=c('x','y')) %>%
mutate(pred = predict(g3_full,newdata=.,type='response')) %>%
filter(is.na(pred)==F) %>%
select(x,y,e,frac_vpd_anom,pred,co2) %>%
pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
mutate(dNDVI = 100*((pred_e2-pred_e1)/pred_e1)) %>%
mutate(d_vpd = frac_vpd_anom_e2 - frac_vpd_anom_e1) %>%
select(x,y,dNDVI)}, by=c('x','y')) %>%
ggplot(data=.,aes(x,y,fill=expectation-dNDVI))+
geom_sf(data=oz_poly, inherit.aes = F, fill='gray50',color='black')+geom_tile()+
scale_x_continuous(breaks=seq(140,154,by=5))+coord_sf(xlim = c(140,154),
            ylim = c(-45,-10), expand = FALSE)+
labs(x=NULL,y=NULL,fill=expression(paste(Pred[WUE]-Pred[GAM]~(%%))))+
scale_fill_binned_diverging(palette='Blue-Red',rev=F,n.breaks=10,
                           # cmax=90,
                           p1=1.5,p2=1.5)+
# scico::scale_fill_scico(expression(paste(Delta*NDVI[WUE~Pred.] ("")-Delta*NDVI[GAM~Pred.] (""))),
# #                         palette = 'roma',
# #                         direction = -1,
# #                         limits=c(-15,15), #na.value = 'red',
# #                         oob=scales::squish
# # )+
guides(fill=guide_colorbar())+
theme(panel.background = element_rect(fill='lightblue'),
      panel.grid = element_blank(),
      legend.title = element_text(size=8),
      legend.position = c(1,1),
      legend.justification = c(1,1)); p_res

```



```

p_res <- inner_join({lt_v_sen %>%
  as_tibble() %>%
  filter(b0 > 0) %>%
  filter(between(b1, -0.05, 0.05)) %>%
  mutate(dVPD_VPD = b1*37/b0) %>%
  mutate(expectation = 100*0.5*(dCa_Ca - 0.5*dVPD_VPD)) %>%
  select(x,y,expectation)},
{bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),

```

```

    tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%
inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2')),
            by=c('e')) %>%
mutate(frac_p_anom=0,
       frac_ppet_anom=0,
       frac_pet_anom=0,
       # frac_vpd_anom=0,
       epoch=0) %>%
inner_join(., tibble(unique(dat_annual[,(x,y,map,mapet,mappet,mavpd15)])), by=c('x','y')) %>%
inner_join(., kop,by=c('x','y')) %>%
mutate(pred = predict(g3_full,newdata=.,type='response')) %>%
filter(is.na(pred)==F) %>%
select(x,y,e,frac_vpd_anom,pred,co2) %>%
pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
mutate(dNDVI = 100*((pred_e2-pred_e1)/pred_e1)) %>%
mutate(d_vpd = frac_vpd_anom_e2 - frac_vpd_anom_e1) %>%
select(x,y,dNDVI)}, by=c('x','y')) %>%
ggplot(data=.,aes(x,y,fill=expectation-dNDVI))+
geom_sf(data=oz_poly, inherit.aes = F, fill='gray50',color='black')+  

