# Comparative Analysis of Path-Level, and Fragmentation-as-a-Picture Methods for Blocking Prediction in Elastic Optical Networks

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Index Terms—Elastic optical networks (EON), Bandwidth fragmentation, Fragmentation metrics, Blocking rate, Machine Learning, Fragmentation Visual Representation, Fragmentation as a picture.

#### I. Notes on Results

This document contains detailed results for the performance of models used to predict blocking of connections at k, k+1, k+3, k+5, k+7, k+9, and k+11 steps for Low, Medium, High, and Very High loads. It is important to note that High and Very high, loads were evaluated only for experimentation purposes in the fragmentation as a picture approach since they cause network blocking to go beyond 20%.

## APPENDIX A COMPLETE TABLES AND FIGURES

Load	Steps	Performance					
Load	Sicps	Accuracy	Precision	5128         0.66125           2720         0.62716           0237         0.60234           0514         0.59511           0174         0.59171           0258         0.59254           0368         0.59365           0167         0.64161           0010         0.60004           03192         0.58186           0568         0.57562           0508         0.57502           0406         0.56400	F1		
	0	0.89651	0.66128	0.66125	0.66127		
	1	0.88816	0.62720	0.62716	0.62718		
	3	0.88434	0.60237	0.60234	0.60236		
Low	5	0.88421	0.59514	0.59511	0.59513		
	7	0.88397	0.59174	0.59171	0.59172		
	9	0.88440	0.59258	0.59254	0.59256		
	11	0.88480	7 0.59174 0.59171 0 0.59258 0.59254 0 0.59368 0.59365 5 0.64167 0.64161	0.59367			
	0	0.92445	0.64167	0.64161	0.64164		
	1	0.91731	0.60010	0.60004	0.60007		
	3	0.91636	0.58192	0.58186	0.58189		
Medium	5	0.91660	0.57568	0.57562	0.57565		
	7	0.91696	0.57508	0.57502	0.57505		
	9	0.91497	0.56406	0.56400	0.56403		
	11	0.91737	0.57628	0.57622	0.57625		

TABLE I: NA - Arithmetic Mean. Path based fragmentation.

Load	Stone		Perforn	nance	
Loau	Steps	Accuracy	Precision	Recall	F1
	0	0.92999	0.77087	0.77083	0.77085
	1	0.92083	0.73612	0.73608	0.73610
	3	0.91814	0.71859	0.71855	0.71857
Low	5	0.91870	0.71573	0.71569	0.71571
	7	0.91830	0.71254	0.71250	0.71252
	9	0.91933	0.71570	0.71566	0.71568
	11	0.91868	0.71319	0.71316	0.71318
	0	0.94781	0.75247	0.75239	0.75243
	1	0.93873	0.70373	0.70366	0.70370
	3	0.93858	0.69301	0.69293	0.69297
Medium	5	0.93625	0.67565	0.67558	0.67562
	7	0.93809	0.68321	0.68313	0.68317
	9	0.93888	0.68668	0.68661	0.68664
	11	0.93648	0.67429	0.67422	0.67426

TABLE II: NA - Sum. Path based fragmentation.

			Doufoun	*****	
Load	Steps				
Zoud	Втерь	Accuracy	Precision	Recall	F1
	0	0.93667	0.79273	0.79269	0.79271
	1	0.92544	0.75148	0.75144	0.75146
	3	0.92249	0.73353	0.73349	0.73351
Low	5	0.92382	0.73363	0.73359	0.73361
	7	0.92414	0.73310	0.73306	0.73308
	9	0.92413	0.73260	0.73256	0.73258
	11	0.92433	0.73310	0.75144 0.73349 0.73359 0.73306	0.73308
	0	0.95122	0.76864	0.76857	0.76860
	1	0.94161	0.71763	0.71756	0.71760
	3	0.93996	0.69987	0.69980	0.69983
Medium	5	0.94137	0.70174	0.70166	0.70170
	7	0.94156	0.70099	0.70091	0.70095
	9	0.94129	0.69900	0.69892	0.69896
	11	0.94234	0.70432	0.70425	0.70428

TABLE III: NA - Sum, Kurtosis, Standard Deviation and Skew. Path based fragmentation.

