Clas s No	Chapter Title/Reference Literature	Topics to be Covered	% of Portions Covered	
			Referenc e Chapter	Cumulativ e
1	Big Data Introduction/T1	Big Data definition, Challenges and opportunities with Big Data	1	3.6
2	Big Data Characterisitics/T1	Data intensive scientific discovery and the role of Big Data, History	2	7.2
3	HDFS/T1	Map Reduce – Storage (HDFS)	2	10.8
4	Map Reduce/T1	Map Reduce – Computation model, Map Reduce architecture,	2,4	14.3
5	Hands on – Map Reduce/T1	Demo class: Map-Reduce – Hands on programming	2,4	17.9
6	YARN/T1	Case Study: Google. YARN introduction.	2	21.5
7	Hadoop Ecosystem/T1	Overview of Hadoop Ecosystem – Oozie, Ambari, Sqoop and Flume	2	25
8	Matrix Vector Multiplication/T1	Introduction to sample Big Data Algorithms – Sparse Matrices, matrix vector multiplication with MR	4	28.6
9	Pagerank/T1	Introduction to sample Big Data Algorithms - Pagerank computations	9	32.2
10	Relational Operators with MR/T1	Relational operators on Map-reduce, Select, Project, Join, Grouping, HIVE	4	35.8
11	Hands On with HIVE/T1	HIVE hands on	4	39.3
12.	Hbase- Cassandra/T1	case study: Other storage - Hbase/Cassandra architecture and columnar storage for analytics	3	42.9
13	Hadoop issues/T1	Issues with Hadoop, Spark and Scala	5	46.5
14	PySpark/T1	PySpark programming model	5	50
15	Spark Programming Model/T1	Transformations and Actions, Spark SQL	5	53.6

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Spark	Spark architecture – RDD, DataFrames,	5	57.2
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Algorithm	Complexity of Big Data algorithms –	2	60.8
Complexity/T3	Communication Cost complexity model.		60.8
Hands On with	Snork HandsOn	г	64.3
Spark/T1	Spark natiusOii	5	04.5
Streaming			
	-	7	67.9
	Streaming Spark,		
Kafka/T1	Kafka – use cases, architecture	7	71.5
Streaming	Streaming Algorithms - Sampling, set		7-
Algorithms 1/T1	membership	/	75
Kafka Hands on/T1	Kafka with HandsOn	7	78.6
	Streaming Algorithms - Bloom Filters,		
Streaming	Counting		02.2
Algorithms 2/T1	9	/	82.2
,	Martin Algorithm.		
ML Algorithms/T1	Clustering Algorithms - kmeans and	_	05.0
	collaborative filtering	р	85.8
MI and Dig Data /T1	Scaling Neural Networks for Big Data,	C	89.3
IVIL dilubig Data/ 11	case study MLLib.		89.3
Project work	Project Work feedback		92.9
Project work	Project Work feedback		96.5
Project work	Project Evaluations		100
	Architecture/T1 Algorithm Complexity/T3 Hands On with Spark/T1 Streaming Spark/T1 Kafka/T1 Streaming Algorithms 1/T1 Kafka Hands on/T1 Streaming Algorithms 2/T1 ML Algorithms/T1 ML andBig Data/T1 Project work Project work	Architecture/T1 Wide and Narrow dependencies, Algorithm Complexity of Big Data algorithms — Complexity/T3 Communication Cost complexity model. Hands On with Spark/T1 Streaming Spark/T1 Streaming Spark, Kafka/T1 Kafka — use cases, architecture Streaming Algorithms - Sampling, set membership Kafka Hands on/T1 Kafka with HandsOn Streaming Algorithms - Bloom Filters, Counting Algorithms 2/T1 Counting unique elements — Flajolet Martin Algorithm. ML Algorithms/T1 Clustering Algorithms - kmeans and collaborative filtering ML andBig Data/T1 Scaling Neural Networks for Big Data, case study MLLib. Project work Project Work feedback Project work Project Work feedback	Architecture/T1 Wide and Narrow dependencies, Algorithm Complexity of Big Data algorithms — Complexity/T3 Communication Cost complexity model. Hands On with Spark/T1 Streaming Spark/T1 Streaming Spark/T1 Streaming Spark, Kafka/T1 Kafka — use cases, architecture 7 Streaming Algorithms - Sampling, set membership 7 Kafka Hands on/T1 Kafka with HandsOn 7 Streaming Algorithms - Bloom Filters, Counting Counting Unique elements — Flajolet Martin Algorithm. ML Algorithms/T1 Clustering Algorithms - kmeans and collaborative filtering 6 ML andBig Data/T1 Scaling Neural Networks for Big Data, case study MLLib. Project work Project Work feedback Project work Project Work feedback