geom_tile()+
scale_x_continuous(breaks=seq(140,154,by=5))+  

coord_sf(xlim = c(140,154),
          ylim = c(-45,-10), expand = FALSE)+  

labs(x=NULL,y=NULL,fill=expression(paste(Pred[WUE]-Pred[GAM]~( '%'))))+  

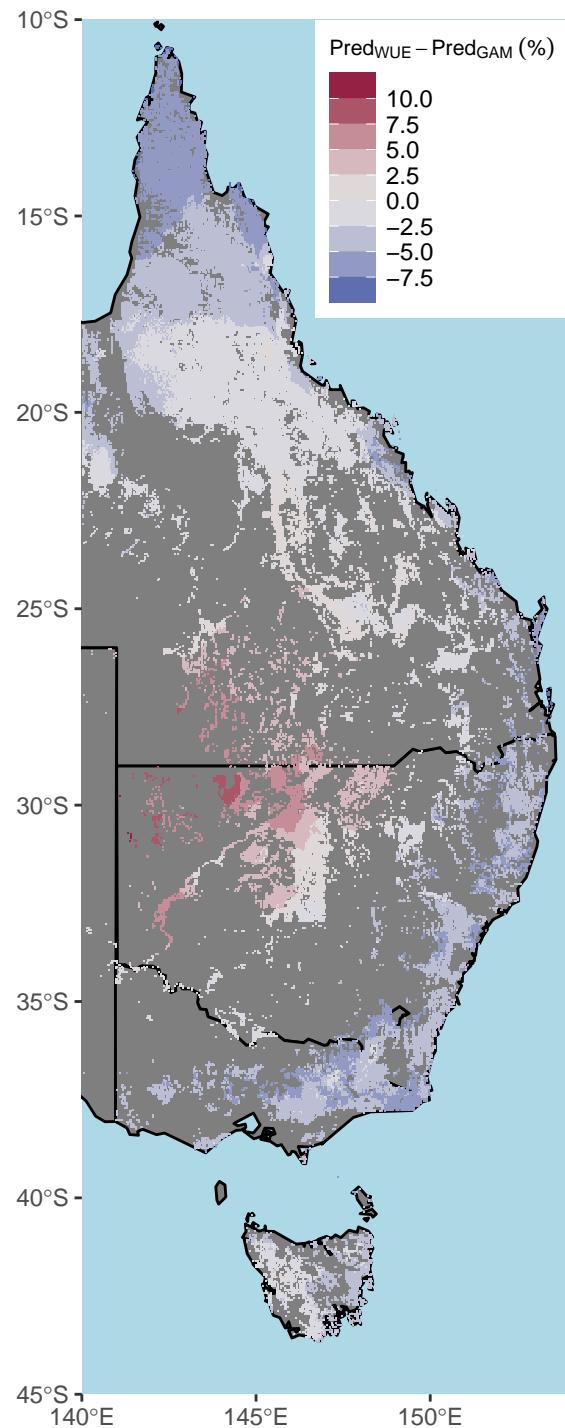
scale_fill_binned_diverging(palette='Blue-Red',rev=F,n.breaks=10,
                            # cmax=90,
                            p1=1.5,p2=1.5)+  

# scico::scale_fill_scico(expression(paste(Delta*NDVI[WUE~Pred.]("%")-Delta*NDVI[GAM~Pred.]("%"))),  

#                         palette = 'roma',
#                         direction = -1,
#                         limits=c(-15,15), #na.value = 'red',
#                         oob=scales::squish
# )+  

guides(fill=guide_colorbar())+
theme(panel.background = element_rect(fill='lightblue'),
      panel.grid = element_blank(),
      legend.title = element_text(size=8),
      legend.position = c(1,1),
      legend.justification = c(1,1)); p_res

```



Boxplots WUE preds. (0.25,0.5,0.75,1 & GAM estimate

```

vec_cols <- scico::scico(n=4, palette = 'bamako',
                           direction = -1, end=0.9, begin=0.2)
p_box <- bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),
                    tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%

```