Load	Steps		Performance           Precision         Recall           0.80721         0.80717           0.76968         0.76964	nance	
Loau	Steps	Accuracy	Precision	Recall	F1
	0	0.94109	0.80721	0.80717	0.80719
	1	0.93090	0.76968	0.76964	0.76966
	3	0.92826	0.75336	0.75332	0.75334
Low	5	0.92824	0.74909	0.74905	0.74907
	7	0.92773	0.74573	0.74569	0.74571
	9	0.92864	0.74851	0.74847	0.74849
	11	0.92884	0.74903	0.74899	0.74901
	0	0.95598	0.79123	0.79115	0.79119
	1	0.94555	0.73672	0.73665	0.73668
	3	0.94474	0.72378	0.72371	0.72374
Medium	5	0.94573	0.72389	0.72381	0.72385
	7	0.94624	0.72492	0.72484	0.72488
	9	0.94598	0.72308	0.72300	0.72304
	11	0.94607	0.72346	0.72338	0.72342

TABLE IV: NA - Granularity Weighted Sum. Path based fragmentation.

Load	Steps		Perforn	nance	
Load	Sicps	Accuracy	Precision	Recall	F1
	0	0.94540	0.82133	0.82129	0.82131
	1	0.93272	0.77576	0.77572	0.77574
	3	0.92962	0.75804	0.75800	0.75802
Low	5	0.93008	0.75553	0.75549	0.75551
	7	0.93197	0.76064	0.76060	0.76062
	9	0.93146	0.75843	0.75839	0.75841
	11	0.93208	0.76043	0.76039	0.76041
	0	0.95930	0.80700	0.80692	0.80696
	1	0.94635	0.74056	0.74048	0.74052
	3	0.94643	0.73225	0.73218	0.73221
Medium	5	0.94697	0.73022	0.73014	0.73018
	7	0.94757	0.73172	0.73164	0.73168
	9	0.94791	0.73298	0.73290	0.73294
	11	0.94768	0.73171	0.73163	0.73167

TABLE V: NA - Granularity Weighted Sum, Kurtosis, Standard Deviation and Skew. Path based fragmentation.

			Porforn	nance	
Load	Steps	Accuracy	Precision		F1
	0	0.94417	0.81444		0.81443
	1	0.93401	0.77936		0.77936
	3	0.93266	0.77288	0.77286	0.77287
Low	5	0.93235	0.77067	0.77066	0.77067
	7	0.93292	0.77198	0.77196	0.77197
	9	0.93197	0.76828	0.76826	0.76827
	11	0.93154	0.76650	0.77066 0.77196	0.76649
	0	0.95044	0.80418	0.80417	0.80417
Ì	1	0.94080	0.76467	0.76465	0.76466
	3	0.93858	0.75349	0.75347	0.75348
Medium	5	0.93849	0.75155	0.75153	0.75154
	7	0.93813	0.74897	011 107 0	0.74896
	9	0.93852	0.74968	011 12 01	0.74968
	11	0.93814	0.74739	0.74737	0.74738

TABLE VIII: Germany - Sum, Kurtosis, Standard Deviation, and Skew. Path based fragmentation.

Load	Steps		Perforn	nance	
Load	Sicps	Accuracy	Precision	Recall	F1
	0	0.91229	0.70850	0.70849	0.70850
	1	0.90605	0.68588	0.68587	0.68587
	3	0.90428	0.67713	0.67712	0.67712
Low	5	0.90381	0.67392	0.67391	0.67392
	7	0.90363	0.67240	0.67238	0.67239
	9	0.90262	0.66827	0.66825	0.66826
	11	0.90281	0.66848	Recall 0.70849 0.68587 0.67712 0.67391 0.67238	0.66848
	0	0.91658	0.67040	0.67039	0.67040
	1	0.91042	0.64391	0.64390	0.64390
	3	0.90949	0.63671	0.63670	0.63671
Medium	5	0.90846	0.63022	0.63021	0.63022
	7	0.90890	0.63034	0.63033	0.63034
	9	0.90790	0.62500	0.62499	0.62500
	11	0.90729	0.62139	0.62138	0.62139

TABLE VI: Germany - Arithmetic Mean. Path based fragmentation.

