

# Spatio Temporal Gaussian Process Real

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library(mvtnorm)

space.kern <- function(D, l2){
  exp(-1/(2*l2)*D)
}

space.D <- function(lat_long){
  D <- as.matrix(dist(lat_long, diag = TRUE, upper = TRUE))^2
}

time.kern <- function(D, l2){
  exp(-1/(2*l2)*D)
}

time.D <- function(tt){
  D <- as.matrix(dist(tt, diag = TRUE, upper = TRUE))^2
}

gau.pro.spat.temp <- function(data, niter, l2_space = 0.005, l2_time = 10,
                              prior_a = 2, prior_b = 2, var_init = 2){
  Y <- log(data[,3]+1)
  tt <- unique(data[,2])

  Ymean <- mean(Y)
  Ysd <- sd(Y)
  Y <- (Y-Ymean)/Ysd

  lat_long <- unique(data[,4:5])
  rownames(lat_long) <- NULL

  N <- length(tt)*nrow(lat_long)

  D_space <- space.D(lat_long)
  K_space <- space.kern(D_space, l2_space) + diag(10e-06, nrow(lat_long))
  D_time <- time.D(tt)
  K_time <- time.kern(D_time, l2_time) + diag(10e-06, length(tt))

  K_space_inv <- solve(K_space)
  K_time_inv <- solve(K_time)

  f_save <- matrix(0, niter, length(Y))
  sig2_save <- rep(0, niter)
  tau2_space_save <- rep(0, niter)
  tau2_time_save <- rep(0, niter)

  sig2_a <- prior_a
  sig2_b <- prior_b
```

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tau2_time_a <- prior_a
tau2_time_b <- prior_b
tau2_space_a <- prior_a
tau2_space_b <- prior_b

sig2_save[1] <- var_init
tau2_space_save[1] <- var_init
tau2_time_save[1] <- var_init

for (i in 2:niter){
  K_space_time <- kronecker(tau2_space_save[i-1]*K_space,
                           tau2_time_save[i-1]*K_time)
  sig2I <- diag(sig2_save[i-1], N)
  f_Sigma <- K_space_time + sig2I

  f_Mu <- K_space_time%%solve(f_Sigma, Y)
  f_Sigma <- K_space_time - K_space_time%%solve(f_Sigma, K_space_time)

  f_save[i,] <- rmvnorm(1, f_Mu, f_Sigma)

  sig2_save[i] <- 1/rgamma(1, sig2_a + N/2,
                        sig2_b + 0.5*t(Y-f_save[i,])%%(Y-f_save[i,]))

  tau2_space_Sigma <- kronecker(K_space_inv, 1/tau2_time_save[i-1]*K_time_inv)

  tau2_space_save[i] <- 1/rgamma(1, tau2_space_a + N/2,
                              tau2_space_b + 0.5*t(f_save[i,])%%tau2_space_Sigma%%f_save[i,])

  tau2_time_Sigma <- kronecker(1/tau2_space_save[i]*K_space_inv, K_time_inv)

  tau2_time_save[i] <- 1/rgamma(1, tau2_time_a + N/2,
                              tau2_time_b + 0.5*t(f_save[i,])%%tau2_time_Sigma%%f_save[i,])
}

cbind(f_save, sig2_save, tau2_space_save, tau2_time_save)
}

load(file = "C://Users/david/Documents/R/Stat 695/Project/uber/trip_data.RData")

gaupro_out <- gau.pro.spat.temp(trip_data, niter = 100)

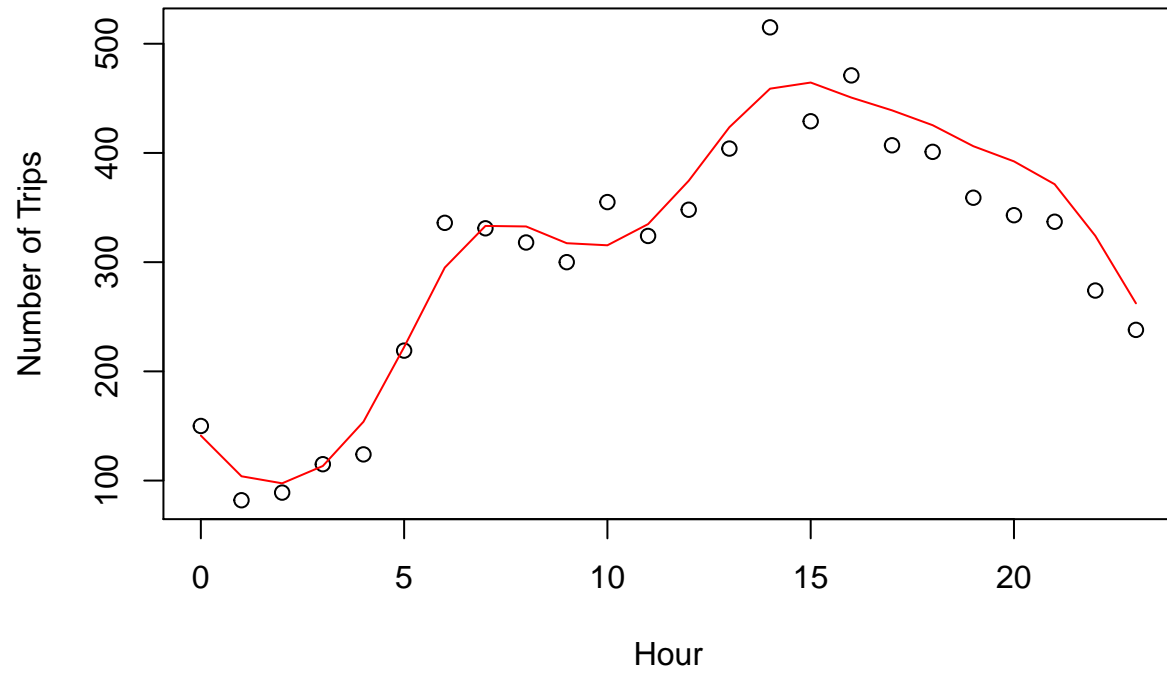
f_out <- gaupro_out[,1:600]
f_pred <- colMeans(exp(f_out*sd(log(trip_data$n_trips+1)) + mean(log(trip_data$n_trips+1))))

f_pred_mat <- matrix(f_pred, nrow = 24, ncol = 25, byrow = FALSE)
trip_mat <- matrix(trip_data$n_trips, nrow = 24, ncol = 25, byrow = FALSE)

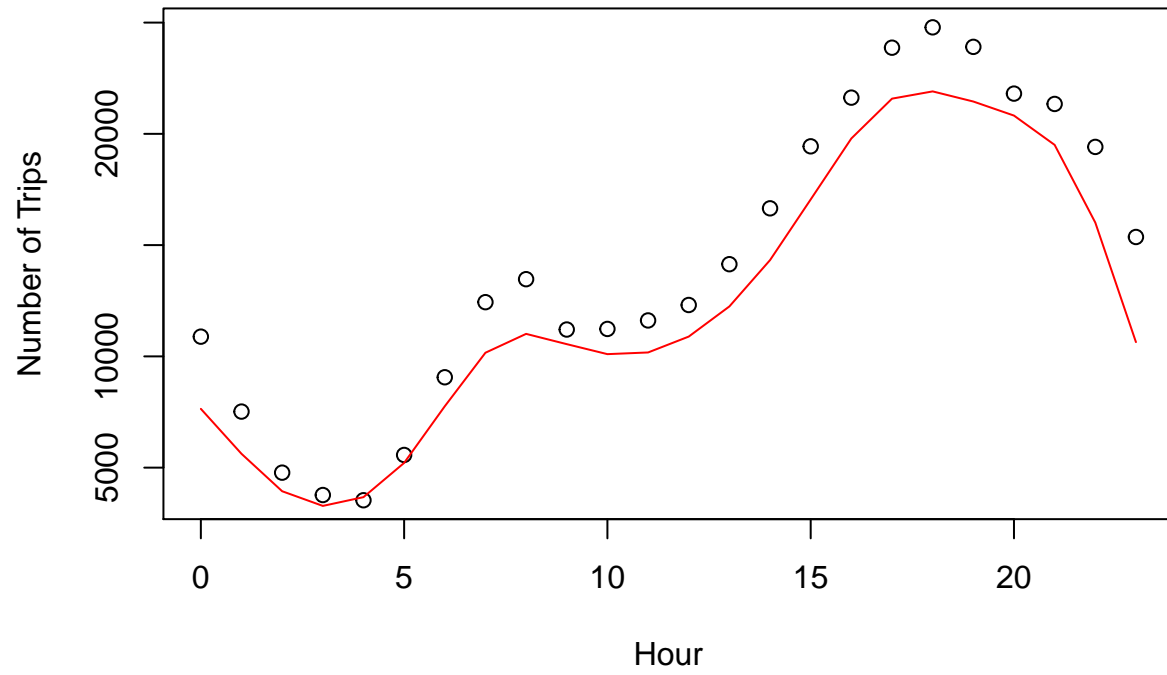
for (i in 1:25){
  plot(0:23, trip_mat[,i], main = paste("Number of Hourly Trips (with Predictions) for Location", i),
       xlab = "Hour", ylab = "Number of Trips")
  lines(0:23, f_pred_mat[,i], col = "red")
}