```

inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2')),
           by=c('e')) %>%
mutate(frac_p_anom=0,
       frac_ppet_anom=0,
       frac_pet_anom=0,
       epoch=0) %>%
inner_join(., tibble(unique(dat_annual[,(x,y,map,mapet,mappet,mavpd15)])), by=c('x','y')) %>%
inner_join(., kop,by=c('x','y')) %>%
mutate(pred = predict(g3_full,newdata=.,type='response')) %>%
filter(is.na(pred)==F) %>%
select(x,y,e,frac_vpd_anom,pred,co2) %>%
pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
mutate(dNDVI = 100*((pred_e2-pred_e1)/pred_e1)) %>%
inner_join(., {
  lt_v_sen %>%
  as_tibble() %>%
  filter(b0 > 0) %>%
  filter(between(b1,-0.05,0.05)) %>%
  mutate(dVPD_VPD = b1*37/b0) %>%
  mutate(wue25 = 100*(0.25*(dCa_Ca - 0.5*dVPD_VPD)),
         wue50 = 100*(0.5*(dCa_Ca - 0.5*dVPD_VPD)),
         wue75 = 100*(0.75*(dCa_Ca - 0.5*dVPD_VPD)),
         wue100 = 100*(1*(dCa_Ca - 0.5*dVPD_VPD))) %>%
  select(x,y,wue25,wue50,wue75,wue100)
},by=c("x","y")) %>%
inner_join(., kop, by=c('x','y')) %>%
mutate(res25 = dNDVI-wue25,
       res50 = dNDVI-wue50,
       res75 = dNDVI-wue75
) %>%
select(zone,
       wue25,
       wue50,
       wue75,
       wue100,
       dNDVI
) %>%
gather(-zone,key='allocation',
       value='res') %>% #ggplot(data=.,aes(value,y=zone,fill=key))+geom_boxplot()