			Perforn	nance	
Load	Steps	Accuracy	Precision	Recall	F1
	0	0.94906	0.83069	0.83068	0.83069
	1	0.93947	0.79764	0.79762	0.79763
	3	0.93788	0.79046	0.79044	0.79045
Low	5	0.93787	0.78937	0.78936	0.78937
	7	0.93685	0.78534	0.78532	0.78533
	9	0.93647	0.78358	0.78357	0.78358
	11	0.93646	0.78326	069 0.83068 764 0.79762 046 0.79044 037 0.78936 034 0.78532 038 0.78357 0326 0.78324 038 0.82817 005 0.78204 043 0.77045 043 0.77042 060 0.76659 017 0.76616	0.78325
	0	0.95651	0.82818	0.82817	0.82817
	1	0.94517	0.78205	0.78204	0.78205
	3	0.94384	0.77458	0.77456	0.77457
Medium	5	0.94317	0.77043	0.77042	0.77042
	7	0.94248	0.76660	0.76659	0.76660
	9	0.94257	0.76617	0.76616	0.76617
	11	0.94211	0.76358	0.76357	0.76357

TABLE IX: Germany - Granularity Weighted Sum. Path based fragmentation.

Load	Steps		Perforn	nance	
Load	ысра	Accuracy	Precision	Recall	F1
	0	0.93474	0.78310	0.78309	0.78309
	1	0.92591	0.75229	0.75228	0.75228
	3	0.92534	0.74818	0.74816	0.74817
Low	5	0.92435	0.74356	0.74355	0.74356
	7	0.92400	0.74165	0.74164	0.74165
	9	0.92347	0.73932	0.73931	0.73931
	11	0.92311	0.73773	Recall 0.78309 0.75228 0.74816 0.74355 0.74164	0.73772
	0	0.94346	0.77663	0.77661	0.77662
	1	0.93517	0.74230	0.74229	0.74229
	3	0.93310	0.73147	0.73146	0.73146
Medium	5	0.93216	0.72599	0.72598	0.72598
	7	0.93242	0.72577	0.72576	0.72577
	9	0.93202	0.72320	0.72319	0.72319
	11	0.93150	0.72024	0.72023	0.72024

TABLE VII: Germany - Sum. Path based fragmentation.

Load	Steps		Perforn	nance	
Loau	Sieps	Accuracy	Precision	Recall	F1
	0	0.95625	0.85459	0.85458	0.85458
	1	0.94489	0.81575	0.81573	0.81574
	3	0.94310	0.80809	0.80808	0.80809
Low	5	0.94245	0.80492	0.80491	0.80492
	7	0.94282	0.80563	0.80561	0.80562
	9	0.94178	0.80169	0.80168	0.80169
	11	0.94241	0.80358	0.80356	0.80357
	0	0.96142	0.84759	0.84758	0.84759
	1	0.95007	0.80153	0.80152	0.80153
	3	0.94772	0.79018	0.79016	0.79017
Medium	5	0.94710	0.78634	0.78633	0.78633
	7	0.94706	0.78517	0.78516	0.78517
	9	0.94659	0.78254	0.78252	0.78253
	11	0.94697	0.78343	0.78342	0.78342

TABLE X: Germany - Granularity Weighted Sum, Kurtosis, Standard Deviation, and Skew. Path based fragmentation.

Load	Steps		Perforn	nance	
Load	Sicps	Accuracy	Precision	Recall	F1
	0	0.93960	0.93534	0.93634	0.93584
	1	0.84232	0.80627	0.87105	0.83741
	3	0.84063	0.80833	0.85564	0.83131
Low	5	0.84286	0.80814	0.85748	0.83208
	7	0.83754	0.79777	0.85811	0.82684
	9	0.84636	0.80851	0.86414	0.83540
	11	0.84724	0.81248	0.85935	0.83526
	0	0.96511	0.91694	0.91870	0.91782
	1	0.90586	0.77076	0.77967	0.77519
	3	0.90341	0.74603	0.79038	0.76757
Medium	5	0.90466	0.74280	0.79501	0.76801
	7	0.90451	0.74090	0.79398	0.76652
	9	0.90530	0.74504	0.78994	0.76684
	11	0.90518	0.74476	0.78984	0.76664
	0	0.97565	0.91425	0.91622	0.91523
	1	0.93018	0.74165	0.77223	0.75663
	3	0.93051	0.74030	0.75304	0.74662
High	5	0.93229	0.73556	0.77067	0.75271
	7	0.93042	0.71514	0.79253	0.75185
	9	0.93406	0.74986	0.75615	0.75299
	11	0.93403	0.74442	0.76714	0.75561
	0	0.98540	0.93899	0.91472	0.92670
	1	0.95240	0.74648	0.78484	0.76518
	3	0.95347	0.74460	0.78133	0.76252
Very High	5	0.95618	0.77298	0.75768	0.76526
	7	0.95499	0.76152	0.75758	0.75954
	9	0.95633	0.76903	0.76341	0.76621
	11	0.95576	0.76215	0.76733	0.76473

TABLE XI: CNN-LSTM Model Results for NA Topology. Fragmentation as a Picture.

