## Fig1

## February 8, 2019

```
In [1]: import numpy as np
        import matplotlib.pyplot as plt
        ### Set random seed
        np.random.seed(3)
        ### Initialize the parameters
        d = 10**2
        n = d
        n_{iter} = 10**2
        ### Initialize the variables
        x_0 = 1./np.sqrt(d)*np.random.randn(d)
        y_0 = 1./np.sqrt(d)*np.random.randn(d)
        norm M = 0
        M = np.zeros((n,d,d))
        b = np.zeros((n,d))
        a = np.zeros((n,d))
        ### Simple bilinear objective.
        for i in range(d):
            M[i] = np.zeros((d,d))
            M[i][i,i] = 1
            a[i,:] = 1./np.sqrt(d)*np.random.randn(d)
            b[i,:] = 1./np.sqrt(d)*np.random.randn(d)
        ### Compute the solution
        M_{mean} = np.mean(M,0)
        b_mean = np.mean(b,0)
        a_mean = np.mean(a,0)
        y_star = np.linalg.solve(M_mean,-a_mean)
        x_star = np.linalg.solve(M_mean.T,-b_mean)
        ### To save the data
        Results = {}
In [2]: # These are the "Tableau 20" colors as RGB.
```

```
tableau20 = [(31, 119, 180), (174, 199, 232), (255, 127, 14), (255, 187, 120),
                     (44, 160, 44), (152, 223, 138), (214, 39, 40), (255, 152, 150),
                     (148, 103, 189), (197, 176, 213), (140, 86, 75), (196, 156, 148),
                     (227, 119, 194), (247, 182, 210), (127, 127, 127), (199, 199, 199),
                     (188, 189, 34), (219, 219, 141), (23, 190, 207), (158, 218, 229)]
        \# Scale the RGB values to the [0, 1] range, which is the format matplotlib accepts.
        for i in range(len(tableau20)):
            red, green, blue = tableau20[i]
            tableau20[i] = (red / 255., green / 255., blue / 255.)
In [3]: from IPython.core.interactiveshell import InteractiveShell
        InteractiveShell.ast_node_interactivity = "all"
        import pylab
        pylab.rcParams['figure.figsize'] = (8, 5)
        import numpy as np
        import matplotlib.pyplot as plt
        import json
        plt.rcParams['text.latex.preamble'] = [r"\usepackage{lmodern}"]
        params = {'text.usetex': True,
                 'font.size': 15,
                 'font.family': 'lmodern',
                 'text.latex.unicode': True,
                 }
        plt.rcParams.update(params)
/home/gidelgau/anaconda3/lib/python3.7/site-packages/matplotlib/__init__.py:846: MatplotlibDep
The text.latex.unicode rcparam was deprecated in Matplotlib 2.2 and will be removed in 3.1.
  "2.2", name=key, obj_type="rcparam", addendum=addendum)
In [4]: class adam:
            def \underline{init} (self,M,a,b, beta_1 = 0,beta_2 = .99,epsilon_0 = 1e-16):
                self.epsilon_0 = epsilon_0
                self.beta_1 = beta_1
                self.beta_2 = beta_2
                self.first_moment_x = np.zeros(d)
                self.first_moment_y = np.zeros(d)
                self.second_moment_y = np.zeros(d)
                self.second_moment_x = np.zeros(d)
                self.M = M
                self.a = a
                self.b = b
            def adam_direction(self,x,y,x_snap,y_snap,i,t):
                (d_x,d_y) = SVRG_dir(x,y,x_snap,y_snap,i)
                self.first_moment_x = self.beta_1 * self.first_moment_x + (1- self.beta_1)* d_:
                self.second_moment_x = self.beta_2 * self.second_moment_x + (1-self.beta_2) *
                self.first_moment_y = self.beta_1 * self.first_moment_y + (1- self.beta_1)* d_
```

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d_x_t = d_x * np.abs(self.first_moment_x) / (1-self.beta_1**(t+1)) / (np.sqrt_moment_x)
                d_y_t = d_y * np.abs(self.first_moment_y) / (1-self.beta_1**(t+1)) / (np.sqrt
                return (d_x_t,d_y_t)
            def old_adam_direction(self,x,y,x_snap,y_snap,i,t):
                (d_x,d_y) = SVRG_dir(x,y,x_snap,y_snap,i)
                self.first_moment_x = self.beta_1 * self.first_moment_x + (1- self.beta_1)* d_:
                self.second_moment_x = self.beta_2 * self.second_moment_x + (1-self.beta_2) *
                self.first_moment_y = self.beta_1 * self.first_moment_y + (1- self.beta_1)* d_
                self.second_moment_y = self.beta_2 * self.second_moment_y + (1-self.beta_2) *
                d x_t = self.first_moment_x / (1-self.beta_1**(t+1)) / (np.sqrt(self.second_moment_x)
                d_y_t = self.first_moment_y / (1-self.beta_1**(t+1)) / (np.sqrt(self.second_m.