mutate(allocation = factor(allocation,
                           levels=c("wue25","wue50","wue75","wue100","dNDVI"),
                           labels = c('10%','50%','90%','100%','GAM. Estimate'),
                           ordered = TRUE)) %>%
ggplot(data=.,aes(y=res,
                   x=zone,
                   fill=allocation))+
geom_hline(aes(yintercept=0),color='grey50',lty=3)+
geom_boxplot(#draw_quantiles = c(0.25,0.5,0.75),
            # trim=TRUE,
            outlier.colour = NA,
            color='black')+
# scico::scale_fill_scico_d(
#   palette='bamako',direction = -1, begin = 0.2,end=0.95,

```

```

#   guide=guide_legend(title.position = 'top'))+
scale_fill_manual(#"% WUE benefit allocated to foliar area:",
  values=c(vec_cols[1], vec_cols[2], vec_cols[3], vec_cols[4],"grey"))+
labs(x=NULL,
  y=expression(paste(Predicted~Delta*NDVI, " (%)" )),
  fill=expression(paste(C0[2]*' x '*WUE, ' gain allocated to foliar area')))+
```

scale\_x\_discrete(limits=(structure(c(1L,2L,3L,4L,5L,6L,7L),# c(5L, 4L, 6L, 2L, 1L, 3L, 7L),
 .Label = c("Equatorial",
 "Tropical",
 "Subtropical",
 "Grassland",
 "Arid",
 "Temperate",
 "Temperate Tas."),
 class = c("ordered", "factor"))))+

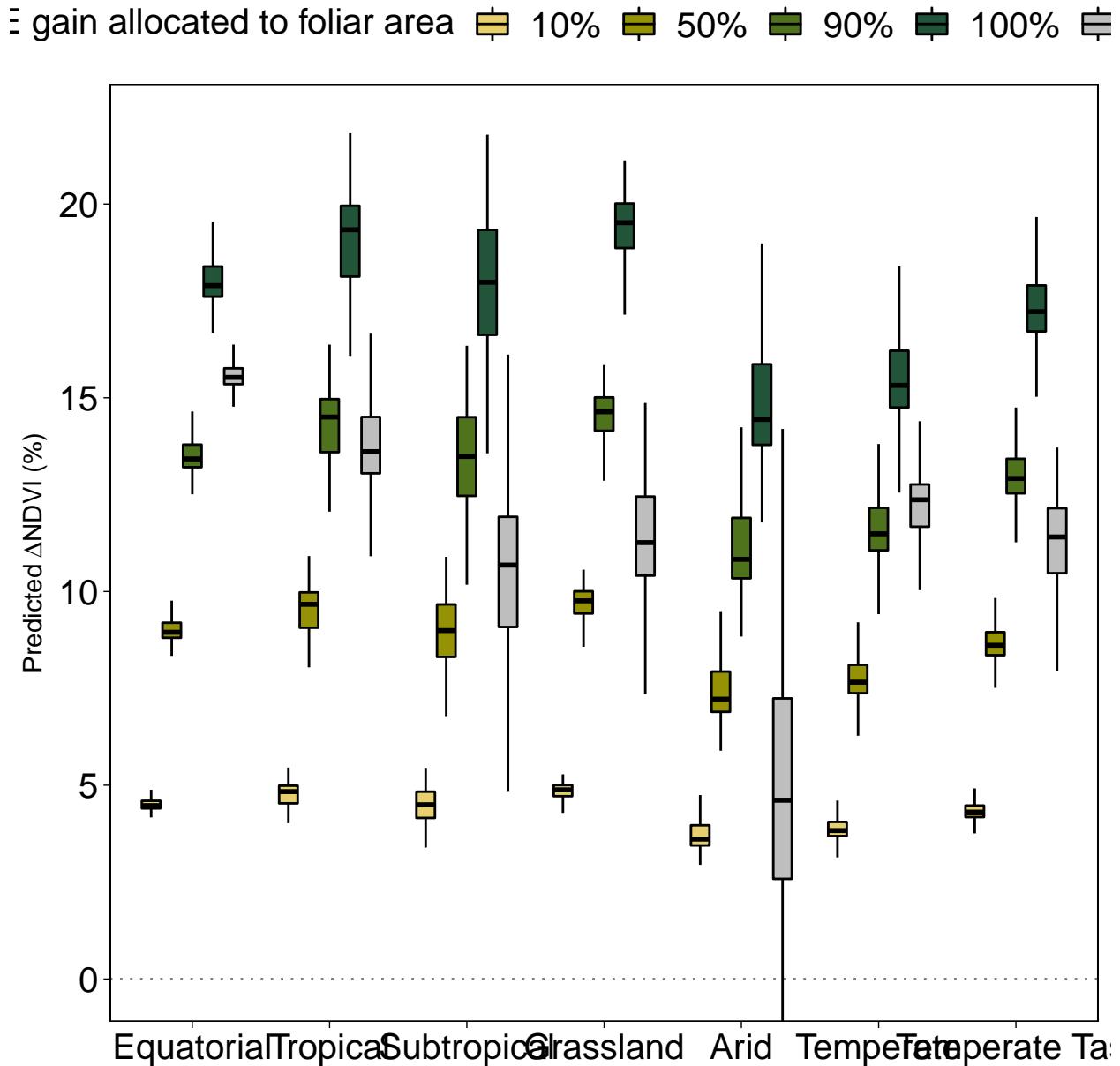
```

