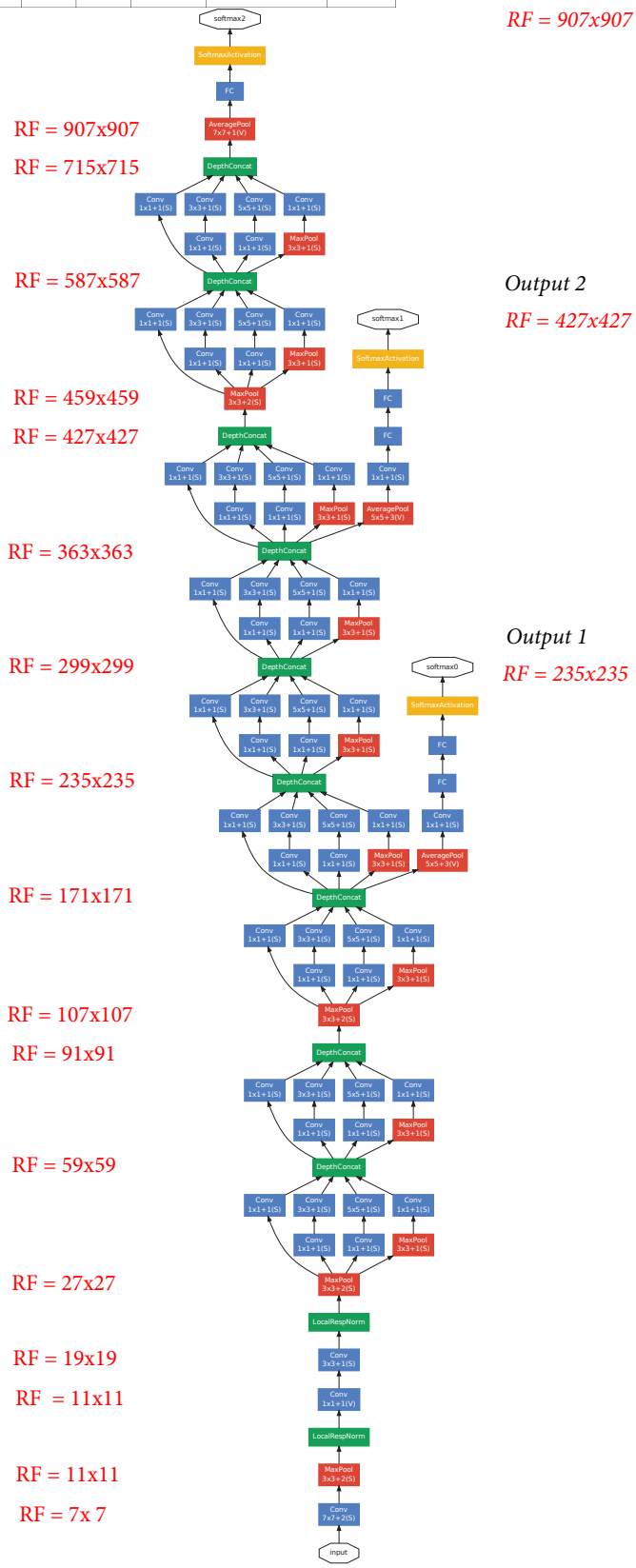
Layers	N-in	k	p	s	N-out	J-in	J-out	R-in	R-out	comments	
Conv	224	7	3	2	112	1	2	1	**7**		
MaxPool	112	3	1	2	56	2	4	7	**11**		
Conv	56	3	1	1	56	4	4	11	**19**		
Maxpool	56	3	1	2	28	4	8	19	**27**		
Inception-3a	28	5	2	1	28	8	8	27	**59**	Max RF taking forward	
Inception-3b	28	5	2	1	28	8	8	59	**91**	Max RF taking forward	
MaxPool	28	3	1	2	14	8	16	91	**107**		
Inception-4a	14	5	2	1	14	16	16	107	**171**	Max RF taking forward	
Inception-4b	14	5	2	1	14	16	16	171	**235**	Max RF taking forward	Output 1
Inception-4c	14	5	2	1	14	16	16	235	**299**	Max RF taking forward	
Inception-4d	14	5	2	1	14	16	16	299	**363**	Max RF taking forward	
Inception-4e	14	5	2	1	14	16	16	363	**427**	Max RF taking forward	Output 2
MaxPool	14	3	1	2	7	16	32	427	**459**		
Inception-5a	7	5	2	1	7	32	32	459	**587**	Max RF taking forward	
Inception-5b	7	5	2	1	7	32	32	587	**715**	Max RF taking forward	
AveragePool	7	7	0	1	1	32	32	715	**907**		Output 3
Softmax	1										



This model has 3 outputs to overcome the vanishing gradient problem, RF at each of these outputs are mentioned.
For the concatenate layers the maximum RF in the parallel paths is carried forward for calculations.

Figure 3: GoogLeNet network with all the bells and whistles