

## WebPPL is a feature-rich probabilistic programming language embedded in Javascript.

Check out some **demos** or try it yourself in the editor below.



```

print("=====")
print("PCM20201212_TriangleModePrior&RiskCalculation          *** 2020/12/12 *** ")
print(" see also Simple Reaction Time, Example 9, Card, Moran & Newell, 1983, p.66 ")
print(" see also https://www.humanbenchmark.com/tests/reactiontime/statistics ")
print(" here we use the triangular distribution as a prior distribution ")
print(" see also https://en.wikipedia.org/wiki/Triangular_distribution ")
print(" CMN-interval 'typical[fast ~ slow]' is interpreted ... ")
print("          as triangle(fast=a, slow=b, 'typical'=mode=c) ")
print("=====")
/**
 * @author - Claus Moebus <claus.moebus@uol.de>
 */
//-----
/**
 * @variable {number} startTime - used in method 'runtime' to compute runtime in sec and min
 */
var startTime = Date.now()
//-----
print("Input parameter:")
/**
 * @variable {integer} nTrials - no of efficient samples (incl. burnout) in MCMC-sampling
 */
var nTrials = 6E4
print("nTrials = " + nTrials)
//-----
/**
 * @variable {integer} nSigma - no of standard deviations between mean and 'slow', 'fast'
 *
 *          interval boundaries
 */
var nSigma = 3
print("nSigma = " + nSigma)
//-----
/**
 * @variable {integer} myBurnPeriod - length of burnin period in MCMC process
 */
var myBurnPeriod = nTrials * 0.10
print("length of burn-in period = " + myBurnPeriod)
//-----
/**
 * @variable {integer} myLag - only every myLag-th sample will be retained during MCMC
 */
var myLag = 10
print("length of lag = " + myLag)
//-----
/**
 * @variable {array} data - author's reaction times in an experiment found here
 */

```

```

*           https://www.humanbenchmark.com/tests/reactiontime/
*           visited March 2018
*/
var data =
  [458, 292, 228, 403, 271, 420, 350, 235, 260, 306]
print("response time data = [" + data + "]")
print("mean of data = " + listMean(data))
print("stdev of data = " + listStdev(data))
print("-----")
/**
 * @function seqOfThresholds - generates an array of thresholds between min and max
 * @property {number} min - minimum = fastman's value
 * @property {number} max - maximum = slowman's value
 */
var seqOfThresholds = function(min, max) {
  var range = max - min
  var stepSize = range/50
  var increment = function(x) {x * stepSize + min}
  mapN(increment, Math.floor(range/stepSize + 1))
}
//-----
/**
 * @variable {array} tauPCrit - critical values-at-risk in msec for tauP
 * @variable {array} tauCCrit - critical values-at-risk in msec for tauC
 * @variable {array} tauMCrit - critical values-at-risk in msec for tauM
 * @variable {array} tauSumCrit - critical values-at-risk in msec for tauSum
 */
var tauPCrit = seqOfThresholds(100, 200) // from typical value upto slowmans value
var tauCCrit = seqOfThresholds( 70, 170) // from typical value upto slowmans value
var tauMCrit = seqOfThresholds( 70, 100) // from typical value upto slowmans value
var tauSumCrit = seqOfThresholds(240, 470) // from typical value upto slowmans value
print("-----")
/**
 * @object hyperParmTauX - parameters 'typical'=c, fast=a, and slow=b for prior Triangle
 * - 'typical', fast, and slow are taken from MHP
 * - 'typical' = mode = c
 * @property {number} 'typical' - value is the 'typical' value of the CMN-interval
 * @property {number} a - value is the 'fast' parameter of Triangle(a, b, c)
 * @property {number} b - value is the 'slow' parameter of Triangle(a, b, c)
 * @property {number} c - value is the mode or c parameter of Triangle(a, b, c)
 */
var hyperParmTauP = {c:100.0, a:50.0, b:200.0}
print("hyperParmTauP = {c:" + hyperParmTauP.c + ", a:" + hyperParmTauP.a +
      ", b:" + hyperParmTauP.b + "}")
var hyperParmTauC = {c: 70.0, a:25.0, b:170.0}
print("hyperParmTauC = {c: " + hyperParmTauC.c + ", a:" + hyperParmTauC.a +
      ", b:" + hyperParmTauC.b + "}")
var hyperParmTauM = {c: 70.0, a:30.0, b:100.0}
print("hyperParmTauM = {c: " + hyperParmTauM.c + ", a:" + hyperParmTauM.a +
      ", b:" + hyperParmTauM.b + "}")
print("-----")
/**
 * @object hyperParmSigmaTauSum - shape=a and scale=b for variance of Gaussian Likelihood
 * @property {number} a - value is the shape parameter of Gamma(a, b)
 * @property {number} b - value is the scale parameter of Gamma(a, b)
 */
var hyperParmSigmaTauSum = {a:4.0, b:20.0}
print("hyperParmSigmaTauSum = {a:" + hyperParmSigmaTauSum.a + ", b:" + hyperParmSigmaTauSum.b + "}")
print("-----")
//-----
// function definitions