coord_cartesian(ylim = c(0,22))+
```

```

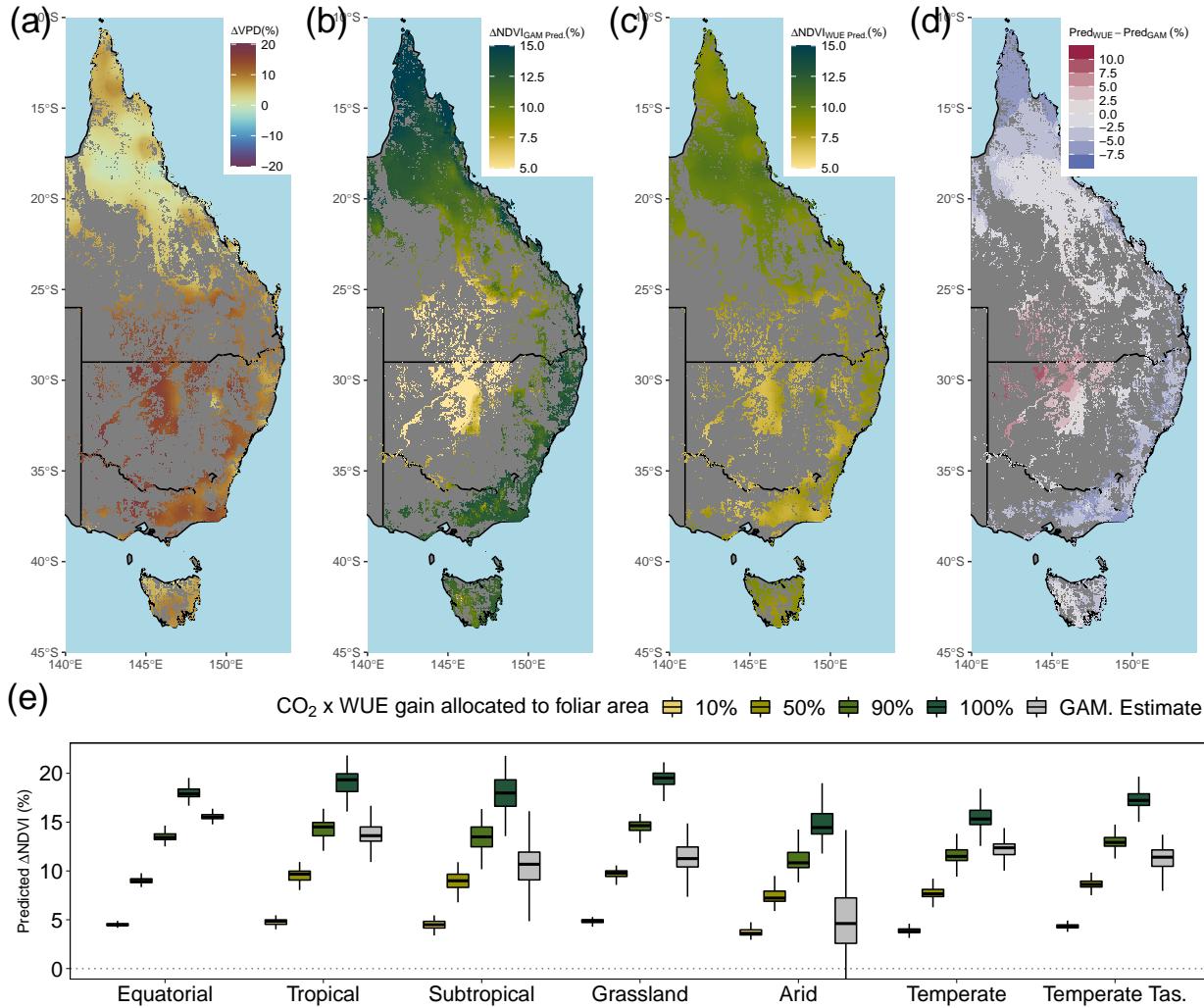
# coord_flip()+
theme_linedraw()+
guides(fill=guide_legend(title.position='left'))+
theme(legend.position = 'top',
  legend.justification = c(1,1),
  legend.text = element_text(size=15),
  legend.title = element_text(size=15),
  # legend.key.height = unit(0.2, 'cm'),
  axis.text = element_text(size=15),
  panel.grid = element_blank());
```

p\_box



merge figures

```
library(cowplot)
p_left <- ggdraw(p_vpd_sen)+draw_label(label='(a)', x=0.07,y=0.985,size = 25)
p_mid1 <- ggdraw(p_gam_pred)+draw_label(label='(b)',x=0.05,y=0.985,size=25)
p_mid2 <- ggdraw(p_wue_sen)+draw_label(label='(c)', x=0.05,y=0.985, size=25)
p_right <- ggdraw(p_res)+draw_label(label = '(d)', x=0.05,y=0.985,size=25)
p_bottom <- ggdraw(p_box)+draw_label(label = '(e)', x=0.015,y=0.95,size=25)
cp_r <- cowplot::plot_grid(p_left,p_mid1,p_mid2,p_right,
                           nrow = 1,
                           rel_widths = c(1,1,1,1))
cp_r/p_bottom+patchwork::plot_layout(heights = c(1,0.5))
```



### Prepare PETA prediction comparison

