## A SUPPLEMENT

Algorithm 3 shows the details of how to sample a minibatch using k-DPP [19] which is used for the DM-SGD and DM-SVI algorithm in the paper.

Algorithm 3 Mini-batch Sampling
<b>Input:</b> Mini-batch size k, eigendecomposition
$\{(v_n, \lambda_n)\}_{n=1}^N$ of similarity matrix <i>L</i> .
Compute the elementary symmetric polynomials
$e_0^n \leftarrow 1 \forall n \in \{0, 1, 2, \dots, N\}$
$e_0^l \leftarrow 1 \forall l \in \{1, 2,, k\}$
for $l = 1, 2,, k$ do
for $n = 1, 2,, N$ do
$e_l^n \leftarrow e_l^{n-1} + \lambda_n e_{l-1}^{n-1}$
end
end
for t=1 to Number of subset samples to generate do
<b>Sampling</b> k eigenvectors V with indices J
$J \leftarrow \emptyset$
$l \leftarrow k$
for $n = N,, 2, 1$ do
if $l = 0$ then
break;
if $u \sim U[0,1] \leq \lambda_n \frac{e_{l-1}}{e^n}$ then
$  J \leftarrow J \cup \{n\}$
$l \leftarrow l-1$
end
end
<b>Sample</b> k data points indexed by Y using V.
$V \leftarrow \{v_i\}_{i \in J}$
$Y \leftarrow \emptyset$
while $ V  > 0$ do
Select <i>i</i> with $Pr(i) = \frac{1}{ V } \sum_{v \in V} (v^T e_i)^2$
$Y \leftarrow Y \cup i \ V \leftarrow V_{\perp}$ , an orthonormal basis for the
subspace of V orthogonal to $e_i$
end
Output: Y

## end

Table 2 and 3 show the top words using K = 30 for LDA using traditional SVI and our proposed DM-SVI respectively. We can see that the topics that are learned by DM-SVI are more diverse and rare topics such as grain (colored in blue) are captured.

Figure 10 shows the synthetic data that are used in the LDA experiment. Each row represents a document and each column represents a word.

The sampling time in seconds for the R8 dataset is listed in Table 4. There are 5485 training documents. The first row in the table shows the sampling time for different minibatch sizes k and different versions of k-DPP sampling. In

Topic 1	pct shares stake and group investment securities stock commission firm
Topic 2	year pct and for last lower growth debt profits
Topic 3	company and merger for will approval companies corp
Topic 5	acquire into letter
Topic 4	and for canadian company management pacific
Topic 5	bid southern court units baker official and that treasury western policy
Tople 5	administration study budget
Topic 6	and president for executive chief shares plc com- pany chairman cyclops
Topic 7	bank pct banks rate rates money interest and
Topic 8	reuter today and unit inc sale sell reuter company systems
Topic 0	corp terms
Topic 9	mln stg and reuter months year for plc market
Topic 10	and national loan federal savings reuter associa-
	tion insurance estate real
Topic 11	trade and for bill not united imports that surplus south
Topic 12	and february for china january gulf issue month
Topic 13	market dollar that had and will exchange system
Topia 14	currency west
Topic 14	and fiscal
Topic 15	billion mln tax year profit credit marks francs
Tomia 16	net pct
Topic To	diluted piedmont
Topic 17	and will union spokesman not two that reuter
Topic 18	offer share tender shares that general and gen-
	corp dlrs not
Торіс 19	and company for that board proposal group made directors proposed
Topic 20	that japan japanese and world industry govern-
Topic 21	ment for told officials american analysts and that analyst chrysler shear-
10010 21	son express stock not
Topic 22	loss profit mln reuter cts net shr dlrs qtr year
Topic 23	mln dlrs and assets for dlr operations year charge
Tania 24	reuler
Topic 24	cts april reuter div pay prior record atly march
Topic 25	sets
Topic 26	dividend stock split for two reuter march payable
Topic 27	oil and prices crude for energy opec petroleum
-	production bpd
Topic 28	agreement for development and years program
Topic 29	and foreign that talks for international industrial
_	exchange not since
Topic 30	corp inc acquisition will company common shares reuter stock purchase
	-

Table 2: Top 10 words for each topics learned from LDA with traditional SVI.