```

```

//-----
/**
 * @function runtime - method to compute the runtime in seconds and minutes
 */
var runTime = function() {
  var stopTime = Date.now()
  var runSecs = (stopTime - startTime)/1000
  var runMins = runSecs/60
  print("runtime in seconds = " + runSecs)
  print("runtime in minutes = " + runMins)}
//-----
/**
 * @description - descriptive statistics of a sample-generated distribution
 * @function myTauXDistribution
 * @param {string} id - The identifier of the tauX distribution.
 * @param {distributionObject} tauXDistribution - tauX distribution (X = P, C, M, T)
 * @param {number} modeTauX - mode of tauX as a function of a and b
 *                               mode = (a-1)*b for a >= 1
 * @returns {object} meanSigmaTauObject - object with mean and sigma of TauX
 * @property {number} meanTauX - mean of tauX (X = P, C, M, T) or tau
 * @property {number} sigmaTauX - standard deviation of tauX (X = P, C, M, T) or tau
 */
var myTauXDescription = function(id, tauXDistribution, modeTauX) {
  var myTauXDistribution = { // extraction of probs and support from WebPPL tauX distribution
    probs: map(function(eventTuple){ // object to compute mean and sigma of tauX
      Math.exp(tauXDistribution.score(eventTuple))}, tauXDistribution.support()),
    support: tauXDistribution.support()}
  print(id)
  // mode(tauX), mean(tauX), variance(tauX) and sigma(tauX)
  print("mode = " + modeTauX)
  var meanTauX = sum(map2(function(value, prob) {
    value*prob},myTauXDistribution.support, myTauXDistribution.probs))
  print("mean = " + meanTauX)
  var sigmaTauX = Math.sqrt(sum(map2(function(value, prob) {
    Math.pow((value-meanTauX), 2)*prob},
    myTauXDistribution.support,
    myTauXDistribution.probs)))
  print("sigma = " + sigmaTauX)
  var tauX_Intval = {fast:meanTauX - nSigma * sigmaTauX, mean:meanTauX,
    slow:meanTauX + nSigma * sigmaTauX}
  return tauX_Intval}
//-----
/**
 * @description - cdf computes the cumulative density function P(X <= c)
 * @function cdf
 * @param {distributionObject} distrObject - must be generated by function 'Infer'
 * @param {real} c - function argument of cdf F(c) = P(X <= c)
 * @returns {real} - F(c) = P(X <= c)
 */
var cdf = function(distrObject, c) {
  var support = distrObject.support()
  var probs = map(function(xValue){
    Math.exp(distrObject.score(xValue))
  }, support)
  sum(map2(function(prob, xValue) {
    xValue <= c ? prob : 0
  }, probs, support))
}
//-----
/**
 * @description - probsAtRisk computes the cumulative density function 1-F(c) = P(X > c)

```

```

* @function probsAtRisk
* @param {distributionObject} distrObject - must be generated by function 'Infer'
* @param {real} valsAtRisk - function arguments of cdf  $1-F(c) = P(X > c)$ 
* @returns {array} -  $F(c_i) = P(X \leq c_i)$  ;  $i = 1, \dots$ 
*/
var probsAtRisk = function (distrObject, valsAtRisk) {
  map(function(valAtRisk) {
    1.0 - cdf(distrObject, valAtRisk)
  }, valsAtRisk)
}
//-----
/**
* @description - prints a table of two column vectors:
*               - values-at-risk and risk probabilities
* @function printRiskProbs
*/
var printRiskProbs = function(valsAtRisk, valsAtRiskText, probs) {
  /*
  map2(function(valAtRisk, prob) {
    print(valsAtRiskText + " = " + valAtRisk + "; risk probability = " + prob)
  }, valsAtRisk, probs)
  */
}
//-----
/**
* @description - prints a table of two column vectors:
*               - values-at-risk and increase of risk probabilities
* @function printDiffProbs
*/
var displayDiffProbs = function(valsAtRisk, valsAtRiskText, probsPrior, probsPosterior) {
  var probDiffs = map2(function(priorPr, postPr) {
    postPr - priorPr // change
  }, probsPrior, probsPosterior)
  map2(function(valAtRisk, probDiff) {
    if (probDiff < 0.05) {print(valsAtRiskText + " = " + valAtRisk
      + "; increase in risk probs = " + probDiff)}
    else {/* empty */ ;}}
    , valsAtRisk, probDiffs)
  viz.line(valsAtRisk, probDiffs, {xLabel: valsAtRiskText, yLabel: "Risk Excess"})
}
//=====
/**
* @description      - draws one sample from the Triangle(a, b, c)-distribution
*                   - // https://en.wikipedia.org/wiki/Triangular\_distribution
* @function          - oneSampleOfTriangle
* @param (number) fast - is the lower bound of the CMN-interval and a of Triangle(a, b, c)
* @param (number) slow - is the upper bound of the CMN-interval and b of Triangle(a, b, c)
* @param (number) mode - is the mode of the CMN-interval and param c of Triangle(a, b, c)
*/
var oneSampleOfTriangle = function(a, b, c) {
  var u = sample(Uniform({a:0, b:1}))
  var ba = b - a
  var bc = b - c
  var ca = c - a
  var Fc = ca / ba
  var x = (0 < u) && (u < Fc) ?
    (a + Math.sqrt(u * ba * ca)) :
    (b - Math.sqrt((1 - u) * ba * bc))
  return x
}
//-----