Load	Steps		Perforn	nance	
Load	Steps	Accuracy	Precision	Recall	F1
	0	0.95757	0.95038	0.95455	0.95246
	1	0.88649	0.87130	0.87294	0.87212
	3	0.88493	0.86635	0.87352	0.86992
Low	5	0.88735	0.87713	0.86439	0.87071
	7	0.88425	0.86437	0.87251	0.86842
	9	0.88498	0.86583	0.87211	0.86896
	11	0.88646	0.87588	0.86227	0.86903
	0	0.97075	0.93013	0.94625	0.93812
	1	0.91915	0.82789	0.82430	0.82609
	3	0.91598	0.81381	0.82475	0.81924
Medium	5	0.91602	0.80951	0.82926	0.81927
	7	0.91600	0.81403	0.81984	0.81692
	9	0.91509	0.80592	0.82638	0.81602
	11	0.91630	0.81193	Recall 0.95455 0.87294 0.87352 0.86439 0.87251 0.87211 0.86227 0.94625 0.82430 0.82475 0.82926 0.81984	0.81710
	0	0.97944	0.92274	0.94040	0.93149
	1	0.94033	0.81032	0.77788	0.79377
	3	0.93547	0.77178	0.79250	0.78200
High	5	0.93437	0.75559	0.80935	0.78155
	7	0.93634	0.78180	0.77576	0.77877
	9	0.93606	0.76986	0.79284	0.78119
	11	0.93630	0.76903	0.79528	0.78193
	0	0.98277	0.91369	0.94364	0.92842
	1	0.94907	0.78390	0.78241	0.78315
	3	0.94173	0.72392		0.76279
Very High	5	0.94488	0.76037		0.76148
	7	0.94638	0.76717		0.76626
	9	0.94566	0.75769		0.76483
	11	0.94415	0.74031	0.78673	0.76282

TABLE XIII: CNN-LSTM Model Results for German Topology. Fragmentation as a Picture

Load	Ctoma	Performance			
Loau	Steps	Accuracy	Precision	Recall	F1
	0	0.89537	0.89390	0.88234	0.88808
	1	0.85763	0.84751	0.84702	0.84726
	3	0.85311	0.84167	0.83749	0.83957
Low	5	0.85280	0.83963	0.83538	0.83750
	7	0.85000	0.83699	0.82974	0.83335
	9	0.85420	0.83865	0.83812	0.83838
	11	0.85173	0.83632	0.83425	0.83528
	0	0.93408	0.84889	0.83845	0.84364
	1	0.91434	0.79436	0.79404	0.79420
	3	0.91283	0.78414	0.78377	0.78395
Medium	5	0.91222	0.78042	0.77622	0.77832
	7	0.91320	0.78312	0.77499	0.77903
	9	0.91292	0.78064	0.77645	0.77854
	11	0.91335	0.78033	0.78025	0.78029
	0	0.94778	0.82219	0.81156	0.81684
	1	0.93597	0.77475	0.76758	0.77115
	3	0.93538	0.76439	0.75842	0.76140
High	5	0.93555	0.75946	0.75806	0.75876
	7	0.93717	0.76388	0.76366	0.76377
	9	0.93677	0.76216	0.76212	0.76214
	11	0.93750	0.76845	0.75840	0.76339
	0	0.96461	0.82678	0.82139	0.82407
Very High	1	0.95421	0.76963	0.76580	0.76771
	3	0.95475	0.76360	0.76287	0.76324
	5	0.95606	0.76924	0.76256	0.76588
	7	0.95554	0.76388	0.76162	0.76275
	9	0.95528	0.76149	0.76136	0.76143
	11	0.95546	0.76342	0.76037	0.76189

TABLE XII: Random Forest Results for NA Topology. Fragmentation as a Picture.

