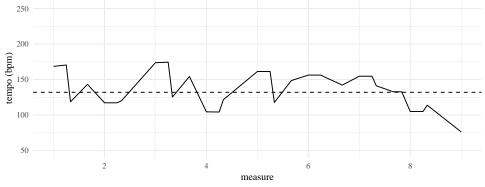
Mazurka paper figures DJM 8/20/2018

Suggested order

- 1. Parameter interpretation in Fliere
- 2. Using parameters to examine two different performances
- 3. Clustering performances (compare the clusters)
 - a. what can we say about the parameters of each cluster? what is different about them?
- 4. Similar performances (Rubinstein)
- 5. Model issues

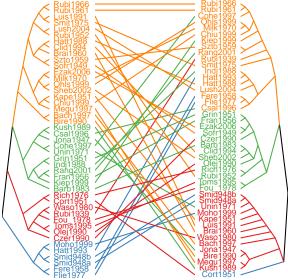
Short tempo

```
ggplot(tempos, aes(x=note_onset, y=Rubinstein_1961)) +
  geom_line() + ylab('tempo (bpm)') + xlab('measure') +
  scale_x_continuous(breaks=1:4*2, limits = c(1,9)) +
  theme_minimal(base_family = 'Times') +
  geom_hline(yintercept = 132, linetype='dashed')
```



Comparing clusters

```
perfs = tempos[,-c(1:3)] %>% as.matrix %>% t
bad_perf = grep('Block',rownames(pvec_ml))
hc_parm = pvec_ml[-bad_perf,] %>% dist %>% percentize %>% hclust
hc_perf = perfs[-bad_perf,] %>% dist %>% percentize %>% hclust
short_labs = rownames(perfs)[-bad_perf]
lens = nchar(short_labs)
short_labs = paste0(substr(short_labs,1,4), substr(short_labs,lens-3,lens))
hc_parm$labels = short_labs
hc_perf$labels = short_labs
dend_parm = hc_parm %>% as.dendrogram
dend_perf = hc_perf %>% as.dendrogram
```



```
heatmap.2(as.matrix(percentize(dist(pvec_ml[-bad_perf,]))),
          Rowv = dend_parm, Colv = dend_parm,
          symm=TRUE,
          density.info = 'none', trace='none',
          labRow = sub('_',' ',row.names(pvec_ml)[-bad_perf]),
          labCol = NA,
          key.title = NA,
          col=colorRampPalette(c('#0b61a4','white')),
          key.xlab = NA,
          margins = c(1,6),
          cexRow = .6,
          cexCol = .6,
          lhei=c(1,8),
          lwid=c(1,8),
          offsetCol = 0, offsetRow = 0,
          key=FALSE
```

```
Rubinstein 1966
Rubinstein 1961
Luisada 1991
Susada 1991
Susada 1991
Rubinstein 1961
Luishak 2004
Rubinstein 1980
Rubinstein 1980
Rubinstein 1980
Sofrontisky 1949
Eraki 2006
Mikina 1989
Eraki 2006
Mikina 1989
Eraki 2006
Kapell 1980
Kapell 1980
Kapell 1980
Rubinstein 1989
Rubinstein 1988
Rangell 2001
Francois 1986
Rubinstein 1989
Rubinstein 1988
Rub
```

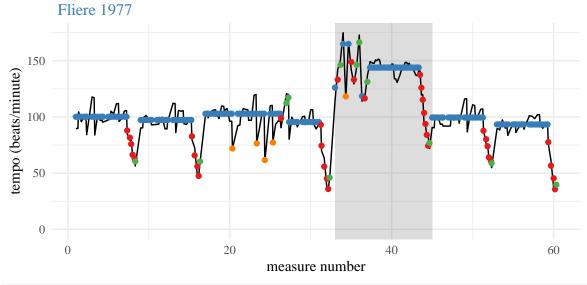
```
clusts = cutree(as.hclust(dend_parm), k = 4)
save(clusts, file = '../extras/ClusterLabels.Rdata')
```

```
heatmap.2(as.matrix(percentize(dist(perfs[-bad_perf,]))),
          Rowv = dend_perf, Colv = dend_perf,
          symm=TRUE,
          density.info = 'none', trace='none',
          labRow = sub('_',' ',row.names(pvec_ml)[-bad_perf]),
          labCol = NA,
         key.title = NA,
          col=colorRampPalette(c('#0b61a4','white')),
          key.xlab = NA,
          margins = c(1,6),
          cexRow = .6,
          cexCol = .6,
         lhei=c(1,8),
          lwid=c(1,8),
          offsetCol = 0, offsetRow = 0,
          key=FALSE
```