```

```

/**
 * @function oneSampleOfPriors      - takes o n e sample from all priors tauP, tauC, tauM,
 *                                  - tauSum = tauP + tauC + tauM, and sigmaTauSum
 * @returns {object} sampleOfPriors - o n e priors-tuple
 * @returns {object} priorSigmaTauSum - o n e sample from the Gamma distr
 *                                  - this is prior sigma for the Gaussian likelihood
 */
var oneSampleOfPriors = function () {
  var priorTauP = oneSampleOfTriangle(hyperParmTauP.a,hyperParmTauP.b,hyperParmTauP.c)
  var priorTauC = oneSampleOfTriangle(hyperParmTauC.a,hyperParmTauC.b,hyperParmTauC.c)
  var priorTauM = oneSampleOfTriangle(hyperParmTauM.a,hyperParmTauM.b,hyperParmTauM.c)
  var priorTauSum = priorTauP + priorTauC + priorTauM
  var priorSigmaTauSum =
    sample(Gamma({shape:hyperParmSigmaTauSum.a, scale:hyperParmSigmaTauSum.b}))
  return {priorTauP:priorTauP, priorTauC:priorTauC, priorTauM:priorTauM,
    priorTauSum:priorTauSum, priorSigmaTauSum:priorSigmaTauSum}
}
//-----
/**
 * @description - Infer generates an univariate prior Gamma distribution for TauSum
 * @variable {distribution} priorTauSum - value is a WebPPL distribution object
 */
var priorTauX = Infer({model:oneSampleOfPriors, method: 'forward', samples: nTrials})
print('Univariate Priors TauX (X=P, C, M, Sum, SigmaTauSum) ~ Gamma(???, ???)')
viz.marginals(priorTauX)
print("-----")
print("model-generated "+ nSigma + "*sigma tau-interval: ")
var priorTauPIntval =
  myTauXDescription("priorTauP", marginalize(priorTauX,'priorTauP'), "unknown")
print("{fast:" + priorTauPIntval.fast + " mean:" + priorTauPIntval.mean + " slow:" + prior
var tauPProbsPrior = probsAtRisk(marginalize(priorTauX,'priorTauP'), tauPCrit)
printRiskProbs(tauPCrit, 'tauPCrit', tauPProbsPrior)
print("-----")
var priorTauCIntval =
  myTauXDescription("priorTauC", marginalize(priorTauX,'priorTauC'), "unknown")
print("{fast:" + priorTauCIntval.fast + " mean:" + priorTauCIntval.mean + " slow:" + prior
var tauCProbsPrior = probsAtRisk(marginalize(priorTauX,'priorTauC'), tauCCrit)
printRiskProbs(tauCCrit, 'tauCCrit', tauCProbsPrior)
print("-----")
var priorTauMIntval =
  myTauXDescription("priorTauM", marginalize(priorTauX,'priorTauM'), "unknown")
print("{fast:" + priorTauMIntval.fast + " mean:" + priorTauMIntval.mean + " slow:" + prior
var tauMProbsPrior = probsAtRisk(marginalize(priorTauX,'priorTauM'), tauMCrit)
printRiskProbs(tauMCrit, 'tauMCrit', tauMProbsPrior)
print("-----")
var priorTauSumIntval =
  myTauXDescription("priorTauSum", marginalize(priorTauX,'priorTauSum'), "unknown")
print("{fast:" + priorTauSumIntval.fast + " mean:" + priorTauSumIntval.mean + " slow:" + pri
var tauSumProbsPrior = probsAtRisk(marginalize(priorTauX,'priorTauSum'), tauSumCrit)
printRiskProbs(tauSumCrit, 'tauSumCrit', tauSumProbsPrior)
print("-----")
var priorSigmaTauSum_Intval =
  myTauXDescription("priorSigmaTauSum", marginalize(priorTauX,'priorSigmaTauSum'), "unknown")
print("model-generated "+ nSigma + "*sigma tau-interval: ")
print("{fast:" + priorSigmaTauSum_Intval.fast + " mean:" + priorSigmaTauSum_Intval.mean + "
print("=====")
/**
 * @function oneSampleOfModel - takes o n e sample from the priors
 * @returns {object} posteriorTauSum - returns o n e sample of posterior TauSum-tuple
 */
var oneSampleOfModel = function() {

```

```

/**
 * @variable {number} PriorTauSum - a sample from Gamma TauSum-distribution
 */
var priorTauP = oneSampleOfTriangle(hyperParmTauP.a,hyperParmTauP.b,hyperParmTauP.c)
var priorTauC = oneSampleOfTriangle(hyperParmTauC.a,hyperParmTauC.b,hyperParmTauC.c)
var priorTauM = oneSampleOfTriangle(hyperParmTauM.a,hyperParmTauM.b,hyperParmTauM.c)
var priorTauSum = priorTauP + priorTauC + priorTauM
/**
 * @variable {number} priorSigmaTauSum - a sample from SigmaTauSum Gamma distribution
 */
var priorSigmaTauSum =
  sample(Gamma({shape:hyperParmSigmaTauSum.a, scale:hyperParmSigmaTauSum.b}))
//
map(function(datum) {
  observe(Gaussian({mu:priorTauSum, sigma:priorSigmaTauSum}),datum)
}, data)
return {postTauP: priorTauP, postTauC: priorTauC, postTauM: priorTauM,
        postTauSum:priorTauSum, postSigmaTauSum:priorSigmaTauSum}
}
//-----
/**
 * @description - Infer generates the posterior distribution 'posteriorTauT'
 * @variable {distributionObject} posteriorTauT - multivariate posterior distribution
 */
print('Univariate Posteriors TauX (X=P, C, M, Sum) Gamma(???, ???) and SigmaTauSum Gamma(???, ???)')
var posterior = Infer({model:oneSampleOfModel, method:'MCMC', samples: nTrials,
                      burn:myBurnPeriod, lag:myLag})
viz.marginals(posterior)
print("-----")
print("model-generated " + nSigma + "*sigma tau-interval: ")
var postTauPIntval =
  myTauXDescription("postTauP", marginalize(posterior,'postTauP'), "unknown")
print("{fast:" + postTauPIntval.fast + " mean:" + postTauPIntval.mean + " slow:" + postTauPIntval.slow + "}")
var tauPProbsPosterior = probsAtRisk(marginalize(posterior,'postTauP'), tauPCrit)
printRiskProbs(tauPCrit, 'tauPCrit', tauPProbsPosterior)
print("-----")
displayDiffProbs(tauPCrit, 'tauPCrit', tauPProbsPrior, tauPProbsPosterior)
print("-----")
var postTauCIntval =
  myTauXDescription("postTauC", marginalize(posterior,'postTauC'), "unknown")
print("{fast:" + postTauCIntval.fast + " mean:" + postTauCIntval.mean + " slow:" + postTauCIntval.slow + "}")
var tauCProbsPosterior = probsAtRisk(marginalize(posterior,'postTauC'), tauCCrit)
printRiskProbs(tauCCrit, 'tauCCrit', tauCProbsPosterior)
print("-----")
displayDiffProbs(tauCCrit, 'tauCCrit', tauCProbsPrior, tauCProbsPosterior)
print("-----")
var postTauMIntval =
  myTauXDescription("postTauM", marginalize(posterior,'postTauM'), "unknown")
print("{fast:" + postTauMIntval.fast + " mean:" + postTauMIntval.mean + " slow:" + postTauMIntval.slow + "}")
var tauMProbsPosterior = probsAtRisk(marginalize(posterior,'postTauM'), tauMCrit)
printRiskProbs(tauMCrit, 'tauMCrit', tauMProbsPosterior)
print("-----")
displayDiffProbs(tauMCrit, 'tauMCrit', tauMProbsPrior, tauMProbsPosterior)
print("-----")
var postTauSumIntval =
  myTauXDescription("postTauSum", marginalize(posterior,'postTauSum'), "unknown")
print("{fast:" + postTauSumIntval.fast + " mean:" + postTauSumIntval.mean + " slow:" + postTauSumIntval.slow + "}")
var tauSumProbsPosterior = probsAtRisk(marginalize(posterior,'postTauSum'), tauSumCrit)
printRiskProbs(tauSumCrit, 'tauSumCrit', tauSumProbsPosterior)
print("-----")
displayDiffProbs(tauSumCrit, 'tauSumCrit', tauSumProbsPrior, tauSumProbsPosterior)