```

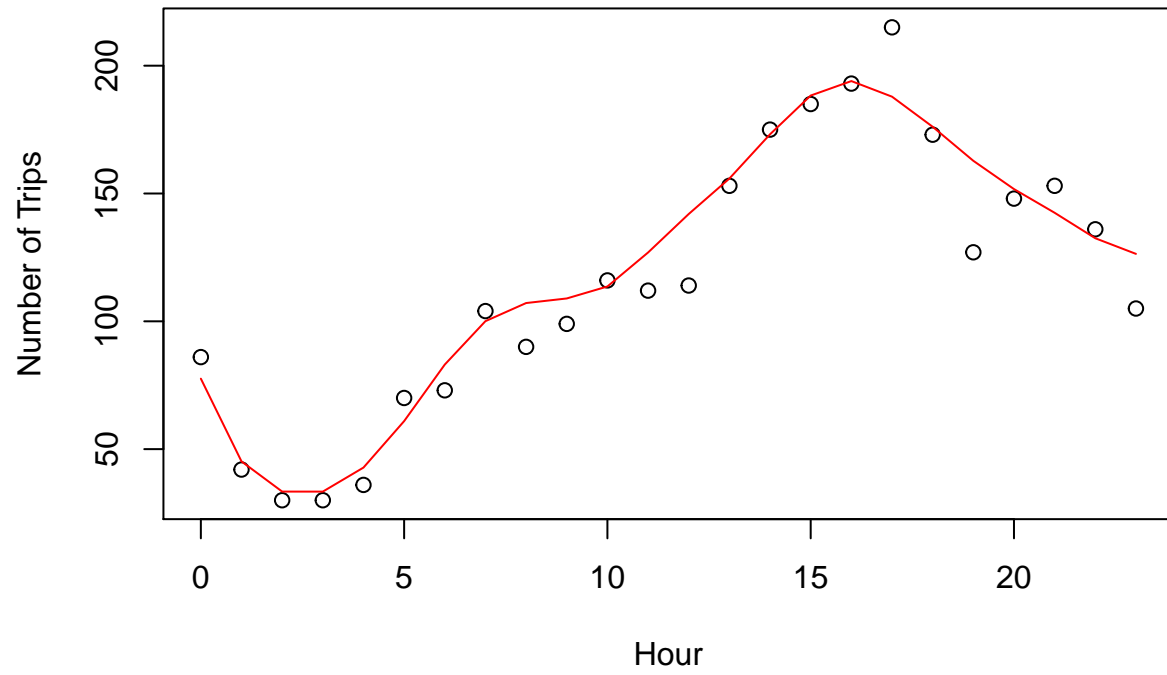
**Number of Hourly Trips (with Predictions) for Location 1**



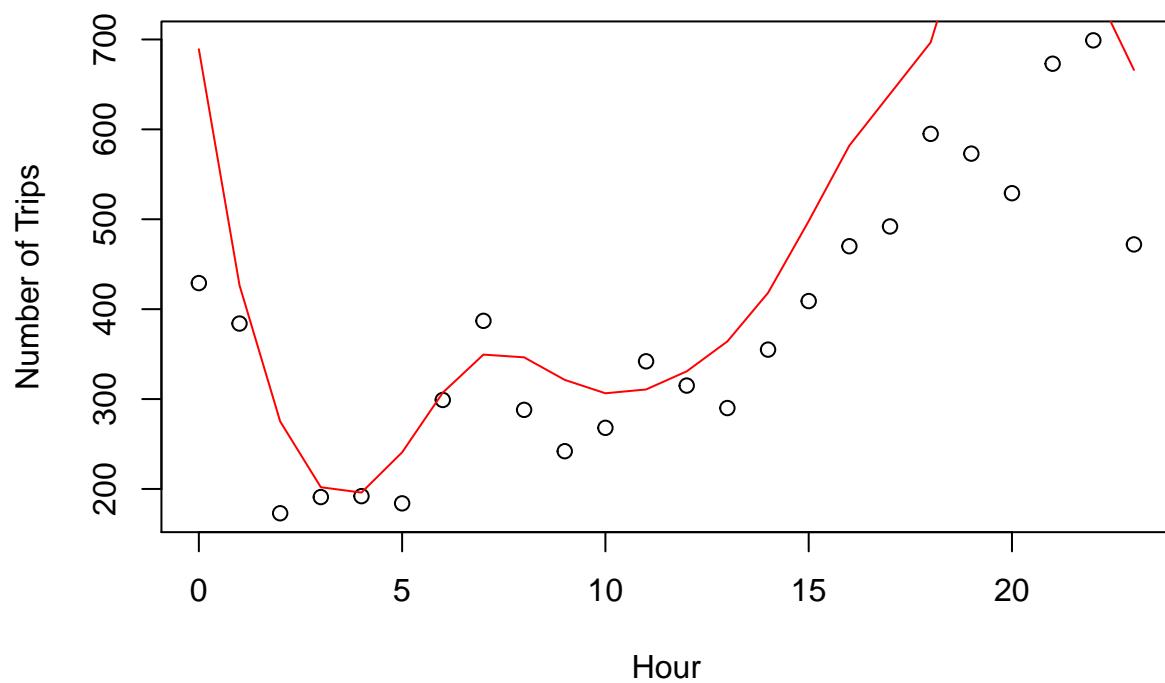
**Number of Hourly Trips (with Predictions) for Location 2**



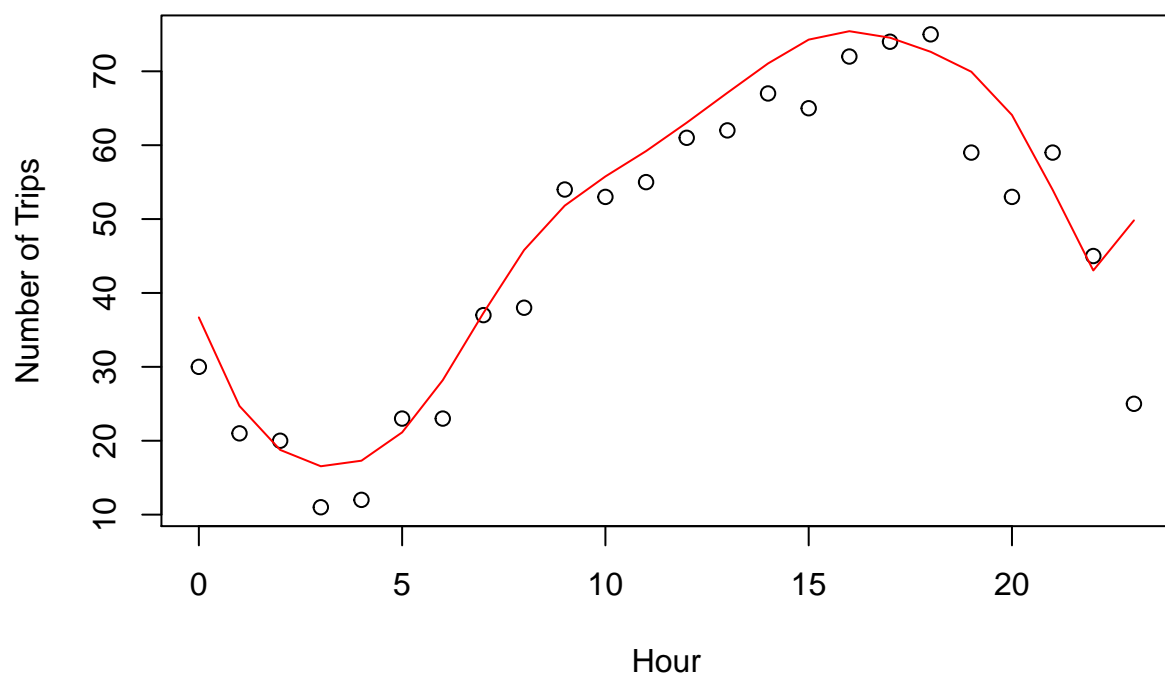
**Number of Hourly Trips (with Predictions) for Location 3**