```
# PETA prediction
dono17 <- {lt_vsq_sen %>%
  as_tibble() %>%
  rowwise() %>%
  filter(b0 > 0) %>%
  # filter(between(b1, -0.05, 0.05)) %>%
  mutate(dVPD_VPD = b1*37/b0) %>%
  mutate(wue_expectation = 0.5*(dCa_Ca - 0.5*dVPD_VPD)) %>% # 0.5*WUE benefit
  mutate(peta = (1+dCa_Ca)/(1+dVPD_VPD) - 1) %>% # eq2 (Donohue 2017)
  # mutate(peta = (1+(dCa_Ca/341.16))/(
  #           (1+(dVPD_VPD/b0))) - 1) %>%
  select(x,y,wue_expectation,peta)} %>%
inner_join(., lai, by=c('x','y')) %>% # join LAI
mutate(alpha = 1-exp(-0.5*lai)) %>% # eq 6
mutate(beta = 1/(1+peta) - 1) %>% # eq 12
mutate(dL = peta*(1-alpha)**2) %>% # eq 9
mutate(dE = -beta*alpha + beta)
```

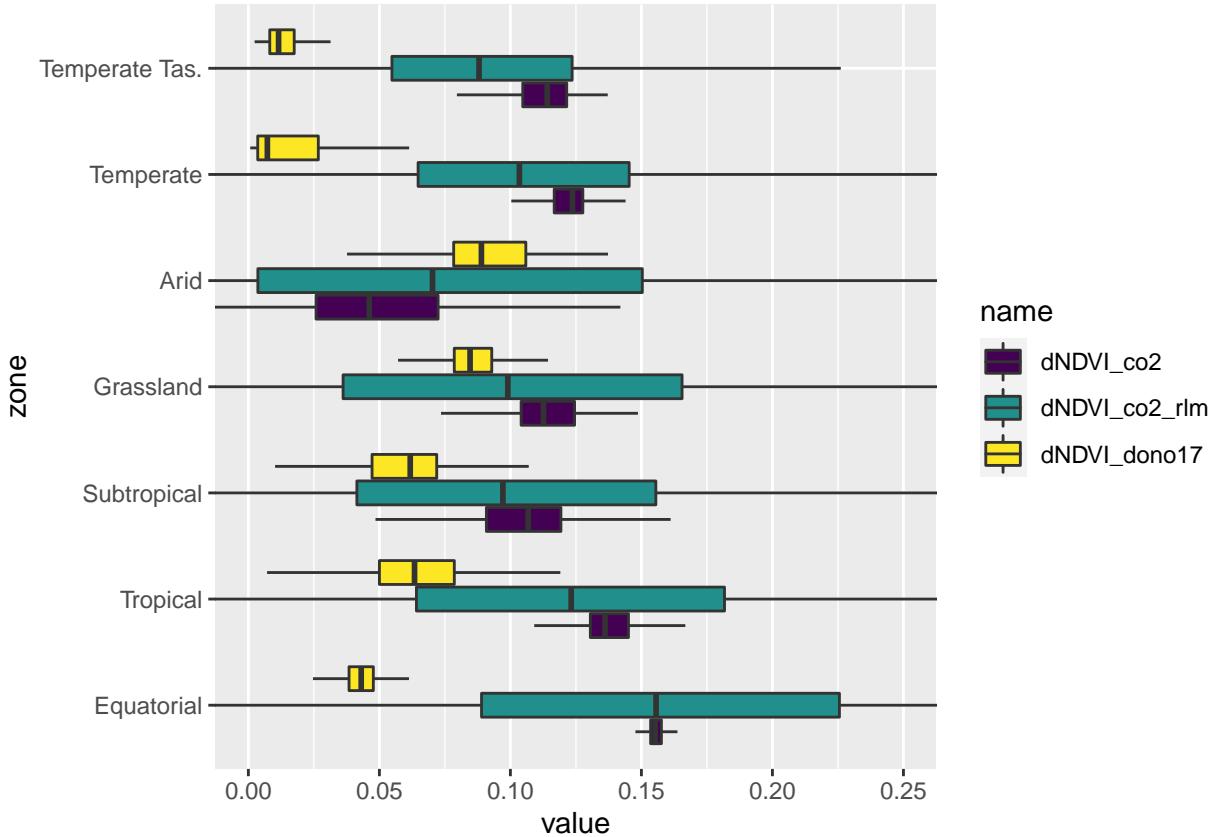
```