                return (d_x_t,d_y_t)
            def adadelta_direction(self,x,y,x_snap,y_snap,i,t):
                (d_x,d_y) = SVRG_dir(x,y,x_snap,y_snap,i)
                self.first_moment_x = self.beta_2 * self.first_moment_x + (1- self.beta_2)* d_:
                self.second_moment_x = self.beta_2 * self.second_moment_x + (1-self.beta_2) *
                self.first_moment_y = self.beta_2 * self.first_moment_y + (1- self.beta_2)* d_
                self.second_moment_y = self.beta_2 * self.second_moment_y + (1-self.beta_2) *
                d_x_t = d_x * self.first_moment_x **2 / (self.second_moment_x+self.epsilon_0)
                d_y_t = d_y * self.first_moment_y **2 / (self.second_moment_y+self.epsilon_0)
                return (d x t,d y t)
In [5]: class gradient:
            def __init__(self,M,a,b ,gamma = .1):
                self.M = M
                self.a = a
                self.b = b
                self.gamma = gamma
            def gradient_direction(self,x_t,y_t,i):
                x_grad = (self.M[i].dot(y_t) + self.a[i])
                y_grad = (self.M[i].T.dot(x_t) + self.b[i])
                return (-self.gamma * x_grad,self.gamma * y_grad)
In [6]: def proj(x):
            n = x.shape[0]
            \#return\ np.minimum(np.ones(n), np.maximum(-np.ones(n), x))
            return x
        def SVRG_dir(x,y,x_snap,y_snap,i):
            d_x = M[i].dot(y) - M[i].dot(y_snap) + M_mean.dot(y_snap) + a_mean
            d_y = M[i].T.dot(x) - M[i].T.dot(x_snap) + M_mean.T.dot(x_snap) + b_mean
            return(d_x,d_y)
        def f_i(x,y,i):
            return np.sum(x * M[i].dot(y) + a[i]*x+ b[i]*y)
```

self.second\_moment\_y = self.beta\_2 \* self.second\_moment\_y + (1-self.beta\_2) \*

```
def progress(x,y,i,L,d_x,d_y,f):
                           x_ = x + d_x
                           y_{-} = y + d_{-}y
                           norm_grad = np.sum((d_x)**2 + (d_y)**2)
                            return np.abs(f(x_y,i) - f(x,y,i) - np.sum(d_x * (M[i].dot(y) + a[i]) + d_y * (M[i].dot(y) + a[i]) + 
                  def lip_test(i,L,d,d_,x,x_,y,y_):
                           return np.linalg.norm(d_ - d) < L * np.sqrt(np.linalg.norm(x-x_)**2 + np.linalg.norm
In [7]: f = 0
                  for i in range(n):
                            f +=f_i(x_star,y_star,i)
                  print(f)
9.615773703310373
In [8]: def AdamSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,alpha = .1, beta_1 = .99,beta_2= ...
