H2 Further Mathematics Contents



|  | - Arc length: $\int \sqrt{d x^{2}+d y^{2}}$ (Pythagoras Theorem) <br> - Arc length of polar curve: $\int \sqrt{\mathrm{dr}^{2}+(r d \theta)^{2}}$ <br> - Surface area: $\int 2 \pi y \sqrt{\mathrm{dx}^{2}+\mathrm{dy}^{2}}$ (perimeter $\times$ thickness) |
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| Further Differential Equation | - Integration factor: For $\frac{d y}{d x}+P(x) y=Q(x)$, multiply both sides by $\mathrm{I}(\mathrm{x})=\mathrm{e}^{\int \mathrm{P}(\mathrm{x}) \mathrm{dx}}$ $\begin{gathered} \mathrm{I}(\mathrm{x}) \frac{d y}{d x}+I(x) P(x) y=Q(x) I(x) \\ \mathrm{I}(\mathrm{x}) \frac{d y}{d x}+I^{\prime}(x) y=Q(x) I(x) \\ \frac{d}{d x}(I(x) y)=Q(x) I(x) \\ \mathrm{y}=\frac{\int Q(x) I(x) d x}{I(x)} \end{gathered}$ |
|  | - Solution to second-order differential equation: $\mathrm{a} \frac{d^{2} y}{d x^{2}}+b \frac{d y}{d x}+c=0$ |
|  | $\Delta>0 \quad y=A e^{m_{1} \mathrm{x}}+\mathrm{Be}^{\mathrm{m}_{2} \mathrm{x}}$ |
|  | $\Delta=0 \quad y=(A+B x) \mathrm{e}^{\mathrm{mx}}$ |
|  | $\Delta<0 \quad \mathrm{y}=e^{\alpha x}(A \cos \beta x+B \sin \beta x)$ |
|  | - Euler's method: $\mathrm{y}_{\mathrm{n}+1}=\mathrm{y}_{\mathrm{n}}+\Delta \mathrm{t} \cdot \mathrm{f}\left(\mathrm{t}_{\mathrm{n}}, \mathrm{y}_{\mathrm{n}}\right)$ <br> - Improved Euler's method: $\left\{\begin{array}{l}y_{n+1}^{*}=y_{n}+\Delta t \cdot f\left(t_{n}, y_{n}\right) \\ y_{n+1}=y_{n}+\Delta t \cdot\left[\frac{f\left(t_{n}, y_{n}\right)+f\left(t_{n+1}, y_{n+1}^{*}\right)}{2}\right]\end{array}\right.$ |
| Further <br> Complex <br> Number | - de Moivre's Theorem: $[\mathrm{r}(\cos \theta+\mathrm{i} \sin \theta)]^{\mathrm{n}}=\mathrm{r}^{\mathrm{n}}(\operatorname{cosn} \theta+\mathrm{i} \operatorname{sinn} \theta)$ <br> - Roots of $z^{n}=r e^{i \theta}: z=r^{\frac{1}{n}} e^{\frac{\theta+2 k \pi}{n}}$ |
| Further <br> Special <br> Discrete <br> Probability <br> Distributions | - Conditions for Poisson Distribution: Independence, constant mean rate, randomness <br> - If $\mathrm{X} \sim \operatorname{Po}(\lambda), \mathrm{E}(\mathrm{X})=\lambda, \operatorname{Var}(\mathrm{X})=\lambda$ <br> - Conditions for geometric distribution: Independence, two outcomes, constant probability of success <br> - Memoryless property: If $\mathrm{X} \sim \mathrm{Geo}(\mathrm{p}), \mathrm{P}(\mathrm{X}>\mathrm{m}+\mathrm{n} \mid \mathrm{X}>\mathrm{n})=$ $\mathrm{P}(\mathrm{X}>\mathrm{m})$ |