Load	Steps	Performance			
Load		Accuracy	Precision	Recall	F1
	0	0.90898	0.89979	0.89531	0.89754
	1	0.88277	0.86943	0.86561	0.86751
	3	0.88230	0.86860	0.86340	0.86599
Low	5	0.88163	0.86774	0.86158	0.86465
	7	0.88030	0.86416	0.86209	0.86313
	9	0.87995	0.86468	0.86007	0.86237
	11	0.88011	0.86368	0.86153	0.86260
	0	0.93764	0.86875	0.86453	0.86664
	1	0.91563	0.81893	0.81891	0.81892
	3	0.91697	0.82019	0.82017	0.82018
Medium	5	0.91678	0.81892	0.81844	0.81868
	7	0.91666	0.81864	0.81626	0.81745
	9	0.91580	0.81613	0.81385	0.81499
	11	0.91604	0.81702	0.81274	0.81487
	0	0.95374	0.84438	0.84435	0.84437
	1	0.93763	0.78967	0.78714	0.78841
	3	0.93853	0.78958	0.78955	0.78956
High	5	0.93875	0.78892	0.78881	0.78887
	7	0.93882	0.79015	0.78483	0.78748
	9	0.93842	0.78612	0.78611	0.78611
	11	0.93843	0.78563	0.78558	0.78561
	0	0.96151	0.83846	0.83598	0.83722
Very High	1	0.94705	0.77636	0.77189	0.77412
	3	0.94814	0.77760	0.77565	0.77662
	5	0.94801	0.77523	0.77370	0.77447
	7	0.94727	0.77055	0.77009	0.77032
	9	0.94815	0.77488	0.77094	0.77290
	11	0.94802	0.77231	0.77229	0.77230

TABLE XIV: Random Forest Results for German Topology. Fragmentation as a Picture.

Load	Steps	Performance				
Loau	Steps	Accuracy	Precision	Recall	F1	
	0	0.94655	0.93944	0.94109	0.94026	
	1	0.85931	0.82808	0.86071	0.84408	
	3	0.85866	0.83000	0.84850	0.83915	
Low	5	0.85995	0.82019	0.86339	0.84124	
	7	0.86243	0.84073	0.83636	0.83854	
	9	0.85701	0.82892	0.83668	0.83278	
	11	0.85488	0.81022	0.85960	0.83418	
	0	0.97095	0.93505	0.92678	0.93090	
	1	0.91602	0.79581	0.79995	0.79787	
	3	0.91607	0.78616	0.79970	0.79287	
Medium	5	0.91463	0.76966	0.81055	0.78958	
	7	0.91776	0.79151	0.78910	0.79031	
	9	0.91496	0.77965	0.78891	0.78426	
	11	0.91713	0.78341	0.79696	0.79012	
	0	0.98151	0.93111	0.94078	0.93592	
	1	0.93847	0.76993	0.80209	0.78568	
	3	0.93986	0.77909	0.77838	0.77873	
High	5	0.94078	0.78031	0.77547	0.77788	
	7	0.94063	0.77359	0.78253	0.77803	
	9	0.94046	0.77541	0.77667	0.77604	
	11	0.94170	0.78493	0.77308	0.77896	
	0	0.98673	0.93404	0.93453	0.93428	
	1	0.95437	0.76270	0.78127	0.77187	
Very High	3	0.95448	0.75204	0.78160	0.76654	
	5	0.95557	0.75712	0.77833	0.76757	
	7	0.95541	0.74860	0.79021	0.76884	
	9	0.95532	0.74870	0.78776	0.76773	
	11	0.95617	0.76025	0.77742	0.76874	

TABLE XV: CNN-LSTM Model Results for NA Topology using KSP Weighted Pictures. Fragmentation as a Picture.

Load	Steps	Performance				
Loau	Steps	Accuracy	Precision	Recall	F1	
	0	0.96139	0.95523	0.95821	0.95672	
	1	0.88843	0.87103	0.87846	0.87473	
	3	0.88929	0.87515	0.87325	0.87420	
Low	5	0.88898	0.87896	0.86629	0.87258	
	7	0.88800	0.86324	0.88426	0.87362	
	9	0.88646	0.86457	0.87787	0.87117	
	11	0.88672	0.86927	0.87177	0.87052	
	0	0.97634	0.94170	0.95838	0.94997	
	1	0.92042	0.83654	0.81829	0.82732	
	3	0.91650	0.81999	0.81783	0.81891	
Medium	5	0.91655	0.81519	0.82303	0.81909	
	7	0.91498	0.80494	0.82892	0.81676	
	9	0.91706	0.82820	0.80247	0.81513	
	11	0.91720	0.82433	0.80796	0.81607	
	0	0.98302	0.93320	0.95402	0.94350	
	1	0.94120	0.80335	0.79667	0.80000	
	3	0.93537	0.76691	0.80092	0.78354	
High	5	0.93557	0.77767	0.77841	0.77804	
	7	0.93441	0.75952	0.79874	0.77863	
	9	0.93410	0.75517	0.80233	0.77804	
	11	0.93668	0.77777	0.78264	0.78020	
	0	0.98429	0.91663	0.95412	0.93500	
Very High	1	0.94901	0.78444	0.78085	0.78264	
	3	0.94518	0.76563	0.76141	0.76351	
	5	0.94541	0.76206	0.76600	0.76403	
	7	0.94463	0.75039	0.77588	0.76292	
	9	0.94591	0.76523	0.76077	0.76300	
	11	0.94534	0.75458	0.77239	0.76338	