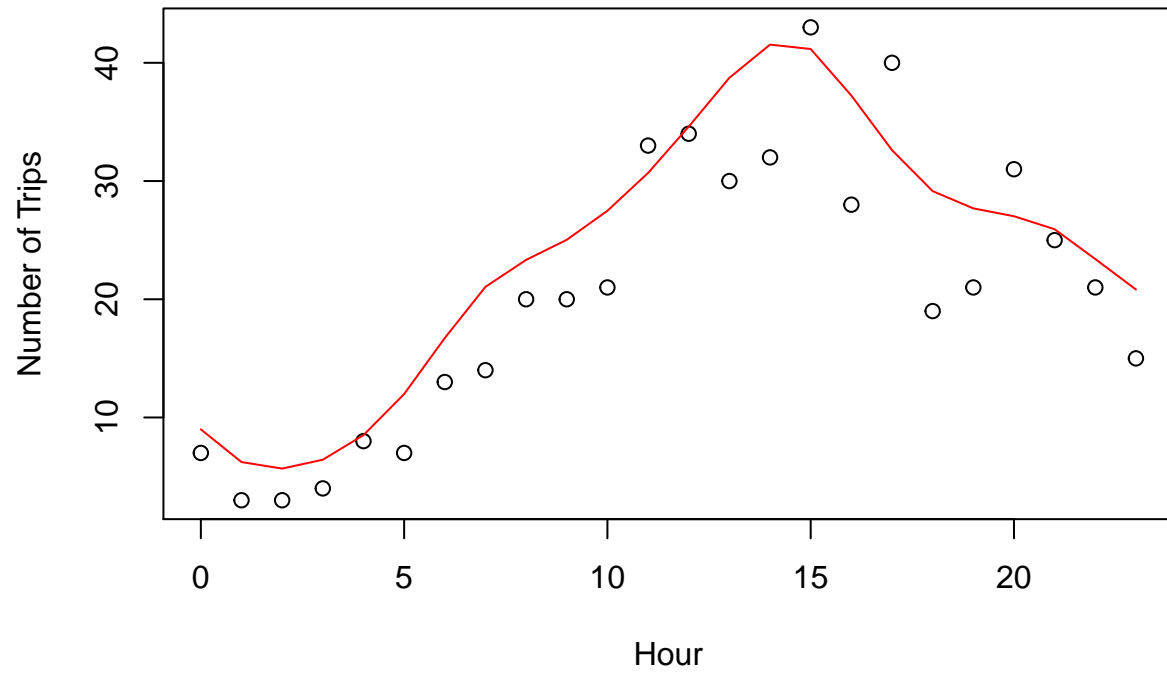
**Number of Hourly Trips (with Predictions) for Location 4**