| Continuous <br> Random <br> Variables | - $\int_{-\infty}^{+\infty} f(x) d x=1$ <br> - Cumulative density function: $F(x)=P(X \leq x)=\int_{-\infty}^{x} f(t) d t$ <br> - Median, $m: F(m)=1$ <br> - Mode, $m^{*}: f\left(m^{*}\right) \geq f(x)$ for all $x$ <br> - $E(X)=\int_{-\infty}^{+\infty} x f(x) d x$ <br> - $E(g(X))=\int_{-\infty}^{+\infty} g(x) f(x) d x$ <br> - $P(X \leq x)=P(Y \leq f(x))$ |
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| Special <br> Continuous Probability Distributions | - Uniform distribution: $E(X)=\frac{a+b}{2}, \operatorname{Var}(X)=\frac{1}{12}(b-a)^{2}$ |
| Confidence Intervals | - Unbiased estimator for $\mu: \overline{\mathrm{X}}$ <br> - Unbiased estimator for $\sigma^{2}: S^{2}=\frac{\mathrm{n}}{\mathrm{n}-1} \sigma_{\mathrm{n}}{ }^{2}$ <br> - Unbiased estimator for $\mathrm{p}: \mathrm{P}_{\mathrm{S}}=\frac{\mathrm{x}}{\mathrm{n}}$ <br> - Explain in context a confidence interval: <br> If a large number of samples are taken randomly, and their (1- $\alpha$ ) confidence intervals for $\mu$ are found, then we say that (1- $\alpha$ ) of such intervals will contain $\mu$. <br> - Confidence interval for population mean: <br> - Confidence interval for population proportion: $\left(P_{s}-z_{1-\frac{\alpha}{2}} \sqrt{\frac{p(1-p)}{n}}, P_{s}+z_{1-\frac{\alpha}{2}} \sqrt{\left.\frac{p(1-p)}{n}\right)}\right.$ |
| Further Hypothesis | - $t$-test is used when population is normally distributed and sample size is small. |


| Testing | - Presentation: <br> Test $H_{0}$ against $H_{1}$ (state both hypotheses). <br> Level of significance: $\alpha \%$ (lower/upper/two-tailed). <br> Under $H_{0}, \bar{X} \sim N\left(\mu, \frac{\sigma^{2}}{n}\right)$, hence $\mathrm{Z}=\frac{\bar{X}-\mu}{\sqrt{\sigma^{2} / n}} \sim N(0,1)$ <br> Method 1 <br> Critical region: $\mathrm{z}<\mathrm{a}$ (taking lower-tailed as an example). <br> Observed test statistic: $\mathrm{z}=\frac{\bar{x}-\mu}{\sqrt{\sigma^{2} / n}}>a$ <br> Hence we do not reject $H_{0}$. <br> Method 2 <br> By using GC, $p$-value $=p$ <br> Since $p>\alpha \%$, we do not reject $H_{0}$. <br> Therefore, there is insufficient evidence at $\alpha \%$ level of significance to claim $H_{1}$ (stated). |
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|  | - When using CLT, remember to convert sample variance to unbiased estimate of population variance. <br> - Assumptions for $t$-test: Normal distribution, randomness, independence. <br> - Value of population mean specified by the null hypothesis is contained in the $\alpha \%$ confidence interval. $\leftrightarrow$ We do not reject the null hypothesis at a $(1-\alpha \%)$ level of significance. <br> - Two samples test: Under $H_{0}, \mathrm{Z}=\frac{\left(\overline{X_{1}}-\overline{X_{2}}\right)-\left(\mu_{1}-\mu_{2}\right)}{\sqrt{\frac{\sigma_{1}{ }^{2}}{n_{1}}+\frac{\sigma_{2}{ }^{2}}{n_{2}}}} \sim \mathrm{~N}(0,1)$ <br> - For two samples test, $X_{1}$ and $X_{2}$ are drawn independently. <br> - Unbiased estimate for common variance: $s=\frac{n_{1} \sigma_{s_{1}}{ }^{2}+n_{2} \sigma_{s 2}{ }^{2}}{n_{1}+n_{2}-2}$ <br> - $t$-test is used when population is normally distributed and at least one sample size is small. <br> - Paired sample $t$-test is used when each subject is measured twice (e.g. before and after tutorials). In this case, under $H_{0}, \quad \mathrm{~T}=$ $\frac{\bar{D}-d_{0}}{S_{D} / \sqrt{n}} \sim t(n-1)$ <br> - For paired-sample test, the difference is normally distributed. <br> - Pearson's statistic: $\chi^{2}=\sum_{i=1}^{n} \frac{\left(o_{i}-E_{i}\right)^{2}}{E_{i}} \sim \chi^{2}(v)$, provided that $E_{i} \geq 5$. <br> - Degree of freedom, $\mathrm{v}=$ number of classes - number of estimated parameters -1 for goodness of fit. <br> - Collapse whenever $\mathrm{E}_{\mathrm{i}}<5$. <br> - Degree of freedom, $v=($ no. of rows -1$) \times($ no. of columns -1$)$ for independence. |


|  | •Non-parametric tests test the median. The null hypotheses are to <br> claim that the median difference is 0. |
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| Non- <br> parametric <br> Tests | -Continuity correction: $\mathrm{P}(\mathrm{X}=\mathrm{k}) \Rightarrow \mathrm{P}\left(\mathrm{k}-\frac{1}{2}<\mathrm{X}<\mathrm{k}+\frac{1}{2}\right) ; \mathrm{P}(\mathrm{X}<$ <br> $\mathrm{k}) \Rightarrow \mathrm{P}\left(\mathrm{X}<\mathrm{k}+\frac{1}{2}\right)$ |