                            adam_grad = adam(M,a,b ,alpha = alpha, beta_1 = beta_1,beta_2= beta_2,epsilon_0 = 6
                            l_norm_adam = np.zeros((nb_rand,10*n_pass+1))
                           l_norm_av = np.zeros((nb_rand,10*n_pass+1))
                            for ran in range(nb_rand):
                                     x_t = x_0.copy()
                                     y_t = y_0.copy()
                                     x_t_av = x_t.copy()
                                     y_t_av = y_t.copy()
                                     norm_grad = np.linalg.norm(x_t-x_star) + np.linalg.norm(y_t-y_star)
                                     1_norm_adam[ran,0] = norm_grad
                                     l_norm_av[ran,0] = norm_grad
                                     S_t = 1
                                     for t in range(int(n_pass*n)):
                                              i = np.random.randint(n)
                                              for k in range(5):
                                                       d_t = adam_grad.adam_direction(x_t,y_t,i,t)
                                                       x_t = proj(x_t + d_t[0])
                                              d_t = adam_grad.adam_direction(x_t,y_t,i,t)
                                              x_t_av += x_t
                                              y_t = proj(y_t+d_t[1])
                                              y_t_av += y_t
                                              S_t += 1
                                              if np.mod(t*10,n) == 0:
                                                       norm_grad = np.linalg.norm(x_t-x_star) + np.linalg.norm(y_t-y_star)
                                                       l_norm_adam[ran,10*t//n+1] = norm_grad
                                                       norm_grad = np.linalg.norm(x_t_av/S_t-x_star) + np.linalg.norm(y_t_av/S_t)
                                                       l_norm_av[ran, 10*t//n+1] = norm_grad
                            return (l_norm_adam,l_norm_av)
In [9]: def AvgSGD(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = .1, restart = .1):
                            1_SimSGD = np.zeros((nb_rand,10*n_pass+1))
```

```
for ran in range(nb_rand):
                                      x_grad = x_0.copy()
                                      y_grad = y_0.copy()
                                      x_grad_av = x_0.copy()
                                      y_grad_av = y_0.copy()
                                      St=1
                                      norm_grad = np.linalg.norm(x_grad_av/S_t-x_star) + np.linalg.norm(y_grad_av/S_-
                                      1_SimSGD[ran,0] = norm_grad
                                      1_AvgSGD[ran,0] = norm_grad
                                      for t in range(int(n_pass * n)):
                                                 \# qamma = 0.1
                                                if restart is not None:
                                                          if np.random.binomial(1,1./n * restart):
                                                                   x_grad = x_grad_av.copy()/S_t
                                                                   y_grad = y_grad_av.copy()/S_t
                                                                   S_t = 1
                                                                   x_grad_av = x_grad.copy()
                                                                   y_grad_av = y_grad.copy()
                                                i = np.random.randint(n)
                                                x_grad = proj(x_grad - gamma * (M[i].dot(y_grad) + a[i]))
                                                y_grad = proj(y_grad + gamma * (M[i].T.dot(x_grad)+ b[i]))
                                                x_grad_av += x_grad
                                                y_grad_av += y_grad
                                                S_t += 1
                                                if np.mod(t*10,n) == 0:
                                                          norm_grad = np.linalg.norm(x_grad-x_star) + np.linalg.norm(y_grad-y_star)
                                                          1_SimSGD[ran,10*t//n+1] = norm_grad
                                                         norm_grad = np.linalg.norm(x_grad_av/S_t-x_star) + np.linalg.norm(y_grad_av/S_t-x_star)
                                                          1_AvgSGD[ran,10*t//n+1] = norm_grad
                             return (1_SimSGD,1_AvgSGD)
In [10]: def ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = .1,linesearch= None,restar,n,gamma = .1,linesearch= 
                               l_SEM = np.zeros((nb_rand, 10*n_pass//2+1))
                               1_avg_SEM = np.zeros((nb_rand,10*n_pass//2+1))
                               l_grad_SEM = np.zeros((nb_rand,10*n_pass//2+1))
                               if ada:
                                                   adam_grad = adam(M,a,b,beta_1 = 0,beta_2= beta_2)
                               for ran in range(nb_rand):
                                         x_grad = x_0.copy()
                                         y_grad = y_0.copy()