**Number of Hourly Trips (with Predictions) for Location 5**

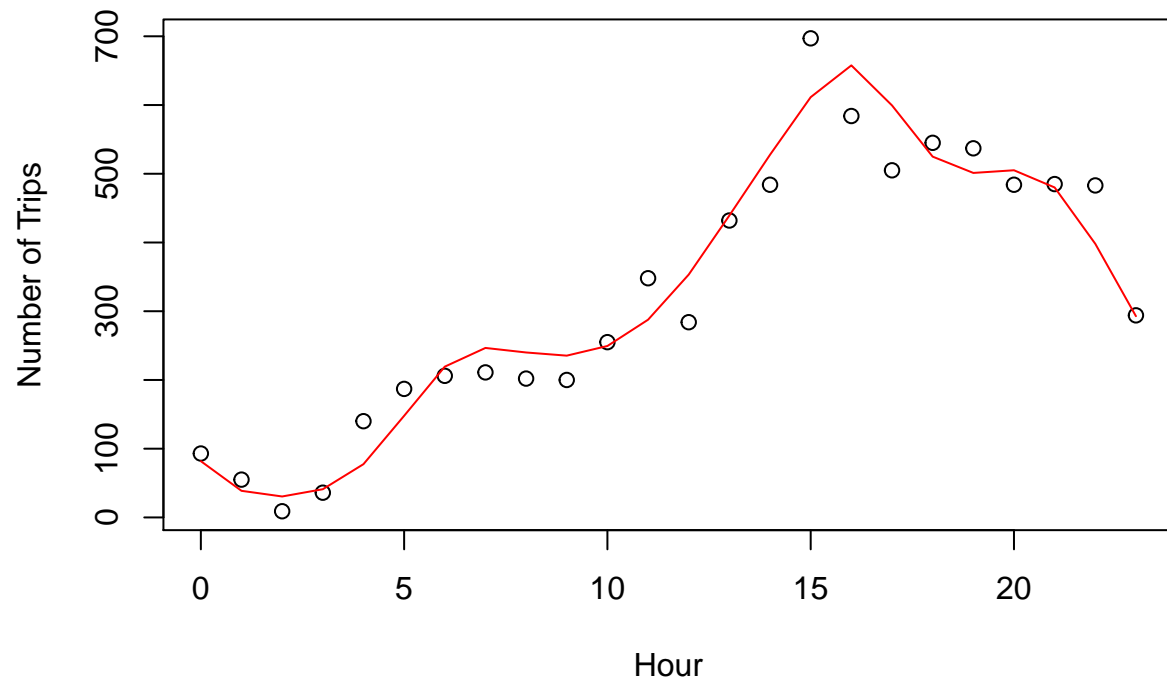


**Number of Hourly Trips (with Predictions) for Location 6**

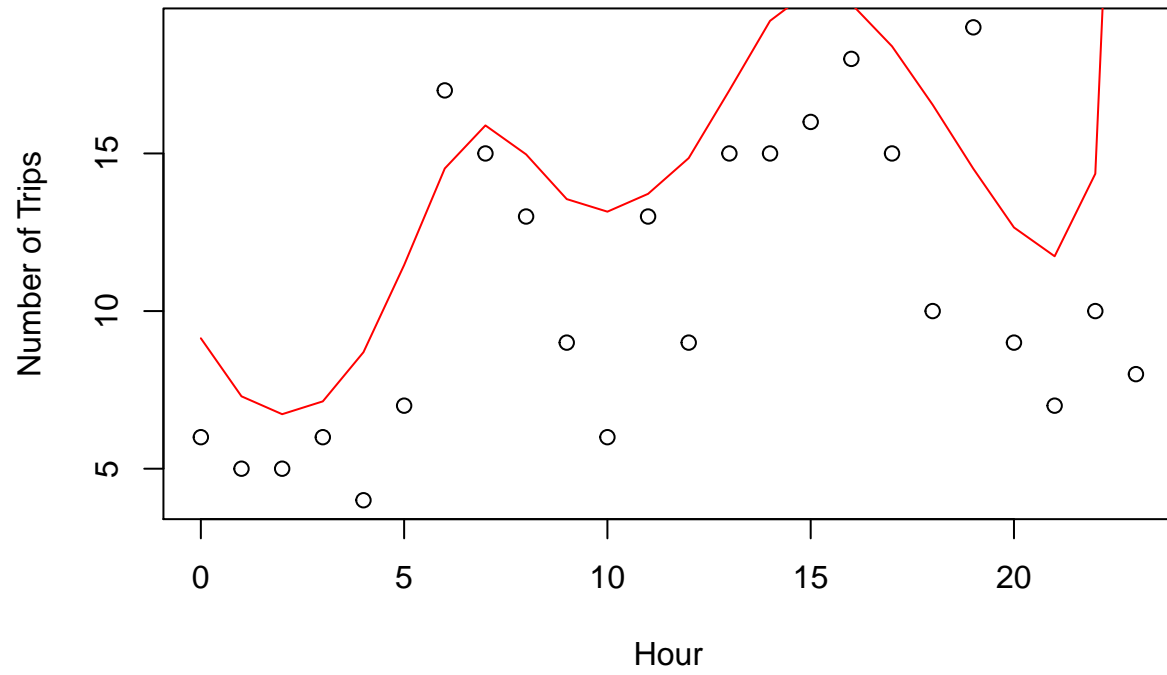




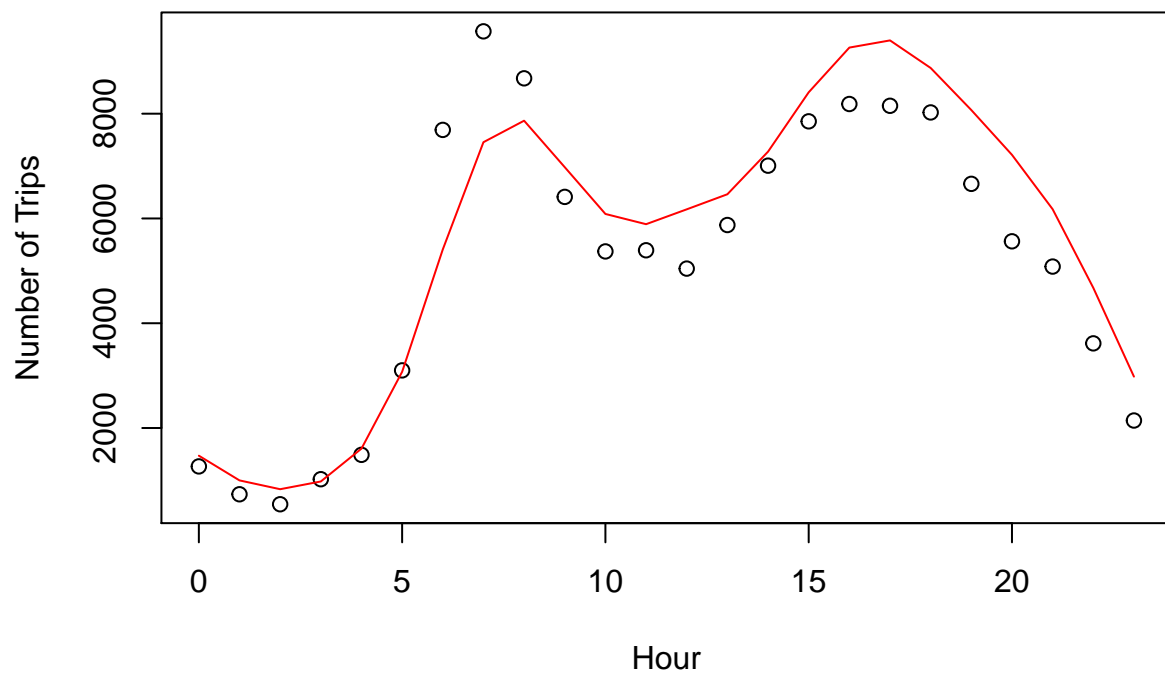
**Number of Hourly Trips (with Predictions) for Location 7**



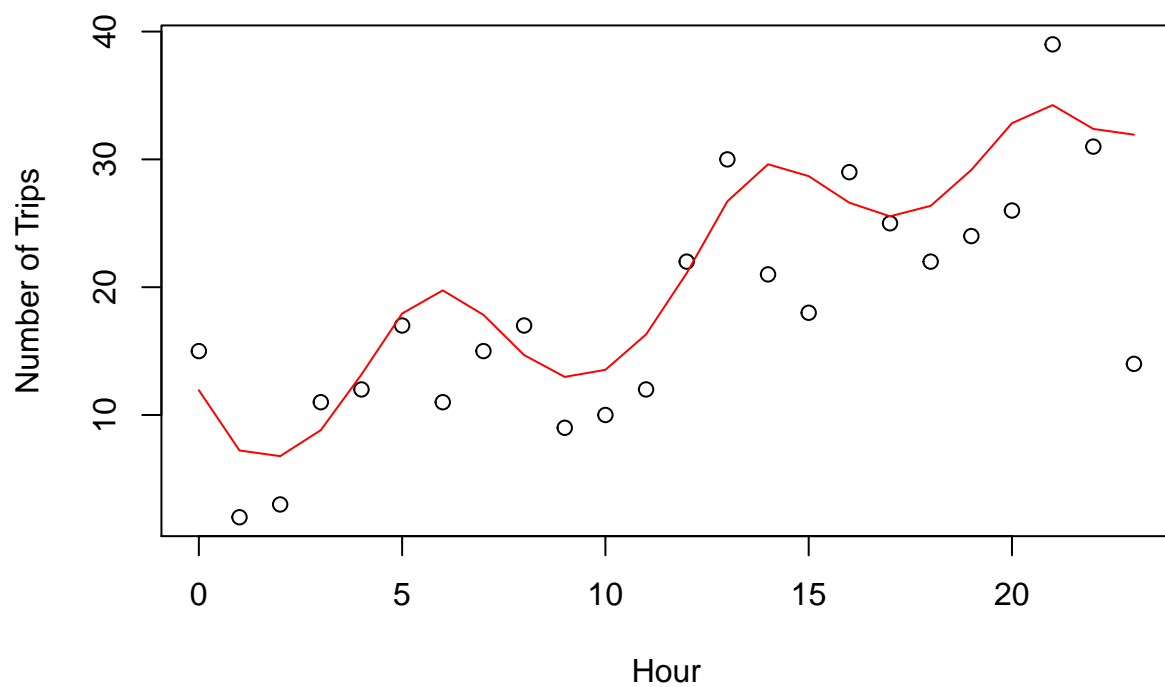
**Number of Hourly Trips (with Predictions) for Location 8**



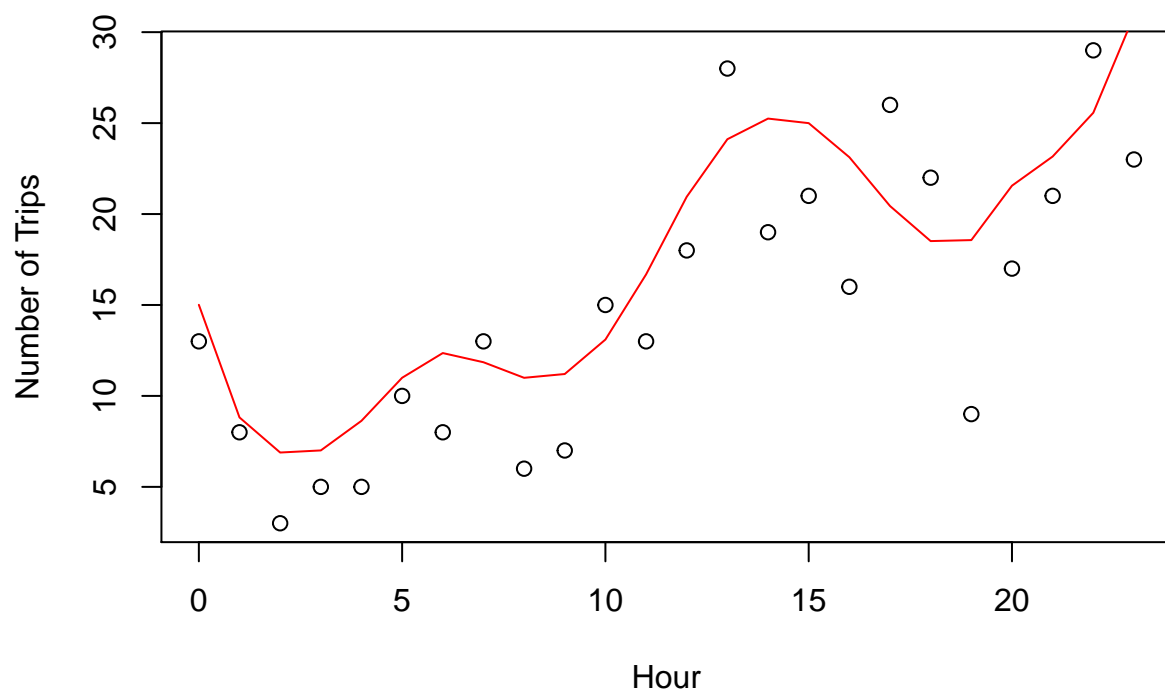
**Number of Hourly Trips (with Predictions) for Location 9**