```

```

print("-----")
var postSigmaTauSumIntval =
  myTauXDescription("postSigmaTauSum", marginalize(posterior, 'postSigmaTauSum'), "unknown")
print("{fast:" + postSigmaTauSumIntval.fast + " mean:" + postSigmaTauSumIntval.mean + " slow:" + postSigmaTauSumIntval.slow}")
print("=====")
runTime()
print("=====")

```

run

```

=====
PCM20201212_TriangleModePrior&RiskCalculation          *** 2020/12/12 ***
  see also Simple Reaction Time, Example 9, Card, Moran & Newell, 1983, p.66
  see also https://www.humanbenchmark.com/tests/reactiontime/statistics
  here we use the triangular distribution as a prior distribution
  see also https://en.wikipedia.org/wiki/Triangular_distribution
  CMN-interval 'typical[fast ~ slow]' is interpreted ...
      as triangle(fast=a, slow=b, 'typical'=mode=c)
=====

Input parameter:
nTrials = 60000
nSigma = 3
length of burn-in period = 6000
length of lag = 10
response time data = [458,292,228,403,271,420,350,235,260,306]
mean of data = 322.3
stdev of data = 77.10389095240265

-----

hyperParmTauP = {c:100, a:50, b:200}
hyperParmTauC = {c: 70, a:25, b:170}
hyperParmTauM = {c: 70, a:30, b:100}

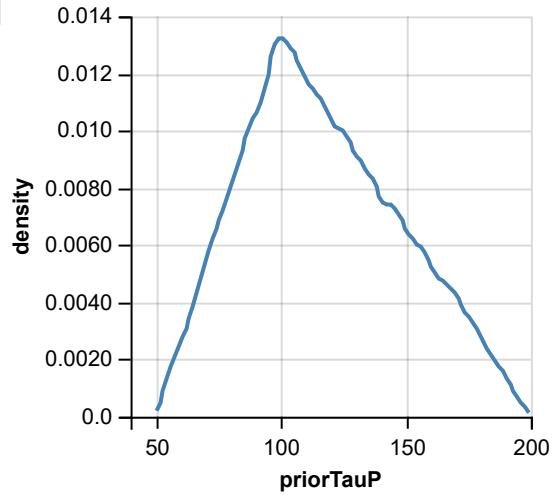
-----

hyperParmSigmaTauSum = {a:4, b:20}

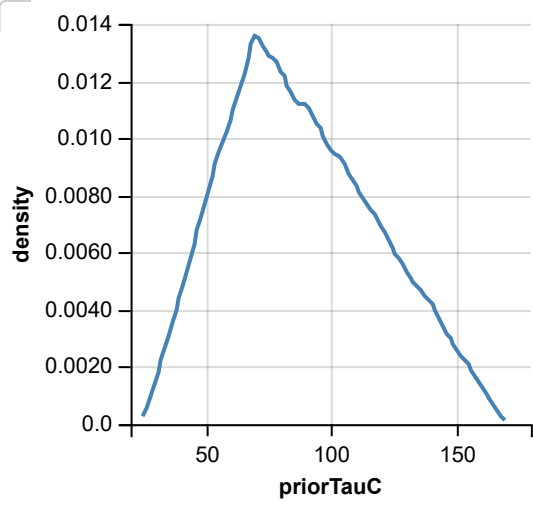
-----

Univariate Priors TauX (X=P, C, M, Sum, SigmaTauSum) ~ Gamma(???, ???)
priorTauP:

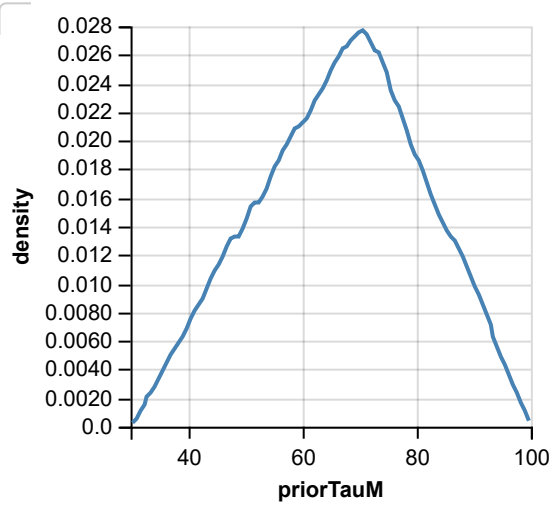
```



priorTauC:

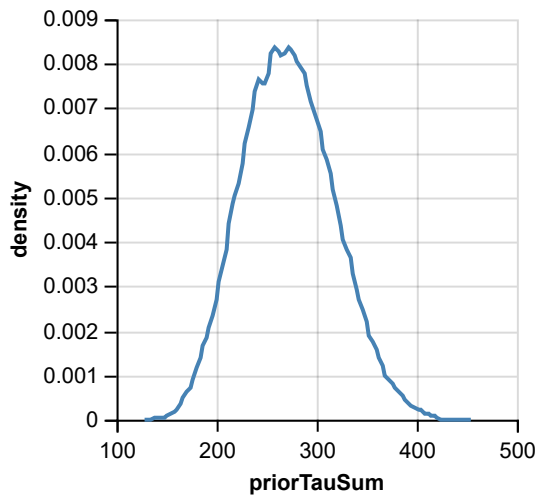


priorTauM:

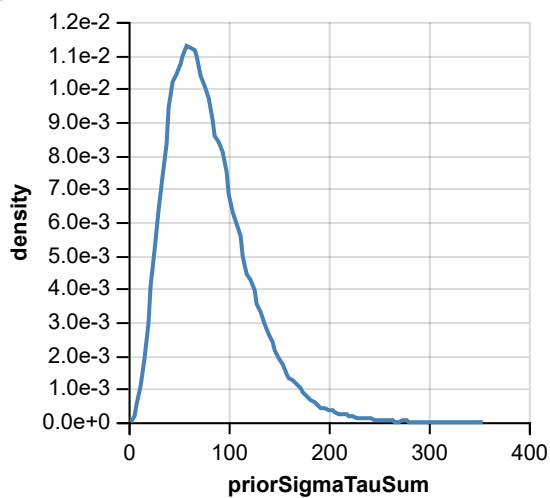


priorTauSum:





priorSigmaTauSum: ▼



----- ▼  
 model-generated 3\*sigma tau-interval: ▼

priorTauP ▼

mode = unknown ▼

mean = 116.69741323038144 ▼

sigma = 31.369305989517542 ▼

{fast:22.589495261828816 mean:116.69741323038144 slow:210.80533119893408} ▼

----- ▼  
 priorTauC ▼

mode = unknown ▼

mean = 88.25270878644645 ▼

sigma = 30.252110728266686 ▼

{fast:-2.5036233983536107 mean:88.25270878644645 slow:179.00904097124652} ▼

----- ▼  
 priorTauM ▼

mode = unknown ▼

mean = 66.73448775450258 ▼

sigma = 14.355816524021051 ▼

{fast:23.667038182439427 mean:66.73448775450258 slow:109.80193732656574} ▼

----- ▼  
 priorTauSum ▼

mode = unknown ▼

```
mean = 271.6846097713323  
sigma = 45.893075153745926  
{fast:134.00538431009454 mean:271.6846097713323 slow:409.36383523257007}
```

-----  
priorSigmaTauSum

mode = unknown

mean = 79.82583170579565

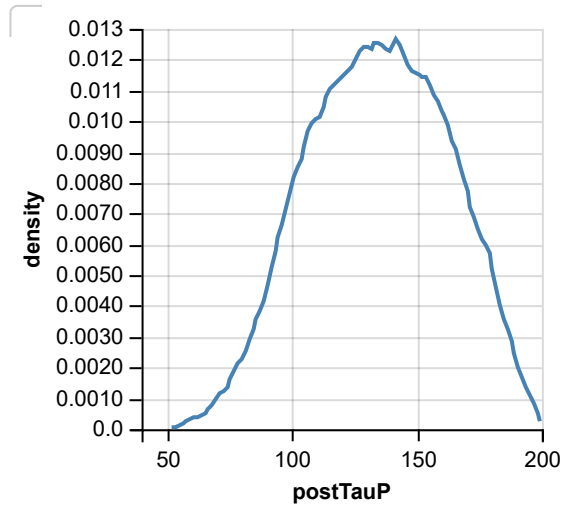
sigma = 40.1605487409151

model-generated 3\*sigma tau-interval:

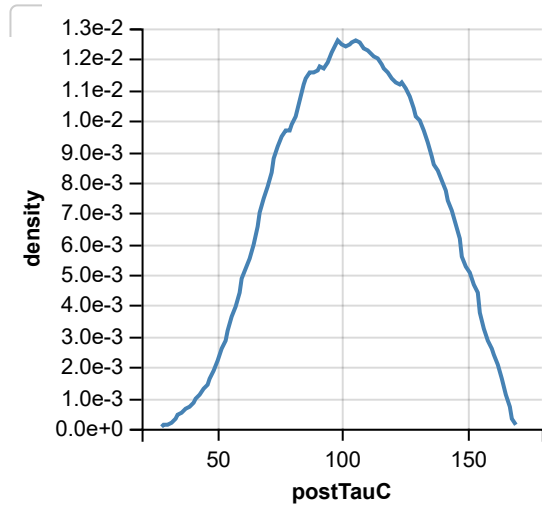
```
{fast:-40.65581451694965 mean:79.82583170579565 slow:200.30747792854095}
```

=====  
Univariate Posteriors TauX (X=P, C, M, Sum) Gamma(???, ???) and SigmaTauSum Gamma(???, ???)

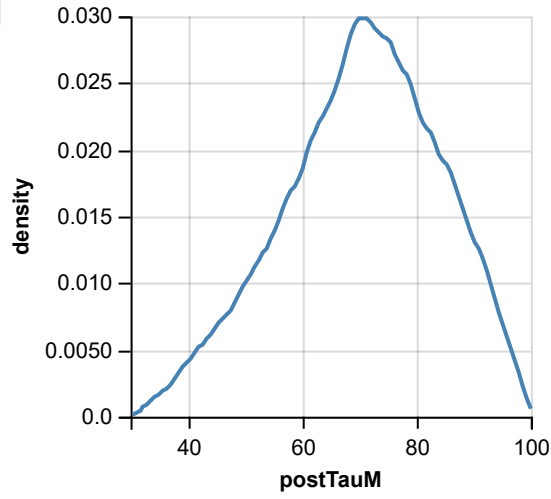
postTauP:



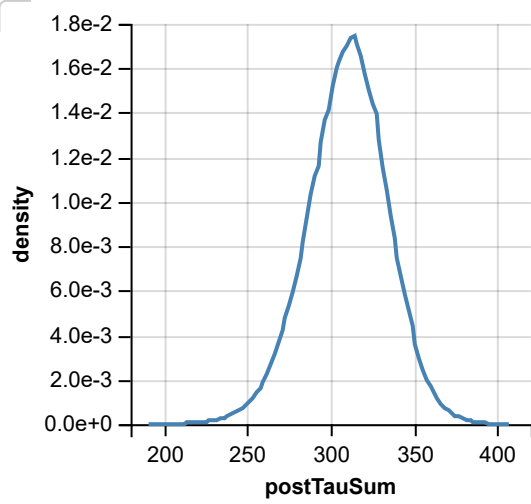
postTauC:



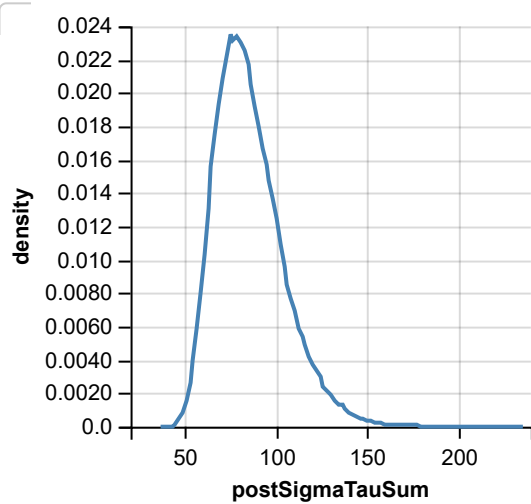
postTauM:



postTauSum:



postSigmaTauSum:



-----  
 model-generated 3\*sigma tau-interval:

postTauP

mode = unknown

mean = 134.47664140029102

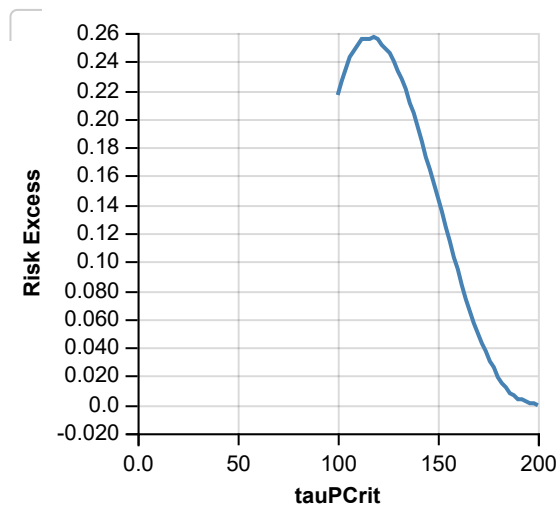
sigma = 27.949695322751094

{fast:50.62755543203774 mean:134.47664140029102 slow:218.32572736854428}

-----  
 tauPCrit = 172; increase in risk probs = 0.042799999999992955

```

tauPCrit = 174; increase in risk probs = 0.0372166666666660236
tauPCrit = 176; increase in risk probs = 0.0310333333333327584
tauPCrit = 178; increase in risk probs = 0.026366666666666132
tauPCrit = 180; increase in risk probs = 0.019749999999995382
tauPCrit = 182; increase in risk probs = 0.0150499999999959
tauPCrit = 184; increase in risk probs = 0.012849999999996031
tauPCrit = 186; increase in risk probs = 0.008549999999996505
tauPCrit = 188; increase in risk probs = 0.006633333333330049
tauPCrit = 190; increase in risk probs = 0.004449999999996956
tauPCrit = 192; increase in risk probs = 0.0031499999999969885
tauPCrit = 194; increase in risk probs = 0.002333333333330412
tauPCrit = 196; increase in risk probs = 0.0009999999999972253
tauPCrit = 198; increase in risk probs = 0.00041666666666395624
tauPCrit = 200; increase in risk probs = -2.6645352591003757e-15
    
```



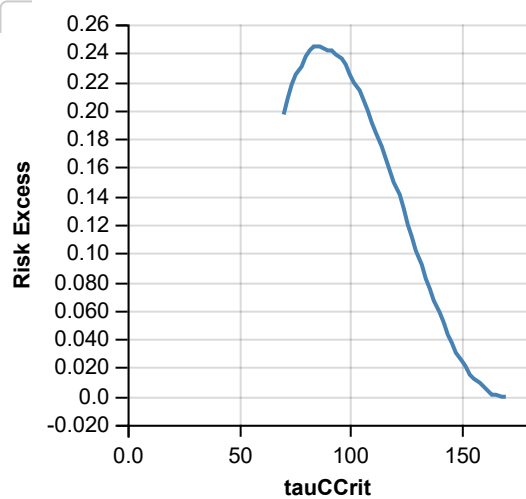
```

-----
postTauC
mode = unknown
mean = 105.09391516556296
sigma = 27.820525848789224
{fast:21.632337619195297 mean:105.09391516556296 slow:188.55549271193064}
    
```

```

-----
tauCCrit = 144; increase in risk probs = 0.04353333333332665
tauCCrit = 146; increase in risk probs = 0.03673333333332729
tauCCrit = 148; increase in risk probs = 0.030783333333327945
tauCCrit = 150; increase in risk probs = 0.024916666666661924
tauCCrit = 152; increase in risk probs = 0.02079999999999571
tauCCrit = 154; increase in risk probs = 0.01503333333332968
tauCCrit = 156; increase in risk probs = 0.01158333333333006
tauCCrit = 158; increase in risk probs = 0.008716666666663597
tauCCrit = 160; increase in risk probs = 0.006433333333330515
tauCCrit = 162; increase in risk probs = 0.0031833333333308733
tauCCrit = 164; increase in risk probs = 0.0015833333333310495
tauCCrit = 166; increase in risk probs = 0.00031666666666441134
tauCCrit = 168; increase in risk probs = 0.00006666666666443888
    
```

```
tauCCrit = 170; increase in risk probs = -2.220446049250313e-15
```



```
-----
```

```
postTauM
```

```
mode = unknown
```

```
mean = 70.32273725361163
```

```
sigma = 13.62182402910084
```

```
{fast:29.457265166309114 mean:70.32273725361163 slow:111.18820934091416}
```

```
-----
```

```
tauMCrit = 83.8; increase in risk probs = 0.047099999999994036
```

```
tauMCrit = 84.4; increase in risk probs = 0.04434999999999423
```

```
tauMCrit = 85; increase in risk probs = 0.04141666666666122
```

```
tauMCrit = 85.6; increase in risk probs = 0.03759999999999508
```

```
tauMCrit = 86.2; increase in risk probs = 0.034883333333328714
```

```
tauMCrit = 86.8; increase in risk probs = 0.03258333333332886
```

```
tauMCrit = 87.4; increase in risk probs = 0.029233333333329226
```

```
tauMCrit = 88; increase in risk probs = 0.027616666666662737
```

```
tauMCrit = 88.6; increase in risk probs = 0.02496666666666303
```

```
tauMCrit = 89.2; increase in risk probs = 0.023099999999996568
```

```
tauMCrit = 89.8; increase in risk probs = 0.021099999999996788
```

```
tauMCrit = 90.4; increase in risk probs = 0.019149999999997003
```

```
tauMCrit = 91; increase in risk probs = 0.01731666666666387
```

```
tauMCrit = 91.6; increase in risk probs = 0.01506666666666412
```

```
tauMCrit = 92.2; increase in risk probs = 0.01278333333331037
```

```
tauMCrit = 92.8; increase in risk probs = 0.01118333333331213
```

```
tauMCrit = 93.4; increase in risk probs = 0.00908333333331445
```

```
tauMCrit = 94; increase in risk probs = 0.00781666666664918
```

```
tauMCrit = 94.6; increase in risk probs = 0.00653333333331726
```

```
tauMCrit = 95.2; increase in risk probs = 0.00531666666665082
```

```
tauMCrit = 95.8; increase in risk probs = 0.00413333333331879
```

```
tauMCrit = 96.4; increase in risk probs = 0.003049999999998665
```

```
tauMCrit = 97; increase in risk probs = 0.00246666666665396
```

