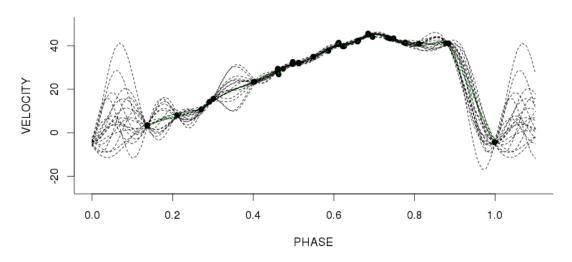
MCMC Midterm 2: Attachments

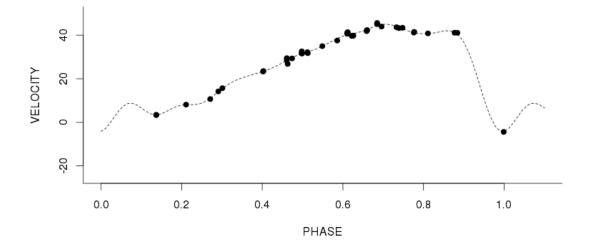
Maurice Diesendruck

April 10, 2015

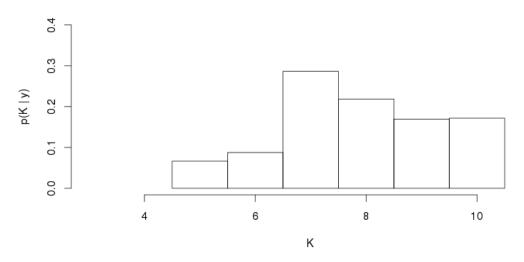
Original Data with Fitted Fourier Regression Estimates



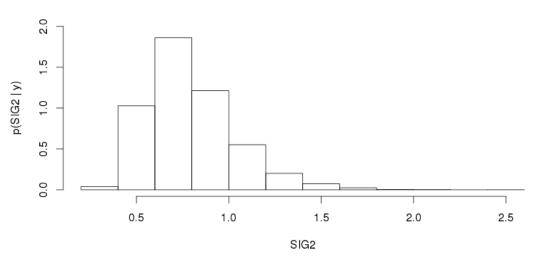
Original Data with E(f|y) Fourier Regression Estimate



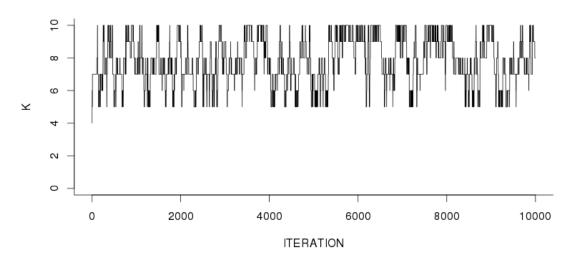
Distribution of K, Given Y

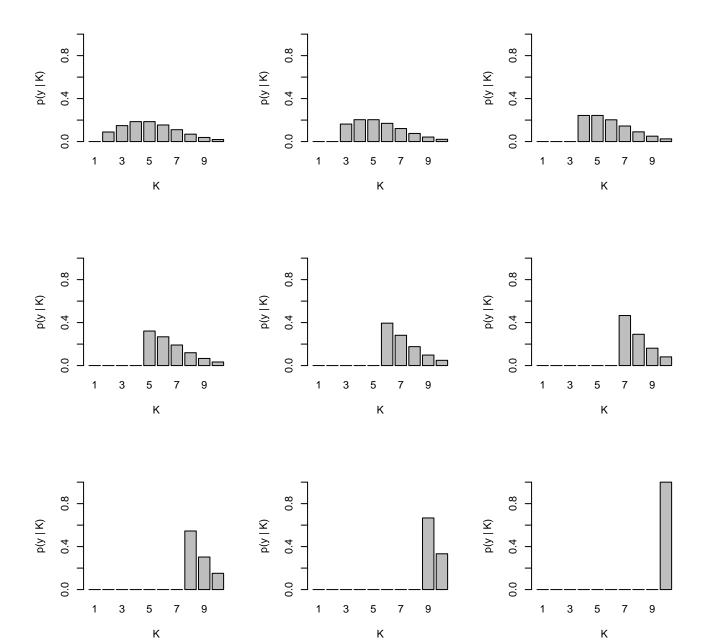


Distribution of σ^2

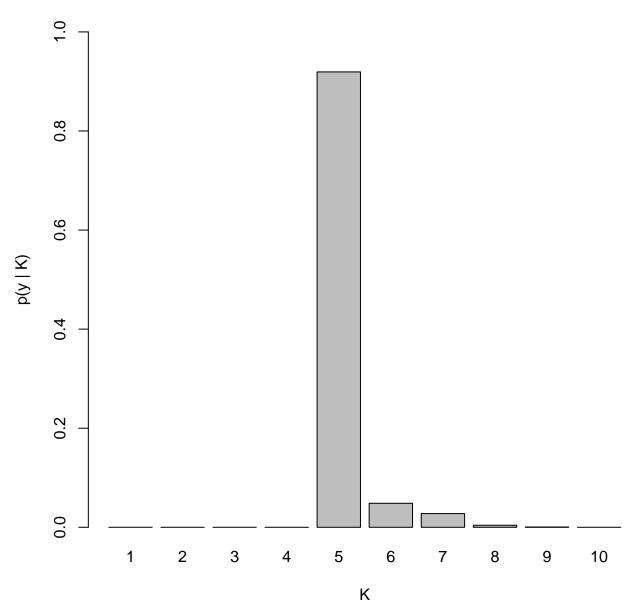


Movement of K Across Iterations









Relevant Code Snippets

0.1 Log Joint Posterior for Ratio

```
ljointpost=function(K, b, sig2){
    #print("calculating posterior with param:");
    n = length(y);
    p <- length(b)
    yhat <- X[,1:p] %*% b</pre>
```

0.2 Gibbs Conditional Posterior Distributions

```
sample.b <- function(K,sig2) { # generate b \sim p(b \mid K, sig2, y)
  idx <- 1:(2*K+1) # select columns (elements) for K harmonics
 Xk <- X[,idx] # Subset of design matrix, with 2K+1 columns.</pre>
 AOk <- AO[idx,idx] # Precision matrix of beta prior.
 b0k <- b0[idx] # Mean vector of zeros from beta prior.
  # Full conditionals from Question 2
 V <- solve(t(Xk)%*%Xk/sig2+A0k)</pre>
 mm <- V%*%(t(Xk)%*%y/sig2)
                              \# LL' = V
 L <- t( chol(V))
  b <- mm + L %*% rnorm(2*K+1) # b ~ N(m, V)
 return (b)
sample.sig2 <- function(K,b) { # generate 1/sig2 \sim p(1/sig2 \mid K,b,y)
 p <- length(b)
 idx <- 1:(2*K+1) # select columns (elements) for K harmonics
 Xk <- X[,idx] # Subset of design matrix, with 2K+1.</pre>
  a1 <- (n/2)+1
 b1 <- t(y-Xk\%*\%b)\%*\%(y-Xk\%*\%b)/2 + 1
  sig2inv <- rgamma(1,shape=a1,rate=b1)</pre>
  sig2 <- 1/sig2inv
  return (sig2)
```

0.3 Auxiliary Variable Transformation and Reverse Transformation

```
qu <- function(K, b, sig2) {
    ## find m,L for mapping T: (b,u) -> b1, below
    ## bnew = m + L*u,
    ## Use a regression of residuals on (K+1)-st harmonic
```

```
## to determine m and L
  idx <- 1:(2*K+1) # select columns (elements) for K harmonics
  Xk <- X[,idx]</pre>
                     # Subset of design matrix, with columns for setting of K.
  eps <- y-Xk%*%b
  regression \leftarrow lm(eps \sim sin((K+1)*2*pi*x) + cos((K+1)*2*pi*x))
  mk <- regression$coefficients[2:3]</pre>
  Vk <- vcov(regression)[2:3,2:3]
  Lk <- t(chol(Vk))</pre>
  return (list(m=mk, V=Vk, L=Lk))
Tinv <- function(K1,b1,sig2) {</pre>
  ## proposed (shorter) par vector
  ## bnew = m + Lu \ or \ u = L^-1 \ (bnew-m)
  K <- K1-1
  p < -2*K+1
  b <- b1[1:p]
  bnew <- b1[c(p+1,p+2)]
  ## back out auxiliary u, and logJ
  fit \leftarrow qu(K, b, sig2)
  u <- solve(fit$L)%*%(bnew-fit$m)
  logJ <- sum(log(diag(fit$L)))</pre>
  return (list(b=b, u=u, logJ=logJ))
```

0.4 Acceptance Probability for Birth Move

```
rho <- function(K, b1, b, u, logJ, sig2) {
    ## acceptance ratio for birth move,
    ## moving from b -> (b,u)
    # current parameter: b
    # proposal: b1
    K1 <- K+1

lqu <- sum(dnorm(u, m=0, sd=1, log=TRUE)) # TODO: Shouldn't this be dnorm?

ljointpostratio <- ljointpost(K1, b1, sig2) - ljointpost(K, b, sig2)

# Priors and likelihood
    rho <- exp(ljointpostratio)

# Transition probabilities</pre>
```

```
if (K==1) {
    rho <- rho*2
} else if (K==Kmx-1) {
    rho <- rho/2
} else {
    # Scaling rule smoothely penalizes larger values of K.
    birth.weight <- qbeta((Kmx-K+0.01)/Kmx, 1, 5)
    rho <- rho*birth.weight
}
rho <- rho/exp(lqu)  # Auxiliary
rho <- rho*exp(logJ)  # Jacobian

return (rho)
}</pre>
```