Let's begin by going over some of the nomenclature from the generalized S-W paper to make sure we've defined all our terms and extensions very clearly:

Variable	Range/Conditions	Name	Description
$I_t$		RGB-D Image at time $t$	This represents the stream of images to be segmented
G < V, E >	$V = v_1, v_2, \dots, v_N$	A Graph containing vertices $V$ and edges $E$	This is the general structure we will use to encode adjacency relationships within an Image $I$
$\pi_n = (V_1, V_2, \dots, V_n)$	$\bigcup_{i=1}^{n} V_i = V$ $V_i \cap V_j = \emptyset, \forall i \neq j$	A partition with $n$ segments	A partition is an assignment of each edge $E$ in graph $G'$ to either "on" or "off". An atomic region must exist uniquely in a segment.
$l \in L$	L={planar, cylindrical, spherical, spline, etc.}	The model family for a region	The model family refers to what sort of surface we think a region contains, e.g. plane, sphere, spline, etc. For the time being, this is assumed to be planar for all regions.
$ heta_i$	$i=1,2,\ldots,n$	The model parameters for a region	These are the coefficients for the given model type. For the plane model, these are the coefficients from the plane equation $ax + by + cz = d$ .
$c_i = (l_i, \theta_i)$	$i=1,2,\ldots,n$	The full model for a region	Combination of family and parameters. Eventually, we would like to explore these variables as well, using Data-Driven MCMC
$W = (n, \pi_n, c_1, c_2, \dots, c_n)$		The full world description	This is the function we are trying to optimize via $W^* = \operatorname{argmax}_{W \in \Omega} P(I W)P(W)$ . For our problem, the search space $\Omega$ is very large, with $\Omega = \bigcup_{n=1}^{ V } \{\Omega_{\pi_n} \times \Omega_l^n \times \Omega_{\theta_1} \times \Omega_{\theta_2}, \ldots, \times \Omega_{\theta_n}\}$
$G_t$		The adjacency graph for the image at time $t$	
$G_t'$		The world graph incorporating all images up to time $t$	
$V_t' = V_{t-1}' \bigcup V_t$		Set of vertices at time $t$	
$E'_t = E'_{t-1} \bigcup E_t \bigcup E(\rho'_s, rho_k)$	$s = 1, 2, \dots, \sum_{\tau=1}^{t-1} N_{\tau}$ $K = 1, 2, \dots, N_t$	Set of edges at time $t$	
$q_e = P(\mu_e = on c_i, c_j)$	$e \in E$	Discriminative probability on edge $e$ .	$\mu_e$ is a binary random variable following a Bernoulli distribution and describing the probability of two adjacent vertices in $G_0'(t)$ being connected in a given step.
$E_t'(\pi) \subset E_t'$		Set of edges turned on for partition $\pi$ at time $t$	

Segmentation pseudocode Build adjacency graph based on  $\rho$  heuristics.