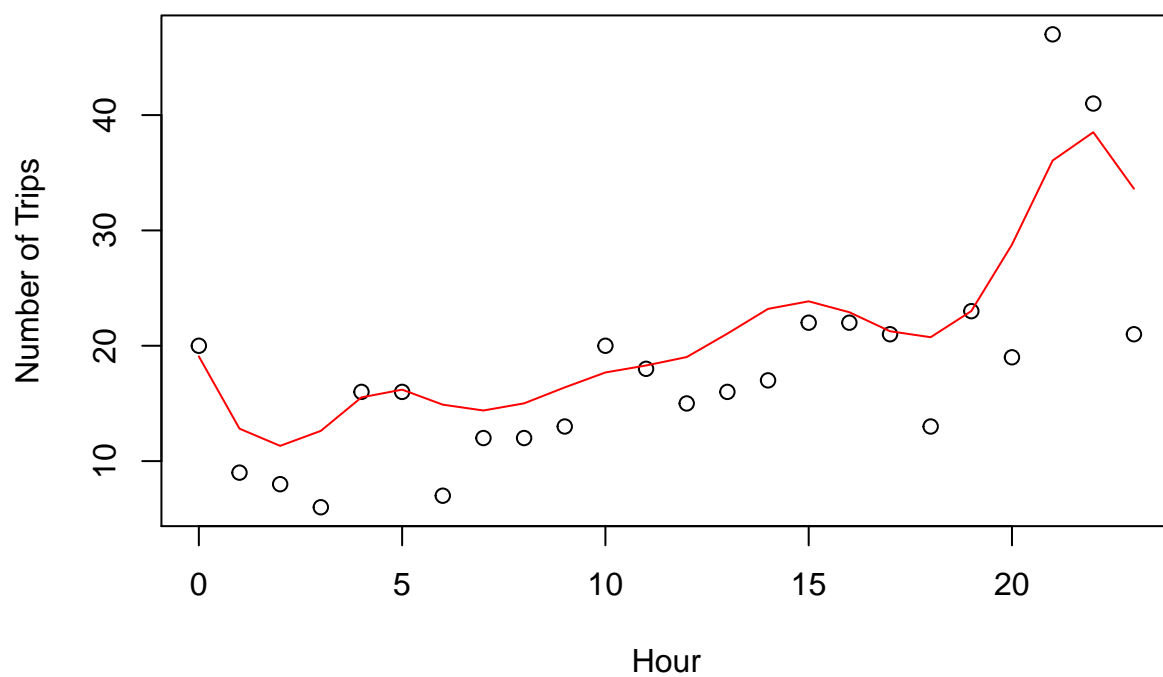
**Number of Hourly Trips (with Predictions) for Location 10**



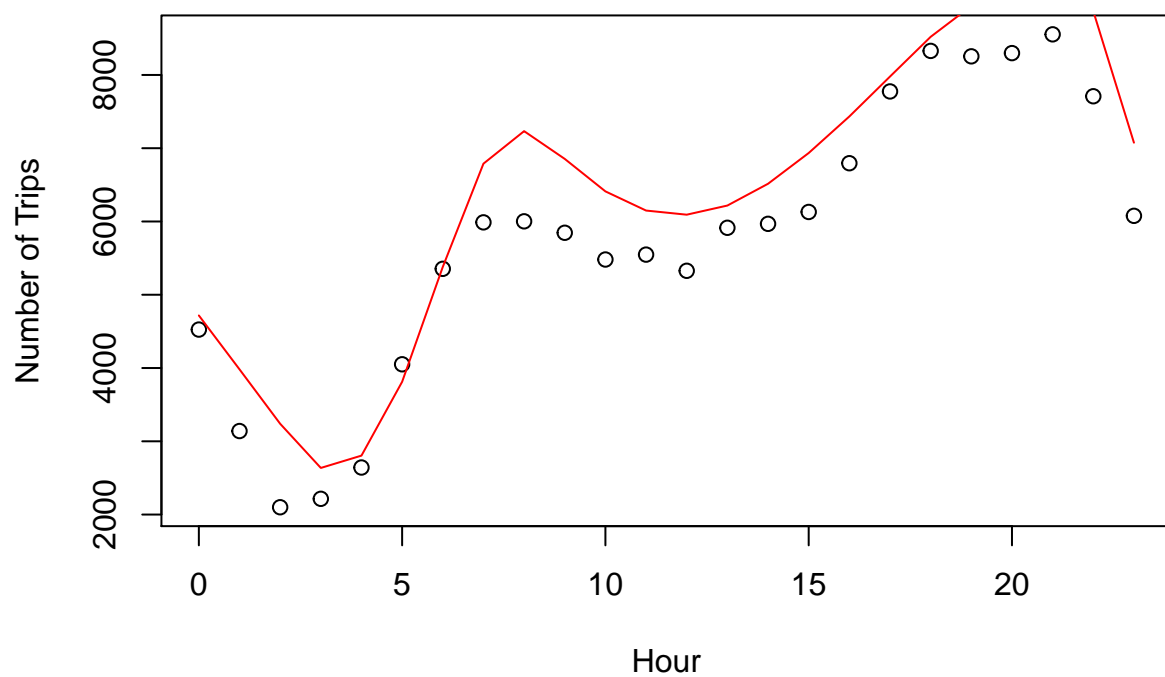
**Number of Hourly Trips (with Predictions) for Location 11**



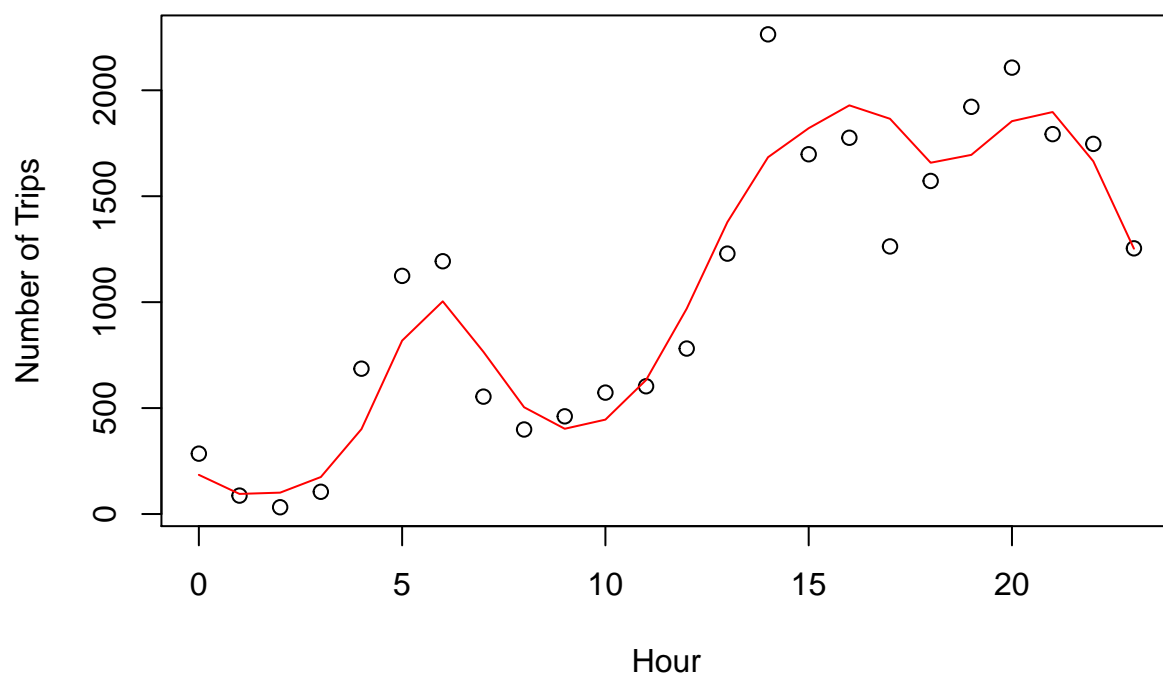
**Number of Hourly Trips (with Predictions) for Location 12**



**Number of Hourly Trips (with Predictions) for Location 13**

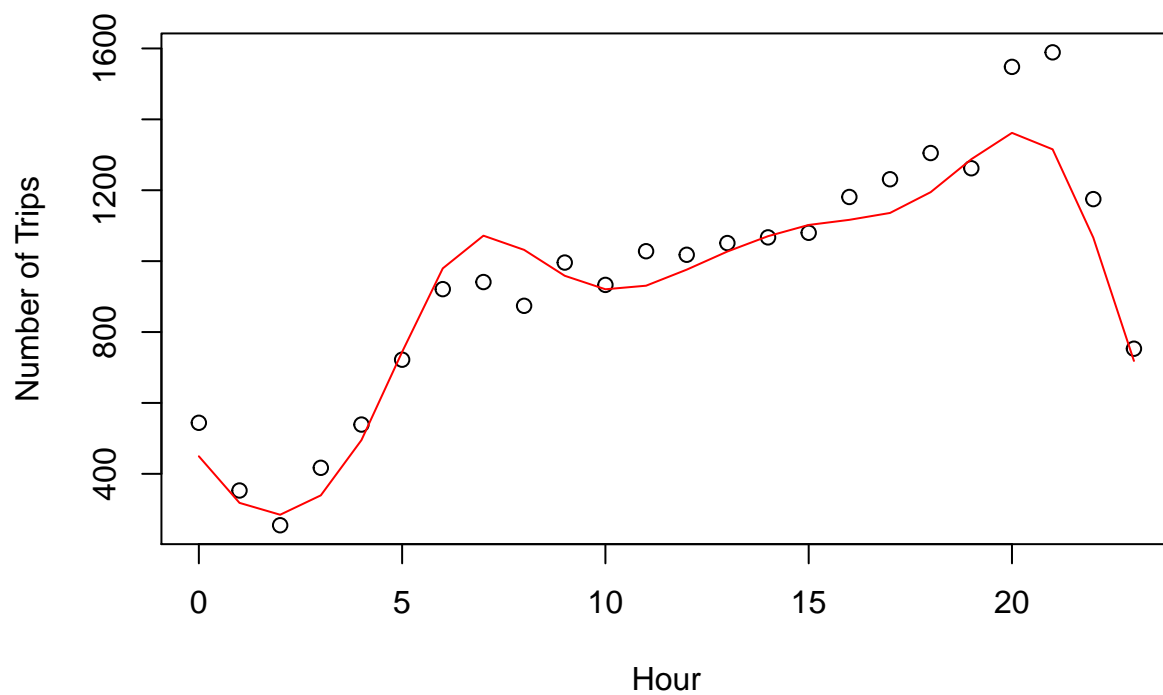


**Number of Hourly Trips (with Predictions) for Location 14**

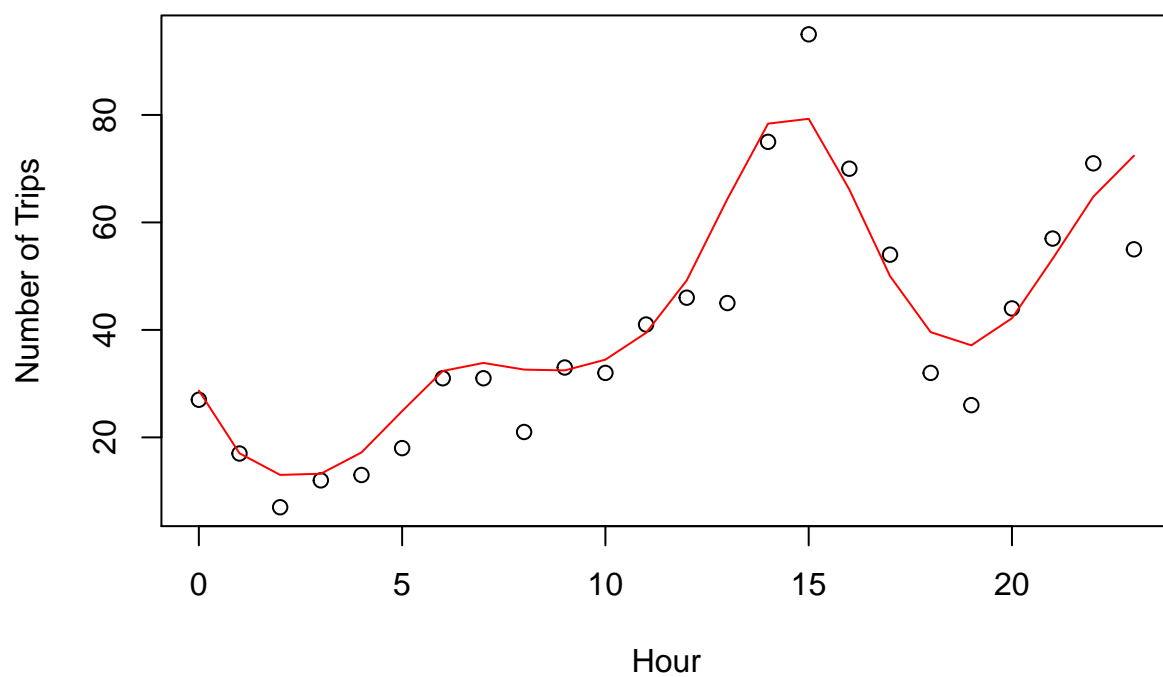




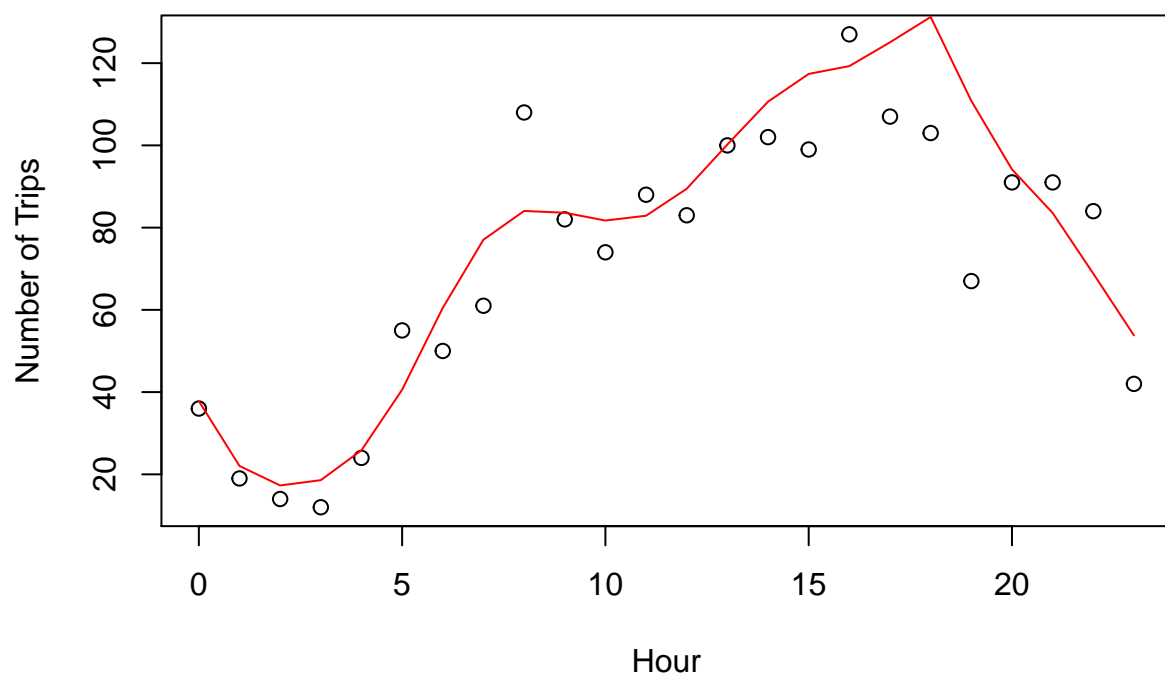
**Number of Hourly Trips (with Predictions) for Location 15**



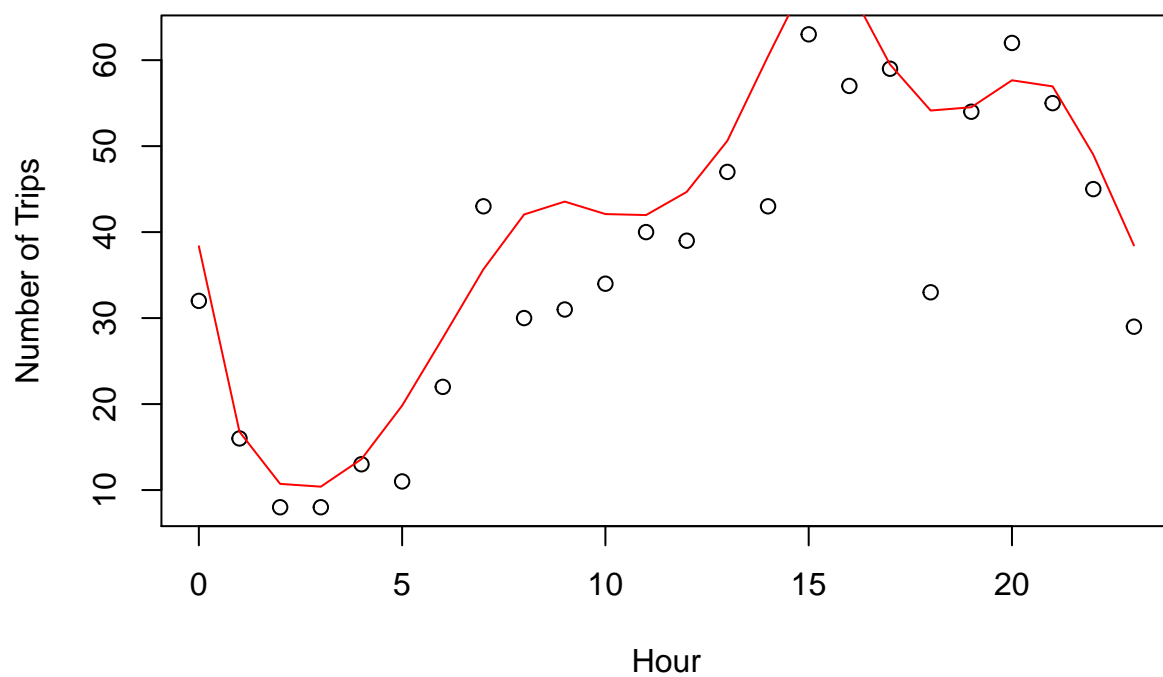
**Number of Hourly Trips (with Predictions) for Location 16**