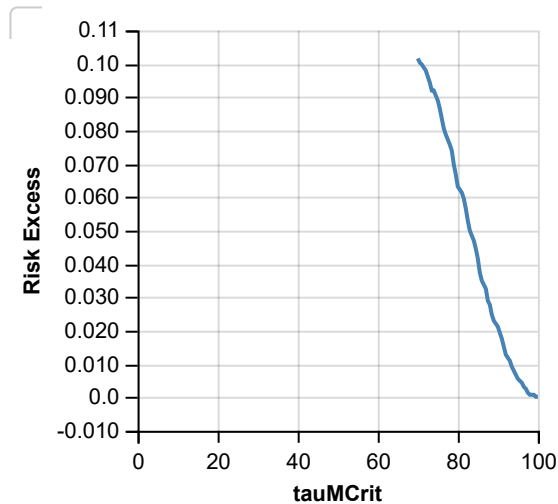
```
tauMCrit = 97.6; increase in risk probs = 0.001183333333322038
```

```
tauMCrit = 98.2; increase in risk probs = 0.000616666666655995
```

```
tauMCrit = 98.8; increase in risk probs = 0.000616666666655995
```

tauMCrit = 99.4; increase in risk probs = 0.00008333333333332495

tauMCrit = 100; increase in risk probs = -9.992007221626409e-16



-----

postTauSum

mode = unknown

mean = 309.89329381946413

sigma = 24.21338842156764

{fast:237.25312855476122 mean:309.89329381946413 slow:382.533459084167}

-----

tauSumCrit = 336.6; increase in risk probs = 0.04188333333332872

tauSumCrit = 341.2; increase in risk probs = 0.01868333333331276

tauSumCrit = 345.8; increase in risk probs = 0.00214999999999763

tauSumCrit = 350.4; increase in risk probs = -0.00821666666665762

tauSumCrit = 355; increase in risk probs = -0.01339999999998524

tauSumCrit = 359.6; increase in risk probs = -0.014833333333317

tauSumCrit = 364.2; increase in risk probs = -0.01558333333331617

tauSumCrit = 368.7999999999995; increase in risk probs = -0.01446666666665073

tauSumCrit = 373.4; increase in risk probs = -0.01294999999998574

tauSumCrit = 378; increase in risk probs = -0.01051666666665508

tauSumCrit = 382.6; increase in risk probs = -0.00881666666665696

tauSumCrit = 387.2; increase in risk probs = -0.00681666666665916

tauSumCrit = 391.7999999999995; increase in risk probs = -0.00529999999999416

tauSumCrit = 396.4; increase in risk probs = -0.00401666666666224

tauSumCrit = 401; increase in risk probs = -0.00294999999999675

tauSumCrit = 405.6; increase in risk probs = -0.00188333333333126

tauSumCrit = 410.2; increase in risk probs = -0.001166666666665382

tauSumCrit = 414.7999999999995; increase in risk probs = -0.000683333333332581

tauSumCrit = 419.4; increase in risk probs = -0.00033333333332966

tauSumCrit = 424; increase in risk probs = -0.000183333333331314

tauSumCrit = 428.6; increase in risk probs = -0.00016666666666483

tauSumCrit = 433.2; increase in risk probs = -0.000066666666665932

tauSumCrit = 437.7999999999995; increase in risk probs = -0.000033333333332966

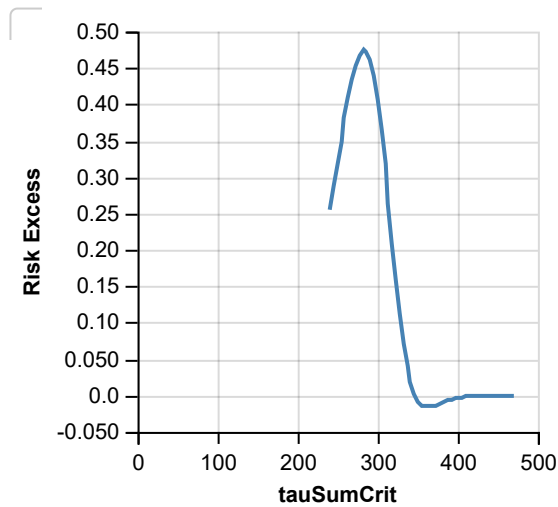
tauSumCrit = 442.4; increase in risk probs = -0.000016666666666483

tauSumCrit = 447; increase in risk probs = -0.000016666666666483

```

tauSumCrit = 451.6; increase in risk probs = -0.00001666666666666483
tauSumCrit = 456.2; increase in risk probs = 0
tauSumCrit = 460.79999999999995; increase in risk probs = 0
tauSumCrit = 465.4; increase in risk probs = 0
tauSumCrit = 470; increase in risk probs = 0

```



```

-----
postSigmaTauSum
mode = unknown
mean = 85.56243421239003
sigma = 18.957748415066664
{fast:28.689188967190034 mean:85.56243421239003 slow:142.43567945759003}
=====
runtime in seconds = 8292.426
runtime in minutes = 138.2071
=====

```

## Features

- Runs on the command line with node.js (<http://nodejs.org/>) or in the browser (<http://docs.webppl.org/en/master/development/workflow.html#browser-version>).
- Supports modular and re-usable code using packages (<http://docs.webppl.org/en/master/packages.html>) built on top of the npm package system, and interoperates with existing Javascript packages in the npm ecosystem.
- Includes a large and expanding library of primitive distributions. (<http://docs.webppl.org/en/master/distributions.html>)
- Implements a variety of inference algorithms (<http://docs.webppl.org/en/master/inference/index.html>), including exact inference via enumeration, rejection sampling, Sequential Monte Carlo, Markov Chain Monte Carlo, Hamiltonian Monte Carlo, and inference-as-optimization (e.g. variational inference).
- Provides inference as a first-class operator in the language, allowing for nested inference ('inference about inference').