Topic 1	oil and that prices for petroleum dlrs energy
T:- 2	crude field
Topic 2	will
Topic 3	billion and pct mln group marks sales year capi-
	tal rose
Topic 4	and saudi oil gulf that arabia december minister
	prices for
Topic 5	and dlrs debt for brazil southern mln will medi-
Topia 6	cal had
Topic 0	and grain that will futures for program farm cer-
	tificates agriculture
Topic 7	bank banks rate and pct interest rates for foreign
Topic 8	and union for national seamen california port
Topic o	security that strike
Topic 9	and trade that for dollar deficit gatt not exports
-	economic
Topic 10	and financial for sale inc services reuter systems
T	agreement assets
	ling market
Topic 12	and for south unit equipment reuter two will state
	corp
Topic 13	and firm stock company will for pct not share
	that
Topic 14	and world that talks economic official for coun-
Topic 15	and gencorn for offer general company partners
Topic 15	that dlrs share
Topic 16	mln canada canadian stg and pct will air that
	royal
Topic 17	usair and twa that analysts not for pct analyst
Topic 18	and that for companies not years study this areas
Topic To	overseas
Topic 19	trade and bill for house that reagan foreign states
	committee
Topic 20	company dlrs offer stock and for corp share
Topic 21	snares min dire year and quarter company for earnings will
10pic 21	tax share
Topic 22	mln cts net loss dlrs profit reuter shr year qtr
Topic 23	exchange paris and rates that treasury baker al-
	lied for western
Topic 24	and shares inc for group dlrs pct offer reuter
Topic 25	share merger and that pacific texas hughes baker com-
Topic 25	merce for company
Topic 26	and american company subsidiary china french
	reuter pct for owned
Topic 27	Japan japanese and that trade officials for gov-
Topic 28	ernment industry pact oil onec min bnd prices production equador and
	output crude
Topic 29	and that had shares block for mln government
-	not san
Topic 30	mln pct and profits dlrs year for billion company
	W1II

Table 3: Top 10 words for each topics learned from LDA with DM-SVI.



Figure 10: Synthetic data used to train the LDA model in the main paper. Each row presents a document and each column represents a word. Documents share topics with highly imbalanced proportions.

Size	k = 10	k =30	k=50	k=80
Fast k-DPP	0.001	0.0139	0.0541	0.2199
k-DPP	0.0098	0.1468	0.6438	2.6698
LDA	0.8777	1.2530	1.6414	2.2312

Table 4: Sampling time (in sec) for LDA on the R8 dataset with different mini-batch sizes.

practice, we use the original implementation from [23] with M = 100. To compare with the traditional k-DPP, we listed the elapsed time with [19]. The last row shows the running time per local LDA update, excluding sampling.

The computational time for training a neural network highly depends on the network structure and implementation details. For example, when using only one softmax layer as in the flower experiment, the cost per gradient step is in the milliseconds. In this setup, k-DPP is not effective from a runtime perspective, but still results in better final classification accuracies. However, the cost for each gradient step for a simple 5 layer NN as in the MNIST experiment with K = 100 is 1.294 seconds. In the latter case, this time is comparable to k-DPP sampling (0.7941 sec) see Table 5. We thus expect our methods to benefit expensive models and imbalanced training datasets more.

Figure 11 shows the bar plots of the frequency of images in each class for Oxford Flower dataset using the number of classes as the mini-batch size. With this setting, we can see that when w = 1, DM-SGD is equivalent to StS.

Size	k = 10	k =100	k=200
Fast k-DPP	0.0012	0.7941	5.4216
NN cost	0.166948	1.29452	2.64811

Table 5: Five Layer NN trained on MNIST with different minibatch sizes. Top row: sampling time (in sec) using the fast k-DPP approach. Bottom row: run time for each update step (excluding mini-batch sampling).



Figure 11: The frequency of images in each class for Oxford Flower dataset, with k = 102.