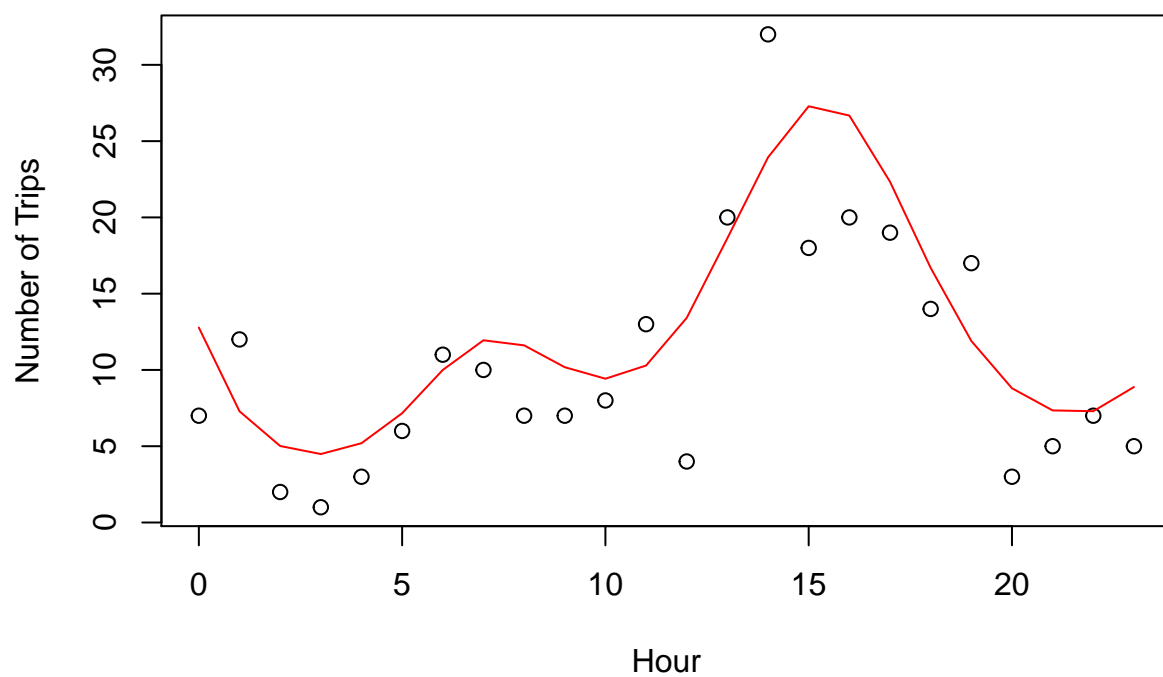
**Number of Hourly Trips (with Predictions) for Location 17**



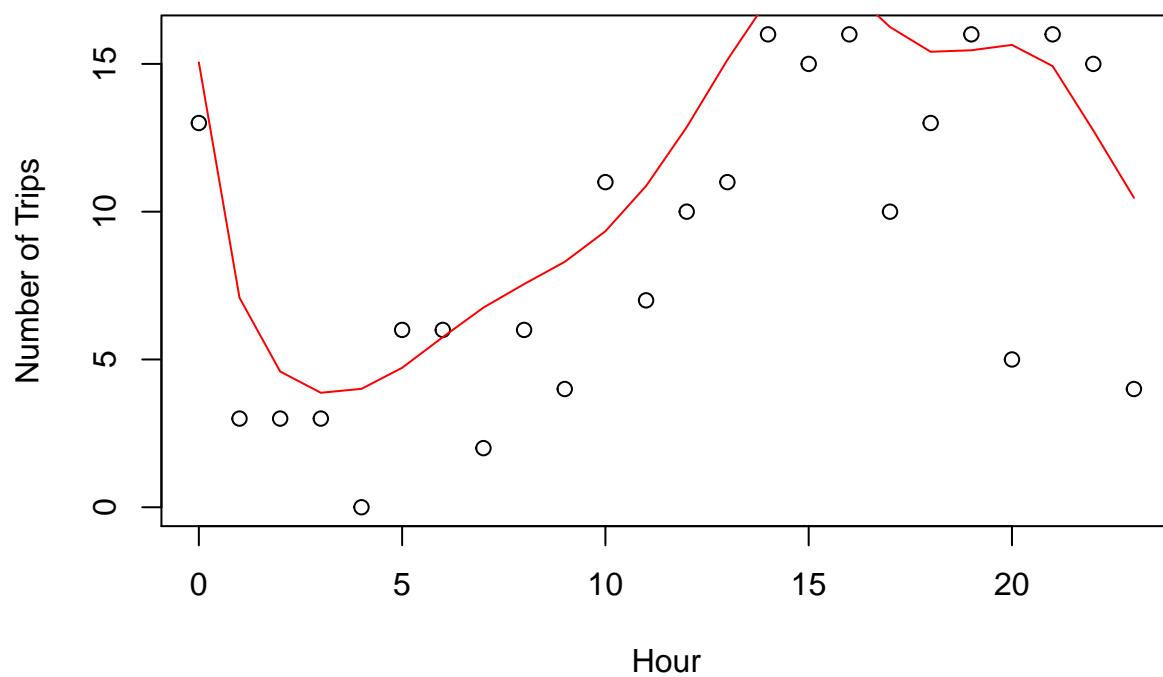
**Number of Hourly Trips (with Predictions) for Location 18**



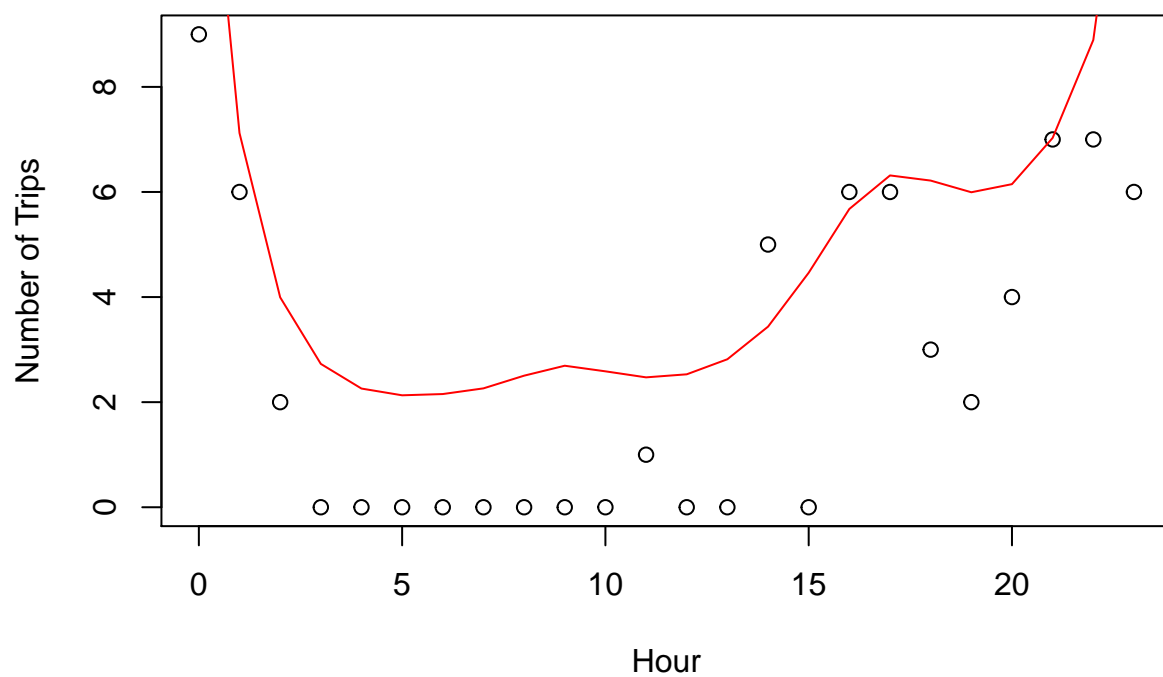
**Number of Hourly Trips (with Predictions) for Location 19**