- Supports optimizable models with neural network components using `adnn` (<https://www.npmjs.com/package/adnn>).

## Demos

Browser-based applications powered by WebPPL.

- Procedural vines with shape constraints (<demos/vines/index.html>)
- 3D procedural spaceships with shape constraints (<http://dritchier.github.io/web-procmod/>)  
(Note: the code in this demo is written in an older version of WebPPL)

## Local install

Install WebPPL in two easy steps:

1. Install `node.js` (<http://nodejs.org>)
2. Run `npm install -g webppl`

Now, the `webppl` command is globally available.

To upgrade to the latest version, run `npm update -g webppl`.

## Documentation

To learn more about how to set up and use WebPPL, take a look at our documentation (<http://docs.webppl.org>) and the examples (<https://github.com/probmods/webppl/tree/master/examples>).

To learn more about how WebPPL works under the hood, check out our web book, *The Design and Implementation of Probabilistic Programming Languages* (<http://dippl.org/>).

For probabilistic modeling in general, our other web book, *Probabilistic Models of Cognition* (<https://probmods.org>), might be of interest.

## License

The WebPPL code base is open source and freely available for commercial and non-commercial use under the MIT license (<https://github.com/probmods/webppl/blob/master/LICENSE.md>).

## Contributions

We encourage you to contribute to WebPPL! Check out our guidelines for contributors (<https://github.com/probmods/webppl/blob/master/CONTRIBUTING.md>) and join the `webppl-dev` (<https://groups.google.com/forum/#!forum/webppl-dev>) mailing list.

## Pronunciation

Say “web people”.

## Citing

If you use WebPPL in academic projects and papers, please cite as:

*N. D. Goodman and A. Stuhlmüller (electronic). The Design and Implementation of Probabilistic Programming Languages. Retrieved from <http://dippl.org> . [bibtex]*



## Publications

If you publish a paper using/extending WebPPL, let us know

(<https://groups.google.com/forum/#!forum/webppl-dev>) and we'll add it to this list:

*D. Ritchie, P. Horsfall, and N. D. Goodman. Deep Amortized Inference for Probabilistic Programs (<https://arxiv.org/abs/1610.05735>). arXiv:1610.05735.*

*L. Ouyang, M. H. Tessler, D. Ly, and N. D. Goodman. Practical optimal experiment design with probabilistic programs (<https://arxiv.org/abs/1608.05046>). arXiv:1608.05046.*

*M. H. Tessler and N. D. Goodman. A Pragmatic Theory of Generic Language (<https://arxiv.org/abs/1608.02926>). arXiv:1608.02926.*

*D. Ritchie, A. Thomas, P. Hanrahan, and N. D. Goodman. Neurally-Guided Procedural Models: Amortized Inference for Procedural Graphics Programs using Neural Networks (<https://arxiv.org/abs/1603.06143>). NIPS 2016.*

*D. Ritchie, A. Stuhlmüller, and N. D. Goodman. C3: Lightweight Incrementalized MCMC for Probabilistic Programs using Continuations and Callsite Caching (<https://arxiv.org/abs/1509.02151>). AISTATS 2016.*

*M. H. Tessler and N. D. Goodman. Communicating generalizations about events (<http://stanford.edu/~mtessler/papers/Tessler2016-cogsci.pdf>). Proceedings of the Thirty-Eighth Annual Conference of the Cognitive Science Society, 2016.*

*E. J. Yoon, M. H. Tessler, N. D. Goodman, and M. C. Frank. Talking with tact: Polite language as a balance between kindness and informativity (<http://stanford.edu/~mtessler/papers/YoonTessler2016-cogsci.pdf>). Proceedings of the Thirty-Eighth Annual Conference of the Cognitive Science Society, 2016.*

*C. Graf, J. Degen, R. X. D. Hawkins, and N. D. Goodman. Animal, dog, or dalmatian? Level of abstraction in nominal referring expressions (<https://cocolab.stanford.edu/papers/GrafEtAl2016-Cogsci.pdf>). Proceedings of the Thirty-Eighth Annual Conference of the Cognitive Science Society, 2016.*

*O. Evans, A. Stuhlmüller, and N. D. Goodman. Learning the Preferences of Ignorant, Inconsistent Agents (<https://stuhlmuller.org/papers/preferences-aaai2016.pdf>). AAI 2016.*

*A. Stuhlmüller, R. X. D. Hawkins, N. Siddharth, and N. D. Goodman. Coarse-to-Fine Sequential Monte Carlo for Probabilistic Programs (<https://arxiv.org/abs/1509.02962>). arXiv:1509.02962.*

*O. Evans, A. Stuhlmüller, and N. D. Goodman. Learning the Preferences of Bounded Agents (<https://stuhlmuller.org/papers/preferences-nipsworkshop2015.pdf>). Workshop on Bounded Optimality, NIPS 2015.*

*R. X. D. Hawkins, A. Stuhlmüller, J. Degen, and N. D. Goodman. Why do you ask? Good questions provoke informative answers (<https://stuhlmuller.org/papers/qa-cogsci2015.pdf>). Proceedings of the Thirty-Seventh Annual Conference of the Cognitive Science Society, 2015.*

*G. Scontras and M. H. Tessler (electronic). Composition in Probabilistic Language Understanding (<http://gscontras.github.io/ESSLLI-2016/>). Retrieved from <http://gscontras.github.io/ESSLLI-2016/>.*

*O. Evans, A. Stuhlmüller, J. Salvatier, and D. Filan (electronic). Modeling Agents with Probabilistic Programs (<http://agentmodels.org>). Retrieved from <http://agentmodels.org>.*

*N. D. Goodman and J. B. Tenenbaum (electronic). Probabilistic Models of Cognition (<http://probmods.org>). Retrieved from <http://probmods.org>.*

*N. D. Goodman and A. Stuhlmüller (electronic). The Design and Implementation of Probabilistic Programming Languages (<http://dippl.org>). Retrieved from <http://dippl.org>.*

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