                                         x_avg = x_0.copy()
                                         y_avg = y_0.copy()
                                         x_snap = x_0.copy()
                                         y_{snap} = y_{0.copy}()
                                         S_t = 1
                                         gamma_x = gamma
                                         gamma_y = gamma
```

1\_AvgSGD = np.zeros((nb\_rand,10\*n\_pass+1))

```
l_grad = []
         dist_opt = np.linalg.norm(x_grad-x_star) + np.linalg.norm(y_grad-y_star)
         1_SEM[ran,0] = dist_opt
        l_avg_SEM[ran,0] = dist_opt
         for t in range(n_pass*n //2):
                  if np.random.binomial(1,1./n):
                           if restart is not None:
                                    if np.random.binomial(1,restart):
                                             x_grad = x_avg.copy()/S_t
                                             y_grad = y_avg.copy()/S_t
                                             S_t = 1
                                             x_avg = x_grad.copy()
                                             y_avg = y_grad.copy()
                          x_{snap} = x_{grad.copy}()
                          y_snap = y_grad.copy()
                  # First step
                  i = np.random.randint(n)
                  if ada == "adam":
                           (d_x,d_y) = adam_grad.adam_direction(x_grad,y_grad,x_snap,y_snap,i,t)
                  elif ada == "delta":
                           (d_x,d_y) = adam_grad.adadelta_direction(x_grad,y_grad,x_snap,y_snap,
                  elif ada == "old adam":
                           (d_x,d_y) = adam_grad.old_adam_direction(x_grad,y_grad,x_snap,y_snap,
                          l_grad.append((d_x,d_y))
                  else:
                           (d_x,d_y) = SVRG_dir(x_grad,y_grad,x_snap,y_snap,i)
                  x_{extra} = proj(x_{grad} - gamma_x * d_x)
                  y_extra = proj(y_grad + gamma_y * d_y)
                  x_avg += x_grad
                  y_avg += y_grad
                  # Extrapolation step
                  i = np.random.randint(n)
                  if ada == "adam":
                           (d_x,d_y) = adam_grad.adam_direction(x_extra,y_extra,x_snap,y_snap,i,
                  elif ada == "delta":
                           (d_x,d_y) = adam_grad.adadelta_direction(x_extra,y_extra,x_snap,y_sna
                  elif ada == "old adam":
                           (d_x,d_y) = adam_grad.old_adam_direction(x_extra,y_extra,x_snap,y_sna
                  else:
                           (d_x,d_y) = SVRG_dir(x_extra,y_extra,x_snap,y_snap,i)
                  x_grad = proj(x_grad - gamma_x * d_x)
                  y_grad = proj(y_grad + gamma_y * d_y)
                  S_t += 1
                  if np.mod(t*10,n) == 0:
                           #print("diff point %f, diff grad:%f" % (np.linalg.norm(x_extra-x_snap)
                          1_SEM[ran,10*t//n+1] = np.linalg.norm(x_grad-x_star) + np.linalg.norm
                          l_avg_SEM[ran, 10*t//n+1] = np.linalg.norm(x_avg/S_t-x_star) + np.linalg.norm(x_avg/
return(1_SEM,1_avg_SEM)
```

```
In [11]: # ADAM baseline
        n_pass = 600
        nb\_rand = 5
         alpha = .002
        beta 1 = 0
         beta_2 = .999
         epsilon_0 = 1e-8
         gamma = .3
In [12]: l_adam = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma, beta_2 = 0.5
         \#l\_SimADAM = l\_adam[O]
         1_AvgADAM = 1_adam[1]
In [13]: l_old_adam = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma, beta_2 =
In [14]: l_grad = AvgSGD(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma, restart = .1)
         1\_SGD = 1\_grad[0]
         l_AvgSGD = l_grad[1]
In [15]: l_SVRG, l_avg_SVRG = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma)
In [16]: l_SVRG_2, l_avg_SVRG_2 = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gam
         1_SVRG_3, 1_avg_SVRG_3 = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma
         1_SVRG_4, 1_avg_SVRG_4 = ExtraSVRG(n_pass,nb_rand,x_0,y_0,x_star,y_star,n,gamma = gamma
In [17]: def av_std(1):
             av = np.mean(1,0)
             std = np.std(1,0)/2
             return (av,av + std,av-std)
         AltSGD1,supAltSGD,infAltSGD = av_std(l_AvgSGD)
         ExtraSVRG1,supExtraSVRG,infExtraSVRG = av_std(l_avg_SVRG)
         ExtraSVRG2,supExtraSVRG2,infExtraSVRG2 = av_std(1_SVRG_2)
         ExtraSVRG3,supExtraSVRG3,infExtraSVRG3 = av_std(1_SVRG_3)
         ExtraSVRG4,supExtraSVRG4,infExtraSVRG4 = av_std(1_SVRG_4)
         AdaSVRG1, supAdaSVRG, infAdaSVRG = av_std(l_adam[0])
         AdaSVRG2, supAdaSVRG2, infAdaSVRG2 = av_std(1_old_adam[0])
In [18]: ax = plt.subplot(111)
         ax.spines["top"].set_visible(False)
         ax.spines["bottom"].set_visible(False)
         ax.spines["right"].set_visible(False)
         ax.spines["left"].set_visible(False)
         # Ensure that the axis ticks only show up on the bottom and left of the plot.
         # Ticks on the right and top of the plot are generally unnecessary chartjunk.
         ax.get_xaxis().tick_bottom()
         ax.get_yaxis().tick_left()
```

```
1,1_{-},1 = av_{std}(1)
         t_extra = 2.*np.array(range(n_pass*10//2+1)) / 10.
         t_grad = np.array(range(n_pass*10+1)) / 10.