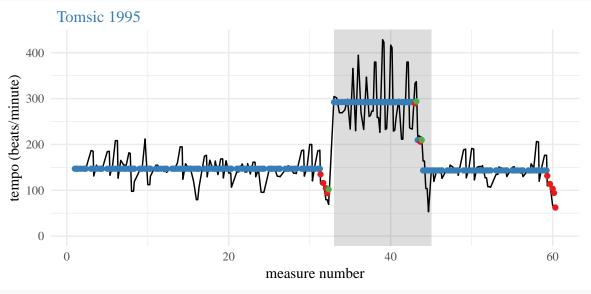


Interpreting parameters

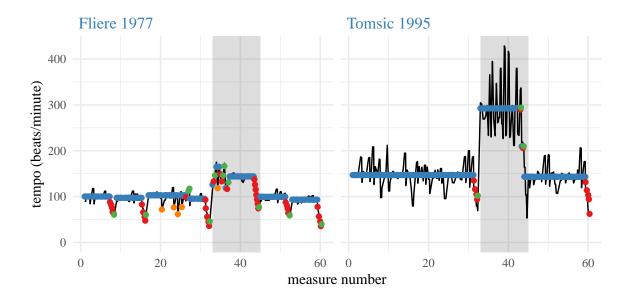
```
plotStates <- function(performers, pars, tempos,</pre>
                       noplot=FALSE,
                       particleNumber = 200,
                       initialMean = c(132,0),
                       initialVariance = c(400,10)){
  lt = diff(c(tempos$note_onset, 61))
  alldfs = NULL
  for(perf in performers){
   params = unlist(pars[row.names(pars)==perf,])
   y = matrix(tempos[[perf]], nrow = 1)
   mats = yupengMats(lt, params[1], params[2:4], params[5:8],
                    params[9:12], initialMean, initialVariance)
   bs = beamSearch(mats\$a0, mats\$P0, c(1,0,0,0,0,0,0,0), mats\$dt,
                    mats$ct, mats$Tt, mats$Zt,
                    mats$Rt, mats$Qt, mats$GGt, y, mats$transMat, particleNumber)
   bestpath = bs$paths[which.max(bs$weights),]
   kal = kalman(mats, bestpath, y)
   df = data.frame(performer=perf, measure = tempos$note_onset, tempo = c(y),
                  inferred = c(kal$ests), state = convert8to4(bestpath))
    alldfs = rbind(alldfs, df)
  if(noplot) return(alldfs)
  ggplot(alldfs, aes(x=measure, y=tempo)) + ylim(0, max(df$tempo)) +
    annotate('rect',xmin = 33, xmax = 45, ymin = -Inf, ymax = Inf,
             alpha=.2) +
   theme_minimal(base_family = 'Times') +
    geom_line(aes(y=tempo), color='black')+
   geom_point(aes(y=inferred),color=fivecolors[alldfs$state])+
   facet_wrap(~performer, labeller = as_labeller(function(x) gsub('_',' ', x))) +
    theme(legend.position = 'none', legend.title = element_blank())+
   ylab('tempo (beats/minute)') + xlab('measure number') +
```



plotStates(perfs[2], pvec_ml, tempos)



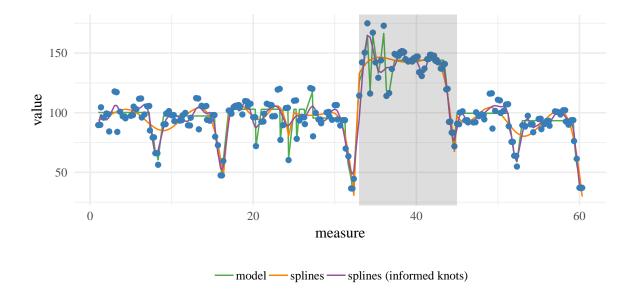
plotStates(perfs, pvec_ml, tempos)



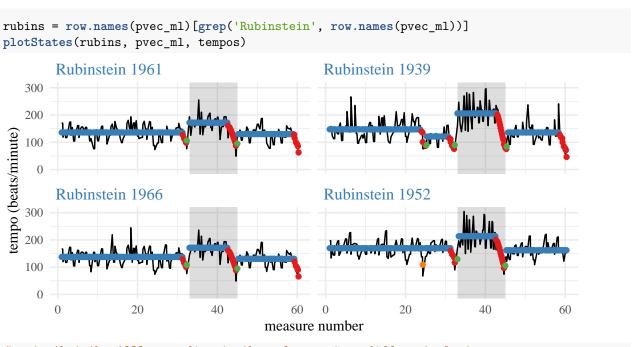
Different smoothing

Try splines, replicating knots, l1tf?

```
nsplines = 64 # 1 knot per bar plus boundary
B = bs(tempos$note_onset, df=nsplines, intercept = TRUE)
preds_smooth = fitted(lm(tempos[[perfs[1]]]~B-1))
single.knots = match(seq(4,56,by=4)+1,tempos$meas_num)
double.knots = match(c(16,24,32,44)+1, tempos\$meas num)
triple.knots = match(c(16,24,32,44)+1, tempos\$meas num)
quad.knots = match(c(16,24,32,44)+1, tempos\$meas_num)
all.knots = tempos$note onset[
  sort(c(single.knots,double.knots,triple.knots,quad.knots))]
B1 = bs(tempos$note_onset, knots = all.knots, intercept = TRUE, Boundary.knots = c(1,61))
preds_music = fitted(lm(tempos[[perfs[1]]]~B1-1))
extras=data.frame(x=tempos$note_onset,y1=preds_smooth,y2=preds_music)
perf1 = plotStates(perfs[1], pvec_ml, tempos, noplot = TRUE)
perf1$ss = preds_smooth
perf1$ms = preds_music
perf1 %>% select(measure, ss, ms, inferred) %>%
  gather(key='key', value='value', -measure) %>%
  ggplot(aes(x=measure)) + geom_line(aes(y=value, color=key)) +
  scale_color_manual(values = fivecolors[c(3,4,5)],
                     labels = c('model','splines','splines (informed knots)')) +
  geom_point(data=perf1, aes(x=measure, y=tempo), color=fivecolors[1]) +
  annotate('rect',xmin = 33, xmax = 45, ymin = -Inf, ymax = Inf,
             alpha=.2) +
  theme_minimal(base_family = 'Times') +
  theme(legend.position = 'bottom',legend.title = element_blank())
```



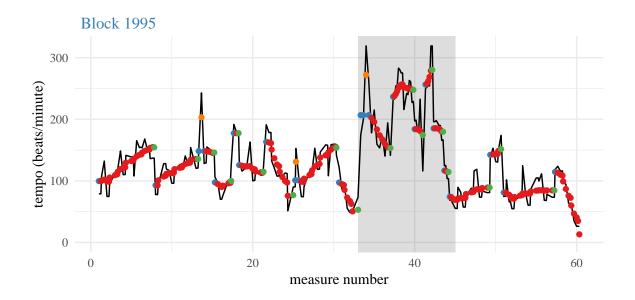
Similar performances



note that the 1939 recording is the only one in a different cluster

Bad estimation

```
plotStates('Block_1995', pvec_ml, tempos)
```



Problems with the model

- Problem with retransitioning to state 1
- states 2 and 3 aren't constrained to always decrease/increase, only in mean
- state 4 may not always emphasize a slow down
- previous 2 have to do with Gaussian assumptions
- necessity for strong priors
- but priors are on parameters, not on path (how would we want this to change?)