# GAM estimate
gest <- bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),
                  tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%
  inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2')),
             by=c('e')) %>%
  mutate(frac_p_anom=0,
         frac_ppet_anom=0,
         frac_pet_anom=0,
         # frac_vpd_anom=0,
         epoch=0) %>%
  inner_join(., tibble(unique(dat_annual[,(x,y,map,mapet,mappet,mavpd15)])), by=c('x','y')) %>%
  inner_join(., kop,by=c('x','y')) %>%
  mutate(pred = predict(g3_full,newdata=.,type='response')) %>%
  filter(is.na(pred)==F) %>%
  select(x,y,e,frac_vpd_anom,pred,co2) %>%
  pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
  mutate(dNDVI_co2 = ((pred_e2-pred_e1)/pred_e1)) %>%
  mutate(d_vpd = frac_vpd_anom_e2 - frac_vpd_anom_e1) %>%
  select(x,y,dNDVI_co2)

# RLM estimates
lest <- bind_rows(tibble(pred_vpd_e1) %>% select(x,y,e,frac_vpd_anom),
                  tibble(pred_vpd_e2) %>% select(x,y,e,frac_vpd_anom)) %>%
  inner_join(., tibble(co2=c(340.8,411.9),e=c('e1','e2')),
             by=c('e')) %>%
  mutate(frac_p_anom=0,
         frac_ppet_anom=0,
         frac_pet_anom=0,
         epoch=0) %>%
  inner_join(., tibble(unique(dat_annual[,(x,y,map,mapet,mappet,mavpd15)])), by=c('x','y')) %>%
  inner_join(., kop,by=c('x','y')) %>%
  inner_join(., rlm_ndvi_annual_co2_ppet_epoch,by=c("x","y")) %>%
  mutate(pred = b0 + b1*(co2-min_co2)+b2*frac_vpd_anom) %>%
  filter(is.na(pred)==F) %>%
  select(x,y,e,frac_vpd_anom,pred,co2) %>%
  pivot_wider(., names_from=c('e'), values_from=c('pred','co2','frac_vpd_anom')) %>%
  mutate(dNDVI_co2_rlm = ((pred_e2-pred_e1)/pred_e1)) %>%
  mutate(d_vpd = frac_vpd_anom_e2 - frac_vpd_anom_e1) %>%
  select(x,y,dNDVI_co2_rlm)

# boxplot
inner_join(dono17, gest) %>%
  inner_join(., lest) %>%
  inner_join(., dono17 %>% rename(dNDVI_dono17 = dL)) %>%
  inner_join(., kop) %>%
  pivot_longer(cols=starts_with("dNDVI")) %>%
  ggplot(data=.,aes(value,zone,fill=name))+
  geom_boxplot(outlier.colour = NA)+
  coord_cartesian(xlim=c(0,0.25),clip = 'on')+
  scale_fill_viridis_d()

```



```

gest %>% rename(GAM = dNDVI_co2) %>%
  inner_join(., lest %>% rename(RLM=dNDVI_co2_rlm)) %>%
  inner_join(., dono17 %>% rename(PETA = dL)) %>%
  inner_join(., kop) %>%
  pivot_longer(cols=c(starts_with("dNDVI"),'PETA','RLM','GAM')) %>%
  ggplot(data=.,aes(lai,value,color=name))+ 
  geom_point(data=. %>% filter(name %in% c("RLM","GAM"))),
  aes(lai,value,color=name),
  alpha=0.03,size=0.5)+ 
  geom_smooth(method='gam',
  formula=y~s(x,bs='cs'),
  method.args=list(method='ML',
  select=TRUE))+ 
  scale_color_manual(#option='D',end=0.6,direction = 1,
  values=c("darkgreen",
  "black",
  "blue"),
  breaks=c("PETA","GAM","RLM"),
  labels=c("PETA Prediction","GAM Estimate","RLM Estimate"),
  # drop=TRUE,
  na.translate=FALSE)+ 
  labs(x=expression(paste('Leaf Area Index ',(m**2/m**2))), 
  color='',
  y=expression(paste(C0[2] ~'effect on NDVI')))+ 
  coord_cartesian(ylim=c(0,0.25),
  xlim=c(0,4.5),
  expand=0)

```

```

      expand=FALSE,
      default = FALSE,
      clip = 'on')+
theme_linedraw()+
theme(legend.position = c(1,1),
      legend.justification = c(0.99,0.99),
      legend.background = element_rect(fill=NA),
      panel.grid = element_blank())

```