         plt.xlim([0,600])
         plt.ylim([.000005,3000])
         plt.semilogy(t_grad,AltSGD1,label="AVG-AltSGD", color = tableau20[1])
         plt.fill_between(t_grad,infAltSGD,supAltSGD, facecolor=tableau20[1], alpha = 0.3)
         plt.plot(t_extra,ExtraSVRG1,label=r"AVG-SVRE", color = tableau20[2])
         plt.fill_between(t_extra,infExtraSVRG,supExtraSVRG, facecolor=tableau20[2], alpha = 0
         \#plt.plot(t_extra,ExtraSVRG2,label=r"SVRE~p=1/20",~color~=~tableau20[3])
         \#plt.fill\_between(t\_extra,infExtraSVRG2,supExtraSVRG2,facecolor=tableau20[3],alpha
         plt.plot(t_extra, ExtraSVRG4,label=r"SVRE p=1/2", color = tableau20[5])
         plt.fill_between(t_extra,infExtraSVRG4,supExtraSVRG4, facecolor=tableau20[5], alpha =
         plt.plot(t_extra, ExtraSVRG3,label=r"SVRE p=1/10", color = tableau20[4])
         plt.fill_between(t_extra, infExtraSVRG3,supExtraSVRG3, facecolor=tableau20[4], alpha
         plt.plot(t_extra,AdaSVRG1,label=r"SVRE-VRAd p=1/10", color = tableau20[6])
         plt.fill_between(t_extra,infAdaSVRG,supAdaSVRG, facecolor=tableau20[6], alpha = 0.3)
         plt.plot(t_extra, AdaSVRG2,label=r"SVRE-A p=1/10", color = tableau20[7])
         plt.fill_between(t_extra, infAdaSVRG2, supAdaSVRG2, facecolor=tableau20[7], alpha = 0.3
         plt.xlabel('Number of passes')
         plt.ylabel('Distance to the optimum')
         plt.grid()
         plt.legend(fancybox = True, framealpha = .5)
         plt.savefig('bilinear_obj.pdf',bbox_inches='tight')
        plt.show()
Out[18]: (0, 600)
Out[18]: (5e-06, 3000)
Out[18]: [<matplotlib.lines.Line2D at 0x7f725ff17278>]
Out[18]: <matplotlib.collections.PolyCollection at 0x7f725fedea90>
Out[18]: [<matplotlib.lines.Line2D at 0x7f72800ad518>]
Out[18]: <matplotlib.collections.PolyCollection at 0x7f725fedeb38>
```

def plot\_with\_bar(1):

```
Out[18]: [<matplotlib.lines.Line2D at 0x7f725fedeef0>]
```

Out[18]: <matplotlib.collections.PolyCollection at 0x7f7271940710>

Out[18]: [<matplotlib.lines.Line2D at 0x7f72719404a8>]

Out[18]: <matplotlib.collections.PolyCollection at 0x7f7271940748>

Out[18]: [<matplotlib.lines.Line2D at 0x7f7271940a20>]

Out[18]: <matplotlib.collections.PolyCollection at 0x7f7271940c88>

Out[18]: [<matplotlib.lines.Line2D at 0x7f7271940f98>]

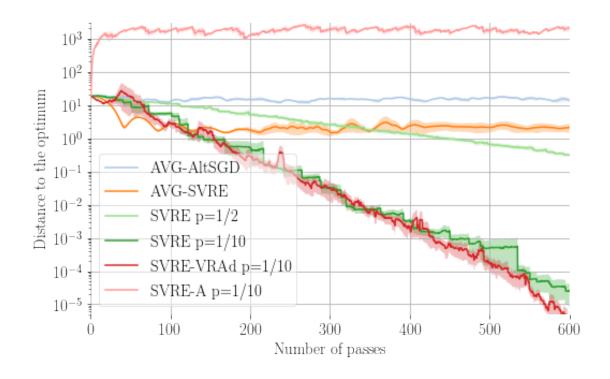
Out[18]: <matplotlib.collections.PolyCollection at 0x7f7271935208>

Out[18]: Text(0.5, 0, 'Number of passes')

Out[18]: Text(0, 0.5, 'Distance to the optimum')

Out[18]: <matplotlib.legend.Legend at 0x7f7271935550>

/home/gidelgau/anaconda3/lib/python3.7/site-packages/matplotlib/font\_manager.py:1241: UserWarn (prop.get\_family(), self.defaultFamily[fontext]))



## In []: