# CPSC 532W - Homework 5

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Public GitHub repo: https://github.com/Xiaoxuan1121/CPSC532W/tree/main/a5

#### 1. Program 1:

```
FOPPL Tests passed
Test passed
All deterministic tests passed
```

('normal', 5, 1.4142136)
p value 0.5850982269507421
('beta', 2.0, 5.0)
p value 0.42946782406988304
('exponential', 0.0, 5.0)
p value 0.12652924669180987
('normal', 5.3, 3.2)
p value 0.3763317545140774
('normalmix', 0.1, -1, 0.3, 0.9, 1, 0.3)
p value 0.35332451374583096
('normal', 0, 1.44)
p value 0.7920528096804577
All probabilistic tests passed

```
class Env(dict):
   def __init__(self, parms=(), args=(), outer=None):
      self.update(zip(parms, args))
      self.outer = outer
   def get(self, var):
     return self[var] if (var in self) else self.outer.get(var)
class Procedure(object):
   def __init__(self, parms, body, env):
     self.parms, self.body, self.env = parms, body, env
   def __call__(self, *args):
   return evaluate(self.body, Env(self.parms, args, self.env))
def evaluate(exp, env=None):#TODO: add sigma, or something
    if env is None:
       env = standard_env()
       proc = evaluate(exp, env)
       return proc([""])
    if isinstance(exp, str):
       if env.get(exp) is not None:
           return env.get(exp)
       return exp
    elif not isinstance(exp, list):
       return torch.tensor(float(exp))
    op, *args = exp
    if op == 'if':
        (test, conseq, alt) = args
        exp = (conseq if evaluate(test, env) else alt)
        return evaluate(exp, env)
    elif op == 'fn':
       parms, body = args
        if len(parms) > 1:
         parms = parms[1:]
        else:
           parms = []
        return Procedure(parms, body, env)
    elif op == 'sample':
        dist = evaluate(args[1], env)
        value = dist.sample()
       return value
    elif op == 'observe':
    return evaluate(args[-1], env)
    else:
       proc = evaluate(op, env)
        vals = [evaluate(arg, env) for arg in args[1:]]
        return proc(*vals)
```

```
env = {
    'normal' : Normal,
    'beta' : Beta,
    'exponential': dist.Exponential,
    'uniform-continuous' : dist.Uniform,
    'flip' : dist.Bernoulli,
    'discrete': dist.Categorical,
    'push-address'_: push_addr,
    '+'_: plus,
    '-'<sub>~</sub>: minus,
    '*'<sub>_</sub>: op.mul,
    '/': op.truediv,
    '>'_: op.gt,
    '<'_: op.lt,
    'sqrt' : torch.sqrt,
    'log' : torch.log,
    'vector': vector,
    'get': get,
    'put': put,
    'first': first,
    'last': last,
    'peek'_: first,
    'rest' : rest,
    'append': append,
    'hash-map' : hashmap,
    'procedure?'_: callable,
    'empty?'_: isempty,
    'cons' : cons,
    'conj' : append
```

```
def isempty(*exp):
    return len(exp[0]) == 0

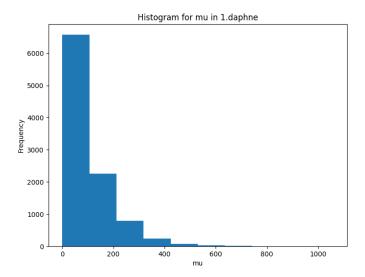
def cons(*exp):
    try:
        return torch.cat((torch.tensor([exp[1]]), exp[0]), dim=0)
    except:
    return torch.cat((torch.tensor([exp[1]]), torch.tensor(exp[0])), dim = 0)
```

## 2. Program 2:

running 1.daphne took 92.527643 seconds

Prior expectation of mu in 1.daphne is 97.88899993896484

Prior variance of mu in 1.daphne is 9446.9794921875

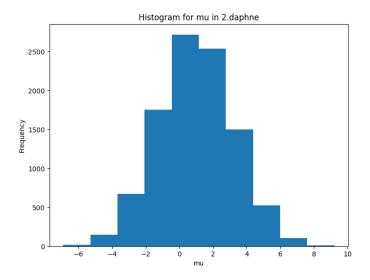


## 3. Program 3:

running 2.daphne took 4.908595 seconds

Prior expectation of mu in 1.daphne is 0.9962315559387207

Prior variance of mu in 1.daphne is 4.93194580078125



4. Program 4:

running 3.daphne took 26.665322 seconds

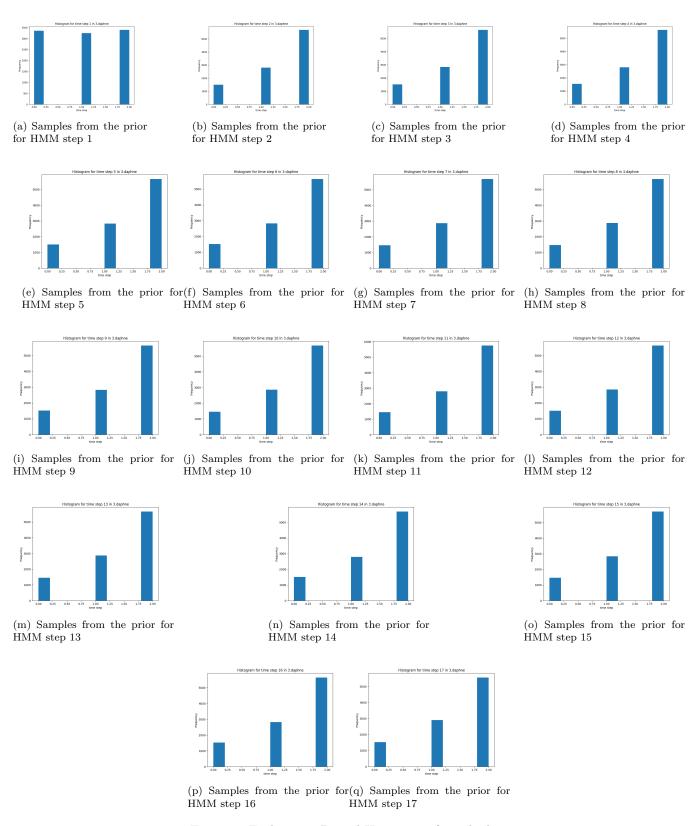


Figure 1: Evaluation: Partial Histograms for 3.daphne