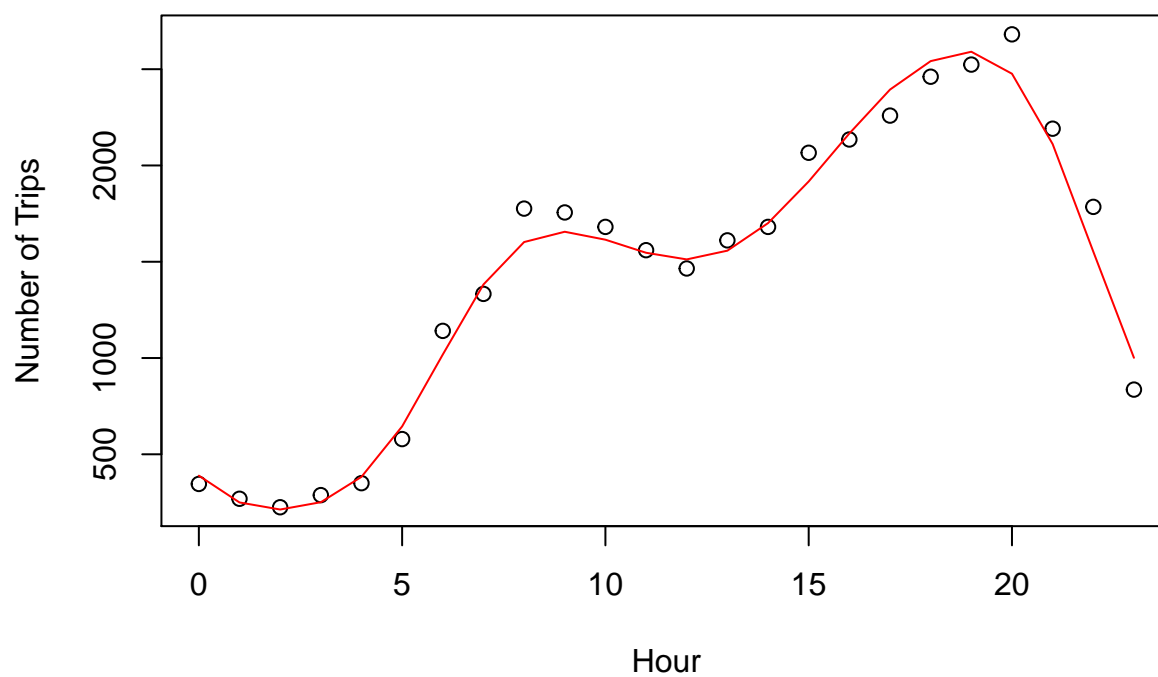
**Number of Hourly Trips (with Predictions) for Location 20**



**Number of Hourly Trips (with Predictions) for Location 21**

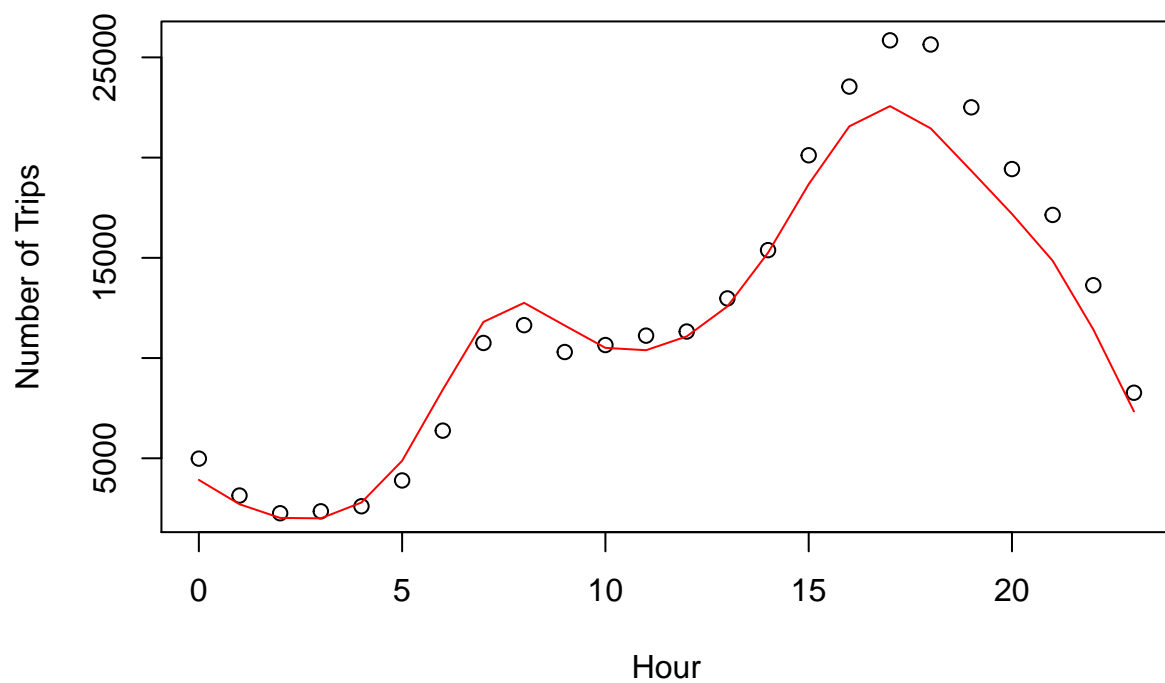


**Number of Hourly Trips (with Predictions) for Location 22**

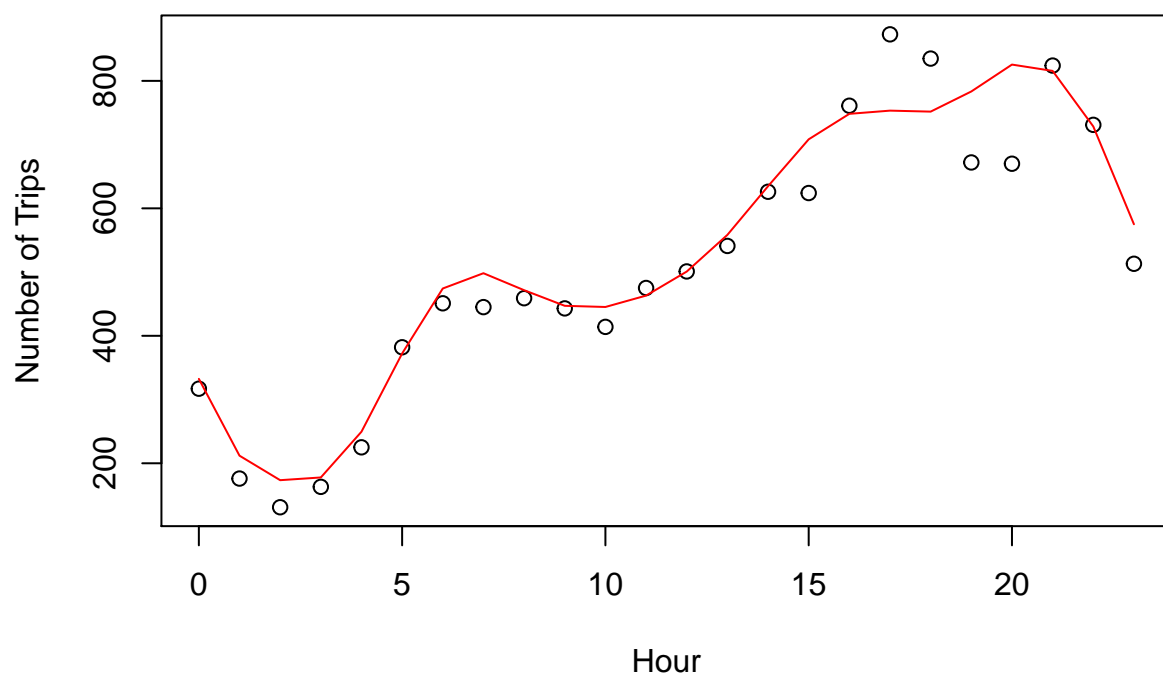




**Number of Hourly Trips (with Predictions) for Location 23**



**Number of Hourly Trips (with Predictions) for Location 24**



**Number of Hourly Trips (with Predictions) for Location 25**

