Note: In the missing theorem, $g(M_1, M_2, ...)$ = $c(j) + \sum_{n=1}^{\infty} \beta^n c(M_n)$. By stationarily, the conditional distribution of $g(X_2, X_3, ...)$ given $X_i = j$ equals that of $g(X_i, X_2, ...)$ given $X_0 = j$, the conditional expectations are also equal.

The "Missing Theorem"

Let X_0, X_1, \ldots be a stationary Markov chain. Then for any function g,

$$Pr\{g(X_n, X_{n+1}, \ldots) = x | X_0 = i_0, X_1 = i_1, \ldots, X_{n-1} = i\} = Pr\{g(X_n, X_{n+1}, \ldots) = x | X_{n-1} = i\}.$$

Because the Markov chain is stationary, this probability does not depend on n. Also, for your entertainment the function has to be measurable, but of course I did not say that.

Our text actually assumes this in quite a few places without saying so explicitly. For example, time until "eventual" absorbtion is definitely a function of an infinite number of values in the future. Is it so obvious that you can say $Pr\{X_T = 0 | X_0 = i_0, X_1 = i_1, \ldots, X_{n-1} = i\} = Pr\{X_T = 0 | X_{n-1} = i\}$ by the Markov property as stated in our book? Not to me.