TABLE XVII: CNN-LSTM Model Results for German Topology using KSP Weighted Pictures. Fragmentation as a Picture

Load	Steps	Performance			
	Sieps	Accuracy	Precision	Recall	F1
	0	0.91279	0.90336	0.90133	0.90234
	1	0.88144	0.86820	0.86306	0.86562
	3	0.87965	0.86304	0.85940	0.86121
Low	5	0.88266	0.86626	0.85968	0.86295
	7	0.88227	0.86392	0.85979	0.86185
	9	0.88227	0.86354	0.85910	0.86131
	11	0.88230	0.86225	0.86028	0.86126
	0	0.94249	0.86474	0.86255	0.86365
	1	0.92580	0.82380	0.81654	0.82016
	3	0.92800	0.82083	0.82070	0.82076
Medium	5	0.92859	0.81934	0.81925	0.81929
	7	0.92957	0.82073	0.82064	0.82068
	9	0.92882	0.81890	0.81747	0.81819
	11	0.92828	0.81718	0.81616	0.81667
	0	0.95827	0.85747	0.85073	0.85409
	1	0.94535	0.80798	0.80192	0.80494
	3	0.94545	0.80080	0.79704	0.79891
High	5	0.94752	0.80559	0.80082	0.80319
	7	0.94741	0.80228	0.80219	0.80223
	9	0.94679	0.79973	0.79964	0.79969
	11	0.94762	0.80300	0.80281	0.80290
	0	0.96722	0.83958	0.83468	0.83712
	1	0.95781	0.78651	0.78646	0.78649
Very High	3	0.95903	0.78624	0.78479	0.78552
	5	0.96033	0.79180	0.78570	0.78874
	7	0.96010	0.78867	0.78526	0.78696
	9	0.96011	0.78727	0.78714	0.78720
	11	0.96017	0.78899	0.78490	0.78694

TABLE XVI: Random Forest Classifier results for NA Topology using KSP Weighed Pictures. Fragmentation as a Picture

Load	Steps	Performance				
Loau	steps	Accuracy	Precision	Recall	F1	
	0	0.91470	0.90616	0.90182	0.90398	
	1	0.88876	0.87650	0.87199	0.87424	
	3	0.88985	0.87547	0.87430	0.87488	
Low	5	0.88946	0.87520	0.87252	0.87386	
	7	0.88853	0.87322	0.87199	0.87260	
	9	0.88843	0.87363	0.87084	0.87223	
	11	0.88905	0.87458	0.87090	0.87273	
	0	0.93861	0.86932	0.86861	0.86897	
	1	0.91817	0.82437	0.82436	0.82437	
	3	0.91972	0.82614	0.82609	0.82611	
Medium	5	0.91972	0.82529	0.82489	0.82509	
	7	0.91912	0.82440	0.82103	0.82271	
	9	0.91876	0.82264	0.82031	0.82148	
	11	0.91924	0.82433	0.81938	0.82185	
	0	0.95445	0.84781	0.84524	0.84653	
	1	0.93845	0.79242	0.78997	0.79119	
	3	0.93998	0.79570	0.79259	0.79414	
High	5	0.93953	0.79161	0.79154	0.79158	
	7	0.93982	0.79260	0.79011	0.79135	
	9	0.93976	0.79218	0.78839	0.79028	
	11	0.93968	0.79071	0.78872	0.78971	
	0	0.96162	0.84014	0.83465	0.83738	
	1	0.94750	0.77746	0.77532	0.77639	
	3	0.94868	0.77922	0.77914	0.77918	
Very High	5	0.94886	0.77844	0.77817	0.77831	
	7	0.94809	0.77414	0.77366	0.77390	
	9	0.94897	0.77782	0.77562	0.77672	
	11	0.94880	0.77680	0.77385	0.77532	

TABLE XVIII: Random Forest Classifier Model Results for German Topology using KSP Weighted Pictures. Fragmentation as a Picture.